



# CUSOLVER LIBRARY

DU-06709-001\_v11.0 | August 2020



# TABLE OF CONTENTS

<b>Chapter 1. Introduction.....</b>	<b>1</b>
1.1. cuSolverDN: Dense LAPACK.....	2
1.2. cuSolverSP: Sparse LAPACK.....	2
1.3. cuSolverRF: Refactorization.....	3
1.4. Naming Conventions.....	3
1.5. Asynchronous Execution.....	5
1.6. Library Property.....	5
1.7. high precision package.....	5
<b>Chapter 2. Using the CUSOLVER API.....</b>	<b>6</b>
2.1. General description.....	6
2.1.1. Thread Safety.....	6
2.1.2. Scalar Parameters.....	6
2.1.3. Parallelism with Streams.....	6
2.1.4. Link Third-party LAPACK Library.....	7
2.1.5. convention of info.....	7
2.1.6. usage of _bufferSize.....	7
2.2. cuSolver Types Reference.....	8
2.2.1. cuSolverDN Types.....	8
2.2.1.1. cusolverDnHandle_t.....	8
2.2.1.2. cublasFillMode_t.....	8
2.2.1.3. cublasOperation_t.....	8
2.2.1.4. cusolverEigType_t.....	8
2.2.1.5. cusolverEigMode_t.....	9
2.2.1.6. cusolverIRSRefinement_t.....	9
2.2.1.7. cusolverDnIRSParams_t.....	10
2.2.1.8. cusolverDnIRSInfos_t.....	10
2.2.1.9. cusolverDnFunction_t.....	10
2.2.1.10. cusolverAlgMode_t.....	10
2.2.1.11. cusolverStatus_t.....	11
2.2.2. cuSolverSP Types.....	11
2.2.2.1. cusolverSpHandle_t.....	11
2.2.2.2. cusparseMatDescr_t.....	11
2.2.2.3. cusolverStatus_t.....	11
2.2.3. cuSolverRF Types.....	12
2.2.3.1. cusolverRfHandle_t.....	12
2.2.3.2. cusolverRfMatrixFormat_t.....	12
2.2.3.3. cusolverRfNumericBoostReport_t.....	13
2.2.3.4. cusolverRfResetValuesFastMode_t.....	13
2.2.3.5. cusolverRfFactorization_t.....	13
2.2.3.6. cusolverRfTriangularSolve_t.....	13

2.2.3.7. cusolverRfUnitDiagonal_t.....	13
2.2.3.8. cusolverStatus_t.....	14
2.3. cuSolver Formats Reference.....	14
2.3.1. Index Base Format.....	14
2.3.2. Vector (Dense) Format.....	14
2.3.3. Matrix (Dense) Format.....	14
2.3.4. Matrix (CSR) Format.....	15
2.3.5. Matrix (CSC) Format.....	16
2.4. cuSolverDN: dense LAPACK Function Reference.....	17
2.4.1. cuSolverDN Helper Function Reference.....	17
2.4.1.1. cusolverDnCreate().....	17
2.4.1.2. cusolverDnDestroy().....	17
2.4.1.3. cusolverDnSetStream().....	18
2.4.1.4. cusolverDnGetStream().....	18
2.4.1.5. cusolverDnCreateSyevjInfo().....	18
2.4.1.6. cusolverDnDestroySyevjInfo().....	19
2.4.1.7. cusolverDnXsyevjSetTolerance().....	19
2.4.1.8. cusolverDnXsyevjSetMaxSweeps().....	19
2.4.1.9. cusolverDnXsyevjSetSortEig().....	20
2.4.1.10. cusolverDnXsyevjGetResidual().....	20
2.4.1.11. cusolverDnXsyevjGetSweeps().....	20
2.4.1.12. cusolverDnCreateGesvdjInfo().....	21
2.4.1.13. cusolverDnDestroyGesvdjInfo().....	21
2.4.1.14. cusolverDnXgesvdjSetTolerance().....	22
2.4.1.15. cusolverDnXgesvdjSetMaxSweeps().....	22
2.4.1.16. cusolverDnXgesvdjSetSortEig().....	22
2.4.1.17. cusolverDnXgesvdjGetResidual().....	23
2.4.1.18. cusolverDnXgesvdjGetSweeps().....	23
2.4.1.19. cusolverDnIRSPParamsCreate().....	24
2.4.1.20. cusolverDnIRSPParamsDestroy().....	24
2.4.1.21. cusolverDnIRSPParamsSetSolverPrecisions().....	24
2.4.1.22. cusolverDnIRSPParamsSetSolverMainPrecision().....	26
2.4.1.23. cusolverDnIRSPParamsSetSolverLowestPrecision().....	26
2.4.1.24. cusolverDnIRSPParamsSetRefinementSolver().....	27
2.4.1.25. cusolverDnIRSPParamsSetTol().....	28
2.4.1.26. cusolverDnIRSPParamsSetTolInner().....	29
2.4.1.27. cusolverDnIRSPParamsSetMaxIters().....	30
2.4.1.28. cusolverDnIRSPParamsSetMaxItersInner().....	30
2.4.1.29. cusolverDnIRSPParamsEnableFallback().....	31
2.4.1.30. cusolverDnIRSPParamsDisableFallback().....	31
2.4.1.31. cusolverDnIRSPParamsGetMaxIters().....	32
2.4.1.32. cusolverDnIRSInfosCreate().....	33
2.4.1.33. cusolverDnIRSInfosDestroy().....	33

2.4.1.34. cusolverDnIRSInfosGetMaxIters()	33
2.4.1.35. cusolverDnIRSInfosGetNiters()	34
2.4.1.36. cusolverDnIRSInfosGetOuterNiters()	34
2.4.1.37. cusolverDnIRSInfosRequestResidual()	35
2.4.1.38. cusolverDnIRSInfosGetResidualHistory()	35
2.4.1.39. cusolverDnCreateParams()	37
2.4.1.40. cusolverDnDestroyParams()	37
2.4.1.41. cusolverDnSetAdvOptions()	37
2.4.2. Dense Linear Solver Reference	38
2.4.2.1. cusolverDn<t>potrf()	38
2.4.2.2. cusolverDnPotrf()	41
2.4.2.3. cusolverDn<t>potrs()	44
2.4.2.4. cusolverDnPotrs()	46
2.4.2.5. cusolverDn<t>potri()	48
2.4.2.6. cusolverDn<t>getrf()	51
2.4.2.7. cusolverDnGetrf()	53
2.4.2.8. cusolverDn<t>getrs()	56
2.4.2.9. cusolverDnGetrs()	58
2.4.2.10. cusolverDn<t1><t2>gesv()	59
2.4.2.11. cusolverDnIRSXgesv()	66
2.4.2.12. cusolverDn<t>geqrf()	72
2.4.2.13. cusolverDnGeqrf()	74
2.4.2.14. cusolverDn<t1><t2>gels()	76
2.4.2.15. cusolverDnIRSXgels()	83
2.4.2.16. cusolverDn<t>ormqr()	89
2.4.2.17. cusolverDn<t>orgqr()	93
2.4.2.18. cusolverDn<t>sytrf()	96
2.4.2.19. cusolverDn<t>potrfBatched()	99
2.4.2.20. cusolverDn<t>potrsBatched()	101
2.4.3. Dense Eigenvalue Solver Reference	103
2.4.3.1. cusolverDn<t>gebrd()	103
2.4.3.2. cusolverDn<t>orgbr()	107
2.4.3.3. cusolverDn<t>sytrd()	111
2.4.3.4. cusolverDn<t>ormtr()	115
2.4.3.5. cusolverDn<t>orgtr()	119
2.4.3.6. cusolverDn<t>gesvd()	122
2.4.3.7. cusolverDnGesvd()	127
2.4.3.8. cusolverDn<t>gesvdj()	131
2.4.3.9. cusolverDn<t>gesvdjBatched()	136
2.4.3.10. cusolverDn<t>gesvdaStridedBatched()	141
2.4.3.11. cusolverDn<t>syevd()	147
2.4.3.12. cusolverDnSyevd()	150
2.4.3.13. cusolverDn<t>syevdx()	154

2.4.3.14. cusolverDnSyevedx()	159
2.4.3.15. cusolverDn<t>sygvd()	164
2.4.3.16. cusolverDn<t>sygvdx()	169
2.4.3.17. cusolverDn<t>syevj()	175
2.4.3.18. cusolverDn<t>sygvj()	180
2.4.3.19. cusolverDn<t>syevjBatched()	186
2.5. cuSolverSP: sparse LAPACK Function Reference	190
2.5.1. Helper Function Reference	190
2.5.1.1. cusolverSpCreate()	191
2.5.1.2. cusolverSpDestroy()	191
2.5.1.3. cusolverSpSetStream()	191
2.5.1.4. cusolverSpXcsrissym()	192
2.5.2. High Level Function Reference	193
2.5.2.1. cusolverSp<t>csrslvlu()	194
2.5.2.2. cusolverSp<t>csrslvqr()	197
2.5.2.3. cusolverSp<t>csrslsvchol()	200
2.5.2.4. cusolverSp<t>csrslsvqr()	203
2.5.2.5. cusolverSp<t>csreigvsi()	207
2.5.2.6. cusolverSp<t>csreigs()	211
2.5.3. Low Level Function Reference	213
2.5.3.1. cusolverSpXcsrsmrcm()	213
2.5.3.2. cusolverSpXcsrsmmdq()	215
2.5.3.3. cusolverSpXcsrsmamd()	216
2.5.3.4. cusolverSpXcsrmetisnd()	218
2.5.3.5. cusolverSpXcsrzfd()	220
2.5.3.6. cusolverSpXcsrperm()	222
2.5.3.7. cusolverSpXcsrqrBatched()	224
2.6. cuSolverRF: Refactorization Reference	232
2.6.1. cusolverRfAccessBundledFactors()	232
2.6.2. cusolverRfAnalyze()	233
2.6.3. cusolverRfSetupDevice()	234
2.6.4. cusolverRfSetupHost()	236
2.6.5. cusolverRfCreate()	238
2.6.6. cusolverRfExtractBundledFactorsHost()	239
2.6.7. cusolverRfExtractSplitFactorsHost()	240
2.6.8. cusolverRfDestroy()	241
2.6.9. cusolverRfGetMatrixFormat()	241
2.6.10. cusolverRfGetNumericProperties()	242
2.6.11. cusolverRfGetNumericBoostReport()	242
2.6.12. cusolverRfGetResetValuesFastMode()	243
2.6.13. cusolverRfGet_Algs()	243
2.6.14. cusolverRfRefactor()	243
2.6.15. cusolverRfResetValues()	244

2.6.16. cusolverRfSetMatrixFormat()	245
2.6.17. cusolverRfSetNumericProperties()	246
2.6.18. cusolverRfSetResetValuesFastMode()	246
2.6.19. cusolverRfSetAlgs()	247
2.6.20. cusolverRfSolve()	247
2.6.21. cusolverRfBatchSetupHost()	249
2.6.22. cusolverRfBatchAnalyze()	251
2.6.23. cusolverRfBatchResetValues()	252
2.6.24. cusolverRfBatchRefactor()	253
2.6.25. cusolverRfBatchSolve()	254
2.6.26. cusolverRfBatchZeroPivot()	255
<b>Chapter 3. Using the CUSOLVERMG API</b>	<b>257</b>
3.1. General description	257
3.1.1. Thread Safety	257
3.1.2. Determinism	257
3.1.3. tile strategy	257
3.1.4. Global matrix versus local matrix	259
3.1.5. usage of _bufferSize	259
3.1.6. synchronization	260
3.1.7. context switch	260
3.1.8. NVLINK	260
3.2. cuSolverMG Types Reference	260
3.2.1. cuSolverMG Types	260
3.2.2. cusolverMgHandle_t	260
3.2.3. cusolverMgGridMapping_t	260
3.2.4. cudaLibMgGrid_t	261
3.2.5. cudaLibMgMatrixDesc_t	261
3.3. Helper Function Reference	261
3.3.1. cusolverMgCreate()	261
3.3.2. cusolverMgDestroy()	261
3.3.3. cusolverMgDeviceSelect()	262
3.3.4. cusolverMgCreateDeviceGrid()	262
3.3.5. cusolverMgDestroyGrid()	263
3.3.6. cusolverMgCreateMatDescr()	263
3.3.7. cusolverMgDestroyMatrixDescr()	264
3.4. Dense Linear Solver Reference	264
3.4.1. cusolverMgPotrf()	264
3.4.2. cusolverMgPotrs()	267
3.4.3. cusolverMgPotri()	270
3.4.4. cusolverMgGetrf()	272
3.4.5. cusolverMgGetrs()	275
3.5. Dense Eigenvalue Solver Reference	277
3.5.1. cusolverMgSyevd()	278

<b>Appendix A. cuSolverRF Examples.....</b>	<b>281</b>
A.1. cuSolverRF In-memory Example.....	281
A.2. cuSolverRF-batch Example.....	285
<b>Appendix B. CSR QR Batch Examples.....</b>	<b>289</b>
B.1. Batched Sparse QR example 1.....	289
B.2. Batched Sparse QR example 2.....	293
<b>Appendix C. QR Examples.....</b>	<b>299</b>
C.1. QR Factorization Dense Linear Solver.....	299
C.2. orthogonalization.....	304
<b>Appendix D. LU Examples.....</b>	<b>310</b>
D.1. LU Factorization.....	310
<b>Appendix E. Cholesky Examples.....</b>	<b>315</b>
E.1. batched Cholesky Factorization.....	315
<b>Appendix F. Examples of Dense Eigenvalue Solver.....</b>	<b>320</b>
F.1. Standard Symmetric Dense Eigenvalue Solver.....	320
F.2. Standard Symmetric Dense Eigenvalue Solver.....	323
F.3. Generalized Symmetric-Definite Dense Eigenvalue Solver.....	326
F.4. Generalized Symmetric-Definite Dense Eigenvalue Solver.....	329
F.5. Standard Symmetric Dense Eigenvalue Solver (via Jacobi method).....	333
F.6. Generalized Symmetric-Definite Dense Eigenvalue Solver (via Jacobi method).....	337
F.7. batch eigenvalue solver for dense symmetric matrix.....	342
<b>Appendix G. Examples of Singular Value Decomposition.....</b>	<b>348</b>
G.1. SVD with singular vectors.....	348
G.2. SVD with singular vectors (via Jacobi method).....	352
G.3. batch dense SVD solver.....	357
G.4. SVD approximation.....	363
G.5. 64-bit GESVD.....	367
<b>Appendix H. Examples of multiGPU eigenvalue solver.....</b>	<b>373</b>
H.1. SYEVD of 1D Laplacian operator (example 1).....	374
H.2. SYEVD of 1D Laplacian operator (example 2).....	386
H.3. SYEVD of 1D Laplacian operator (example 3).....	391
<b>Appendix I. Examples of multiGPU linear solver.....</b>	<b>396</b>
I.1. GETRF and GETRS of 1D Laplacian operator (example 1).....	397
I.2. POTRF and POTRS of 1D Laplacian operator (example 2).....	405
I.3. POTRF and POTRI of 1D Laplacian operator (example 3).....	414
<b>Appendix J. Acknowledgements.....</b>	<b>423</b>
<b>Appendix K. Bibliography.....</b>	<b>426</b>

## LIST OF FIGURES

Figure 1 Example of cusolveMG tiling for 3 Gpus .....	258
Figure 2 global matrix and local matrix .....	259

# Chapter 1.

## INTRODUCTION

The cuSolver library is a high-level package based on the cuBLAS and cuSPARSE libraries. It consists of two modules corresponding to two sets of API:

1. The cuSolver API on a single GPU
2. The cuSolverMG API on a single node multiGPU

Each of which can be used independently or in concert with other toolkit libraries. To simplify the notation, cuSolver denotes single GPU API and cuSolverMg denotes multiGPU API.

The intent of cuSolver is to provide useful LAPACK-like features, such as common matrix factorization and triangular solve routines for dense matrices, a sparse least-squares solver and an eigenvalue solver. In addition cuSolver provides a new refactorization library useful for solving sequences of matrices with a shared sparsity pattern.

cuSolver combines three separate components under a single umbrella. The first part of cuSolver is called cuSolverDN, and deals with dense matrix factorization and solve routines such as LU, QR, SVD and LDLT, as well as useful utilities such as matrix and vector permutations.

Next, cuSolverSP provides a new set of sparse routines based on a sparse QR factorization. Not all matrices have a good sparsity pattern for parallelism in factorization, so the cuSolverSP library also provides a CPU path to handle those sequential-like matrices. For those matrices with abundant parallelism, the GPU path will deliver higher performance. The library is designed to be called from C and C++.

The final part is cuSolverRF, a sparse re-factorization package that can provide very good performance when solving a sequence of matrices where only the coefficients are changed but the sparsity pattern remains the same.

The GPU path of the cuSolver library assumes data is already in the device memory. It is the responsibility of the developer to allocate memory and to copy data between GPU memory and CPU memory using standard CUDA runtime API routines, such as `cudaMalloc()`, `cudaFree()`, `cudaMemcpy()`, and `cudaMemcpyAsync()`.

cuSolverMg is GPU-accelerated ScaLAPACK. By now, cuSolverMg supports 1-D column block cyclic layout and provides symmetric eigenvalue solver.



The cuSolver library requires hardware with a CUDA compute capability (CC) of at least 2.0 or higher. Please see the *CUDA C++ Programming Guide* for a list of the compute capabilities corresponding to all NVIDIA GPUs.

## 1.1. cuSolverDN: Dense LAPACK

The cuSolverDN library was designed to solve dense linear systems of the form

$$Ax = b$$

where the coefficient matrix  $A \in \mathbb{R}^{n \times n}$ , right-hand-side vector  $b \in \mathbb{R}^n$  and solution vector  $x \in \mathbb{R}^n$

The cuSolverDN library provides QR factorization and LU with partial pivoting to handle a general matrix  $\mathbf{A}$ , which may be non-symmetric. Cholesky factorization is also provided for symmetric/Hermitian matrices. For symmetric indefinite matrices, we provide Bunch-Kaufman (LDL) factorization.

The cuSolverDN library also provides a helpful bidiagonalization routine and singular value decomposition (SVD).

The cuSolverDN library targets computationally-intensive and popular routines in LAPACK, and provides an API compatible with LAPACK. The user can accelerate these time-consuming routines with cuSolverDN and keep others in LAPACK without a major change to existing code.

## 1.2. cuSolverSP: Sparse LAPACK

The cuSolverSP library was mainly designed to solve sparse linear system

$$Ax = b$$

and the least-squares problem

$$x = \operatorname{argmin} \|A^*z - b\|$$

where sparse matrix  $A \in \mathbb{R}^{m \times n}$ , right-hand-side vector  $b \in \mathbb{R}^m$  and solution vector  $x \in \mathbb{R}^n$ . For a linear system, we require  $m=n$ .

The core algorithm is based on sparse QR factorization. The matrix  $\mathbf{A}$  is accepted in CSR format. If matrix  $\mathbf{A}$  is symmetric/Hermitian, the user has to provide a full matrix, ie fill missing lower or upper part.

If matrix  $\mathbf{A}$  is symmetric positive definite and the user only needs to solve  $Ax = b$ , Cholesky factorization can work and the user only needs to provide the lower triangular part of  $\mathbf{A}$ .

On top of the linear and least-squares solvers, the **cuSolverSP** library provides a simple eigenvalue solver based on shift-inverse power method, and a function to count the number of eigenvalues contained in a box in the complex plane.

## 1.3. cuSolverRF: Refactorization

The cuSolverRF library was designed to accelerate solution of sets of linear systems by fast re-factorization when given new coefficients in the same sparsity pattern

$$A_i x_i = f_i$$

where a sequence of coefficient matrices  $A_i \in R^{n \times n}$ , right-hand-sides  $f_i \in R^n$  and solutions  $x_i \in R^n$  are given for  $i=1, \dots, k$ .

The cuSolverRF library is applicable when the sparsity pattern of the coefficient matrices  $A_i$  as well as the reordering to minimize fill-in and the pivoting used during the LU factorization remain the same across these linear systems. In that case, the first linear system ( $i=1$ ) requires a full LU factorization, while the subsequent linear systems ( $i=2, \dots, k$ ) require only the LU re-factorization. The later can be performed using the cuSolverRF library.

Notice that because the sparsity pattern of the coefficient matrices, the reordering and pivoting remain the same, the sparsity pattern of the resulting triangular factors  $L_i$  and  $U_i$  also remains the same. Therefore, the real difference between the full LU factorization and LU re-factorization is that the required memory is known ahead of time.

## 1.4. Naming Conventions

The cuSolverDN library provides two different APIs; **legacy** and **generic**.

The functions in the legacy API are available for data types **float**, **double**, **cuComplex**, and **cuDoubleComplex**. The naming convention for the legacy API is as follows:

`cusolverDn<t><operation>`

where `<t>` can be **S**, **D**, **C**, **Z**, or **X**, corresponding to the data types **float**, **double**, **cuComplex**, **cuDoubleComplex**, and the generic type, respectively. `<operation>` can be Cholesky factorization (**potrf**), LU with partial pivoting (**getrf**), QR factorization (**geqrf**) and Bunch-Kaufman factorization (**sytrf**).

The functions in the generic API provide a single entry point for each routine and support for 64-bit integers to define matrix and vector dimensions. The naming convention for the generic API is data-agnostic and is as follows:

`cusolverDn<operation>`

where `<operation>` can be Cholesky factorization (**potrf**), LU with partial pivoting (**getrf**) and QR factorization (**geqrf**).

The cuSolverSP library functions are available for data types **float**, **double**, **cuComplex**, and **cuDoubleComplex**. The naming convention is as follows:

```
cusolverSp[Host]<t>[<matrix data
format>]<operation>[<output matrix data format>]<based on>
```

where **cuSolverSp** is the GPU path and **cusolverSpHost** is the corresponding CPU path. **<t>** can be **S**, **D**, **C**, **Z**, or **X**, corresponding to the data types **float**, **double**, **cuComplex**, **cuDoubleComplex**, and the generic type, respectively.

The **<matrix data format>** is **csr**, compressed sparse row format.

The **<operation>** can be **ls**, **lsq**, **eig**, **eigs**, corresponding to linear solver, least-square solver, eigenvalue solver and number of eigenvalues in a box, respectively.

The **<output matrix data format>** can be **v** or **m**, corresponding to a vector or a matrix.

**<based on>** describes which algorithm is used. For example, **qr** (sparse QR factorization) is used in linear solver and least-square solver.

All of the functions have the return type **cusolverStatus\_t** and are explained in more detail in the chapters that follow.

#### cuSolverSP API

routine	data format	operation	output format	based on
csrlsvlu	csr	linear solver (ls)	vector (v)	LU (lu) with partial pivoting
csrlsvqr	csr	linear solver (ls)	vector (v)	QR factorization (qr)
csrlsvchol	csr	linear solver (ls)	vector (v)	Cholesky factorization (chol)
csrlsqvqr	csr	least-square solver (lsq)	vector (v)	QR factorization (qr)
csreigvsi	csr	eigenvalue solver (eig)	vector (v)	shift-inverse
csreigs	csr	number of eigenvalues in a box (eigs)		
csrsymrcm	csr	Symmetric Reverse Cuthill-McKee (symrcm)		

The cuSolverRF library routines are available for data type **double**. Most of the routines follow the naming convention:

```
cusolverRf_<operation>_[[Host]](...)
```

where the trailing optional Host qualifier indicates the data is accessed on the host versus on the device, which is the default. The <operation> can be **Setup**, **Analyze**, **Refactor**, **Solve**, **ResetValues**, **AccessBundledFactors** and **ExtractSplitFactors**.

Finally, the return type of the cuSolverRF library routines is **cusolverStatus\_t**.

## 1.5. Asynchronous Execution

The cuSolver library functions prefer to keep asynchronous execution as much as possible. Developers can always use the **cudaDeviceSynchronize()** function to ensure that the execution of a particular cuSolver library routine has completed.

A developer can also use the **cudaMemcpy()** routine to copy data from the device to the host and vice versa, using the **cudaMemcpyDeviceToHost** and **cudaMemcpyHostToDevice** parameters, respectively. In this case there is no need to add a call to **cudaDeviceSynchronize()** because the call to **cudaMemcpy()** with the above parameters is blocking and completes only when the results are ready on the host.

## 1.6. Library Property

The **libraryPropertyType** data type is an enumeration of library property types. (ie. CUDA version X.Y.Z would yield **MAJOR\_VERSION=X**, **MINOR\_VERSION=Y**, **PATCH\_LEVEL=Z**)

```
typedef enum libraryPropertyType_t
{
    MAJOR_VERSION,
    MINOR_VERSION,
    PATCH_LEVEL
} libraryPropertyType;
```

The following code can show the version of cusolver library.

```
int major=-1,minor=-1,patch=-1;
cusolverGetProperty(MAJOR_VERSION, &major);
cusolverGetProperty(MINOR_VERSION, &minor);
cusolverGetProperty(PATCH_LEVEL, &patch);
printf("CUSOLVER Version (Major,Minor,PatchLevel): %d.%d.%d\n",
major,minor,patch);
```

## 1.7. high precision package

The **cusolver** library uses high precision for iterative refinement when necessary.

# Chapter 2.

## USING THE CUSOLVER API

### 2.1. General description

This chapter describes how to use the cuSolver library API. It is not a reference for the cuSolver API data types and functions; that is provided in subsequent chapters.

#### 2.1.1. Thread Safety

The library is thread-safe, and its functions can be called from multiple host threads.

#### 2.1.2. Scalar Parameters

In the cuSolver API, the scalar parameters can be passed by reference on the host.

#### 2.1.3. Parallelism with Streams

If the application performs several small independent computations, or if it makes data transfers in parallel with the computation, then CUDA streams can be used to overlap these tasks.

The application can conceptually associate a stream with each task. To achieve the overlap of computation between the tasks, the developer should:

1. Create CUDA streams using the function `cudaStreamCreate()`, and
2. Set the stream to be used by each individual cuSolver library routine by calling, for example, `cusolverDnSetStream()`, just prior to calling the actual cuSolverDN routine.

The computations performed in separate streams would then be overlapped automatically on the GPU, when possible. This approach is especially useful when the computation performed by a single task is relatively small, and is not enough to fill the GPU with work, or when there is a data transfer that can be performed in parallel with the computation.

## 2.1.4. Link Third-party LAPACK Library

Starting with CUDA 10.1 update 2, NVIDIA LAPACK library `liblapack_static.a` is a subset of LAPACK and only contains GPU accelerated `stedc` and `bdsqr`. The user has to link `libcusolver_static.a` with `liblapack_static.a` in order to build the application successfully. Prior to CUDA 10.1 update 2, the user can replace `liblapack_static.a` with a third-party LAPACK library, for example, MKL. In CUDA 10.1 update 2, the third-party LAPACK library no longer affects the behavior of cusolver library, neither functionality nor performance. Furthermore the user cannot use `liblapack_static.a` as a standalone LAPACK library because it is only a subset of LAPACK.



The `liblapack_static.a` library, which is the binary of CLAPACK-3.2.1, is a new feature of CUDA 10.0.

- ▶ If you use `libcusolver_static.a`, then you must link with `liblapack_static.a` explicitly, otherwise the linker will report missing symbols. No conflict of symbols between `liblapack_static.a` and other third-party LAPACK library, you are free to link the latter to your application.
- ▶ The `liblapack_static.a` is built inside `libcusolver.so`. Hence, if you use `libcusolver.so`, then you don't need to specify a LAPACK library. The `libcusolver.so` will not pick up any routines from the third-party LAPACK library even you link the application with it.

## 2.1.5. convention of info

Each LAPACK routine returns an `info` which indicates the position of invalid parameter. If `info = -i`, then *i*-th parameter is invalid. To be consistent with base-1 in LAPACK, `cusolver` does not report invalid `handle` into `info`. Instead, `cusolver` returns `CUSOLVER_STATUS_NOT_INITIALIZED` for invalid `handle`.

## 2.1.6. usage of `_bufferSize`

There is no `cudaMalloc` inside `cuSolver` library, the user must allocate the device workspace explicitly. The routine `xyz_bufferSize` is to query the size of workspace of the routine `xyz`, for example `xyz = potrf`. To make the API simple, `xyz_bufferSize` follows almost the same signature of `xyz` even it only depends on some parameters, for example, device pointer is not used to decide the size of workspace. In most cases, `xyz_bufferSize` is called in the beginning before actual device data (pointing by a device pointer) is prepared or before the device pointer is allocated. In such case, the user can pass null pointer to `xyz_bufferSize` without breaking the functionality.

## 2.2. cuSolver Types Reference

### 2.2.1. cuSolverDN Types

The `float`, `double`, `cuComplex`, and `cuDoubleComplex` data types are supported. The first two are standard C data types, while the last two are exported from `cuComplex.h`. In addition, cuSolverDN uses some familiar types from cuBlas.

#### 2.2.1.1. cusolverDnHandle\_t

This is a pointer type to an opaque cuSolverDN context, which the user must initialize by calling `cusolverDnCreate()` prior to calling any other library function. An un-initialized Handle object will lead to unexpected behavior, including crashes of cuSolverDN. The handle created and returned by `cusolverDnCreate()` must be passed to every cuSolverDN function.

#### 2.2.1.2. cublasFillMode\_t

The type indicates which part (lower or upper) of the dense matrix was filled and consequently should be used by the function. Its values correspond to Fortran characters `'L'` or `'l'` (lower) and `'U'` or `'u'` (upper) that are often used as parameters to legacy BLAS implementations.

Value	Meaning
<code>CUBLAS_FILL_MODE_LOWER</code>	the lower part of the matrix is filled
<code>CUBLAS_FILL_MODE_UPPER</code>	the upper part of the matrix is filled

#### 2.2.1.3. cublasOperation\_t

The `cublasOperation_t` type indicates which operation needs to be performed with the dense matrix. Its values correspond to Fortran characters `'N'` or `'n'` (non-transpose), `'T'` or `'t'` (transpose) and `'C'` or `'c'` (conjugate transpose) that are often used as parameters to legacy BLAS implementations.

Value	Meaning
<code>CUBLAS_OP_N</code>	the non-transpose operation is selected
<code>CUBLAS_OP_T</code>	the transpose operation is selected
<code>CUBLAS_OP_C</code>	the conjugate transpose operation is selected

#### 2.2.1.4. cusolverEigType\_t

The `cusolverEigType_t` type indicates which type of eigenvalue solver is. Its values correspond to Fortran integer `1` ( $A*x = \lambda*B*x$ ), `2` ( $A*B*x = \lambda*x$ ), `3` ( $B*A*x = \lambda*x$ ), used as parameters to legacy LAPACK implementations.

Value	Meaning
-------	---------

CUSOLVER_EIG_TYPE_1	$A^*x = \text{lambda} * B^*x$
CUSOLVER_EIG_TYPE_2	$A^*B^*x = \text{lambda} * x$
CUSOLVER_EIG_TYPE_3	$B^*A^*x = \text{lambda} * x$

### 2.2.1.5. cusolverEigMode\_t

The `cusolverEigMode_t` type indicates whether or not eigenvectors are computed. Its values correspond to Fortran character 'N' (only eigenvalues are computed), 'V' (both eigenvalues and eigenvectors are computed) used as parameters to legacy LAPACK implementations.

Value	Meaning
CUSOLVER_EIG_MODE_NOVECTOR	only eigenvalues are computed
CUSOLVER_EIG_MODE_VECTOR	both eigenvalues and eigenvectors are computed

### 2.2.1.6. cusolverIRSRefinement\_t

The `cusolverIRSRefinement_t` type indicates which solver type would be used for the specific cusolver function. Most of our experimentation shows that CUSOLVER\_IRS\_REFINE\_GMRES is the best option.

*More details about the refinement process can be found in Azzam Haidar, Stanimire Tomov, Jack Dongarra, and Nicholas J. Higham. 2018. Harnessing GPU tensor cores for fast FP16 arithmetic to speed up mixed-precision iterative refinement solvers. In Proceedings of the International Conference for High Performance Computing, Networking, Storage, and Analysis (SC '18). IEEE Press, Piscataway, NJ, USA, Article 47, 11 pages.*

Value	Meaning
CUSOLVER_IRS_REFINE_NOT_SET	Solver is not set, this value is what is set when creating the params structure. IRS solver will return error.
CUSOLVER_IRS_REFINE_NONE	No refinement solver, the IRS solver performs a factorisation followed by a solve without any refinement. For example if the IRS solver was <code>cusolverDnIRSXgesv()</code> , this is equivalent to a <code>Xgesv</code> routine without refinement and where the factorisation is carried out in the lowest precision. If for example the main precision was CUSOLVER_R_64F and the lowest was CUSOLVER_R_64F as well, then this is equivalent to a call to <code>cusolverDnDgesv()</code> .
CUSOLVER_IRS_REFINE_CLASSICAL	Classical iterative refinement solver. Similar to the one used in LAPACK routines.
CUSOLVER_IRS_REFINE_GMRES	GMRES (Generalized Minimal Residual) based iterative refinement solver. In recent study, the GMRES method has drawn the scientific community attention for its ability to be used as refinement solver that outperforms the classical iterative refinement method. based on our experimentation, we recommend this setting.

CUSOLVER_IRS_REFINE_CLASSICAL_GMRES	Classical iterative refinement solver that uses the GMRES (Generalized Minimal Residual) internally to solve the correction equation at each iteration. We call the <i>classical refinement iteration</i> the outer iteration while the <i>GMRES</i> is called inner iteration. Note that if the tolerance of the inner GMRES is set very low, let say to machine precision, then the outer <i>classical refinement iteration</i> will performs only one iteration and thus this option will behaves like CUSOLVER_IRS_REFINE_GMRES.
CUSOLVER_IRS_REFINE_GMRES_GMRES	Similar to CUSOLVER_IRS_REFINE_CLASSICAL_GMRES which consists of classical refinement process that uses GMRES to solve the inner correction system, here it is a GMRES (Generalized Minimal Residual) based iterative refinement solver that uses another GMRES internally to solve the preconditioned system.

### 2.2.1.7. cusolverDnIRSParams\_t

This is a pointer type to an opaque `cusolverDnIRSParams_t` structure, which holds parameters for the iterative refinement linear solvers such as `cusolverDnXgesv()`. Use corresponding helper functions described below to either Create/Destroy this structure or Set/Get solver parameters.

### 2.2.1.8. cusolverDnIRSInfos\_t

This is a pointer type to an opaque `cusolverDnIRSInfos_t` structure, which holds information about the performed call to an iterative refinement linear solver (e.g., `cusolverDnXgesv()`). Use corresponding helper functions described below to either Create/Destroy this structure or retrieve solve information.

### 2.2.1.9. cusolverDnFunction\_t

The `cusolverDnFunction_t` type indicates which routine needs to be configured by `cusolverDnSetAdvOptions()`. The value `CUSOLVERDN_GETRF` corresponds to the routine `Getrf`.

Value	Meaning
CUSOLVERDN_GETRF	corresponds to <code>Getrf</code>

### 2.2.1.10. cusolverAlgMode\_t

The `cusolverAlgMode_t` type indicates which algorithm is selected by `cusolverDnSetAdvOptions()`. The set of algorithms supported for each routine is described in detail along with the routine documentation.

The default algorithm is `CUSOLVER_ALG_0`. The user can also provide `NULL` to use the default algorithm.

### 2.2.1.11. cusolverStatus\_t

This is the same as `cusolverStatus_t` in the sparse LAPACK section.

## 2.2.2. cuSolverSP Types

The `float`, `double`, `cuComplex`, and `cuDoubleComplex` data types are supported. The first two are standard C data types, while the last two are exported from `cuComplex.h`.

### 2.2.2.1. cusolverSpHandle\_t

This is a pointer type to an opaque cuSolverSP context, which the user must initialize by calling `cusolverSpCreate()` prior to calling any other library function. An un-initialized Handle object will lead to unexpected behavior, including crashes of cuSolverSP. The handle created and returned by `cusolverSpCreate()` must be passed to every cuSolverSP function.

### 2.2.2.2. cusparseMatDescr\_t

We have chosen to keep the same structure as exists in cuSparse to describe the shape and properties of a matrix. This enables calls to either cuSparse or cuSolver using the same matrix description.

```
typedef struct {
    cusparseMatrixType_t MatrixType;
    cusparseFillMode_t FillMode;
    cusparseDiagType_t DiagType;
    cusparseIndexBase_t IndexBase;
} cusparseMatDescr_t;
```

Please read documentation of CUSPARSE Library to understand each field of `cusparseMatDescr_t`.

### 2.2.2.3. cusolverStatus\_t

This is a status type returned by the library functions and it can have the following values.

<code>CUSOLVER_STATUS_SUCCESS</code>	The operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INIT:</code>	The cuSolver library was not initialized. This is usually caused by the lack of a prior call, an error in the CUDA Runtime API called by the cuSolver routine, or an error in the hardware setup.  <b>To correct:</b> call <code>cusolverCreate()</code> prior to the function call; and check that the hardware, an appropriate version of the driver, and the cuSolver library are correctly installed.
<code>CUSOLVER_STATUS_ALLOC_FA:</code>	Resource allocation failed inside the cuSolver library. This is usually caused by a <code>cudaMalloc()</code> failure.  <b>To correct:</b> prior to the function call, deallocate previously allocated memory as much as possible.

<code>CUSOLVER_STATUS_INVALID_*</code>	<p>An unsupported value or parameter was passed to the function (a negative vector size, for example).</p> <p><b>To correct:</b> ensure that all the parameters being passed have valid values.</p>
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	<p>The function requires a feature absent from the device architecture; usually caused by the lack of support for atomic operations or double precision.</p> <p><b>To correct:</b> compile and run the application on a device with compute capability 2.0 or above.</p>
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	<p>The GPU program failed to execute. This is often caused by a launch failure of the kernel on the GPU, which can be caused by multiple reasons.</p> <p><b>To correct:</b> check that the hardware, an appropriate version of the driver, and the cuSolver library are correctly installed.</p>
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	<p>An internal cuSolver operation failed. This error is usually caused by a <code>cudaMemcpyAsync()</code> failure.</p> <p><b>To correct:</b> check that the hardware, an appropriate version of the driver, and the cuSolver library are correctly installed. Also, check that the memory passed as a parameter to the routine is not being deallocated prior to the routine's completion.</p>
<code>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code>	<p>The matrix type is not supported by this function. This is usually caused by passing an invalid matrix descriptor to the function.</p> <p><b>To correct:</b> check that the fields in <code>descrA</code> were set correctly.</p>

## 2.2.3. cuSolverRF Types

cuSolverRF only supports **double**.

### 2.2.3.1. cusolverRfHandle\_t

The `cusolverRfHandle_t` is a pointer to an opaque data structure that contains the cuSolverRF library handle. The user must initialize the handle by calling `cusolverRfCreate()` prior to any other cuSolverRF library calls. The handle is passed to all other cuSolverRF library calls.

### 2.2.3.2. cusolverRfMatrixFormat\_t

The `cusolverRfMatrixFormat_t` is an enum that indicates the input/output matrix format assumed by the `cusolverRfSetupDevice()`, `cusolverRfSetupHost()`, `cusolverRfResetValues()`, `cusolverRfExtractBundledFactorsHost()` and `cusolverRfExtractSplitFactorsHost()` routines.

Value	Meaning
<code>CUSOLVER_MATRIX_FORMAT_CSR</code>	matrix format CSR is assumed. (default)

CUSOLVER_MATRIX_FORMAT_CSC	matrix format CSC is assumed.
----------------------------	-------------------------------

### 2.2.3.3. cusolverRfNumericBoostReport\_t

The `cusolverRfNumericBoostReport_t` is an enum that indicates whether numeric boosting (of the pivot) was used during the `cusolverRfRefactor()` and `cusolverRfSolve()` routines. The numeric boosting is disabled by default.

Value	Meaning
CUSOLVER_NUMERIC_BOOST_NOT_USED	numeric boosting not used. (default)
CUSOLVER_NUMERIC_BOOST_USED	numeric boosting used.

### 2.2.3.4. cusolverRfResetValuesFastMode\_t

The `cusolverRfResetValuesFastMode_t` is an enum that indicates the mode used for the `cusolverRfResetValues()` routine. The fast mode requires extra memory and is recommended only if very fast calls to `cusolverRfResetValues()` are needed.

Value	Meaning
CUSOLVER_RESET_VALUES_FAST_MODE_OFF	fast mode disabled. (default)
CUSOLVER_RESET_VALUES_FAST_MODE_ON	fast mode enabled.

### 2.2.3.5. cusolverRfFactorization\_t

The `cusolverRfFactorization_t` is an enum that indicates which (internal) algorithm is used for refactorization in the `cusolverRfRefactor()` routine.

Value	Meaning
CUSOLVER_FACTORIZATION_ALG0	algorithm 0. (default)
CUSOLVER_FACTORIZATION_ALG1	algorithm 1.
CUSOLVER_FACTORIZATION_ALG2	algorithm 2. Domino-based scheme.

### 2.2.3.6. cusolverRfTriangularSolve\_t

The `cusolverRfTriangularSolve_t` is an enum that indicates which (internal) algorithm is used for triangular solve in the `cusolverRfSolve()` routine.

Value	Meaning
CUSOLVER_TRIANGULAR_SOLVE_ALG1	algorithm 1. (default)
CUSOLVER_TRIANGULAR_SOLVE_ALG2	algorithm 2. Domino-based scheme.
CUSOLVER_TRIANGULAR_SOLVE_ALG3	algorithm 3. Domino-based scheme.

### 2.2.3.7. cusolverRfUnitDiagonal\_t

The `cusolverRfUnitDiagonal_t` is an enum that indicates whether and where the unit diagonal is stored in the input/output triangular factors

in the `cusolverRfSetupDevice()`, `cusolverRfSetupHost()` and `cusolverRfExtractSplitFactorsHost()` routines.

Value	Meaning
<code>CUSOLVER_UNIT_DIAGONAL_STORED_L</code>	unit diagonal is stored in lower triangular factor. (default)
<code>CUSOLVER_UNIT_DIAGONAL_STORED_U</code>	unit diagonal is stored in upper triangular factor.
<code>CUSOLVER_UNIT_DIAGONAL_ASSUMED_L</code>	unit diagonal is assumed in lower triangular factor.
<code>CUSOLVER_UNIT_DIAGONAL_ASSUMED_U</code>	unit diagonal is assumed in upper triangular factor.

### 2.2.3.8. `cusolverStatus_t`

The `cusolverStatus_t` is an enum that indicates success or failure of the `cuSolverRF` library call. It is returned by all the `cuSolver` library routines, and it uses the same enumerated values as the sparse and dense Lapack routines.

## 2.3. `cuSolver` Formats Reference

### 2.3.1. Index Base Format

The CSR or CSC format requires either zero-based or one-based index for a sparse matrix **A**. The GLU library supports only zero-based indexing. Otherwise, both one-based and zero-based indexing are supported in `cuSolver`.

### 2.3.2. Vector (Dense) Format

The vectors are assumed to be stored linearly in memory. For example, the vector

$$x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

is represented as

$$(x_1 \ x_2 \ \dots \ x_n)$$

### 2.3.3. Matrix (Dense) Format

The dense matrices are assumed to be stored in column-major order in memory. The sub-matrix can be accessed using the leading dimension of the original matrix. For example, the **m\*n** (sub-)matrix

$$\begin{pmatrix} a_{1,1} & \dots & a_{1,n} \\ a_{2,1} & \dots & a_{2,n} \\ \vdots & & \\ a_{m,1} & \dots & a_{m,n} \end{pmatrix}$$

is represented as

$$\begin{pmatrix} a_{1,1} & \dots & a_{1,n} \\ a_{2,1} & \dots & a_{2,n} \\ \vdots & \ddots & \vdots \\ a_{m,1} & \dots & a_{m,n} \\ \vdots & \ddots & \vdots \\ a_{lda,1} & \dots & a_{lda,n} \end{pmatrix}$$

with its elements arranged linearly in memory as

$$(a_{1,1} \ a_{2,1} \ \dots \ a_{m,1} \ \dots \ a_{lda,1} \ \dots \ a_{1,n} \ a_{2,n} \ \dots \ a_{m,n} \ \dots \ a_{lda,n})$$

where  $lda \geq m$  is the leading dimension of  $\mathbf{A}$ .

### 2.3.4. Matrix (CSR) Format

In CSR format the matrix is represented by the following parameters

parameter	type	size	Meaning
<code>n</code>	(int)		the number of rows (and columns) in the matrix.
<code>nnz</code>	(int)		the number of non-zero elements in the matrix.
<code>csrRowPtr</code>	(int *)	<code>n+1</code>	the array of offsets corresponding to the start of each row in the arrays <code>csrColInd</code> and <code>csrVal</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix.
<code>csrColInd</code>	(int *)	<code>nnz</code>	the array of column indices corresponding to the non-zero elements in the matrix. <b>It is assumed that this array is sorted by row and by column within each row.</b>
<code>csrVal</code>	(S D C Z) *	<code>nnz</code>	the array of values corresponding to the non-zero elements in the matrix. <b>It is assumed that this array is sorted by row and by column within each row.</b>

Note that in our CSR format sparse matrices are assumed to be stored in row-major order, in other words, the index arrays are first sorted by row indices and then within each row by column indices. Also it is assumed that each pair of row and column indices appears only once.

For example, the **4x4** matrix

$$A = \begin{pmatrix} 1.0 & 3.0 & 0.0 & 0.0 \\ 0.0 & 4.0 & 6.0 & 0.0 \\ 2.0 & 5.0 & 7.0 & 8.0 \\ 0.0 & 0.0 & 0.0 & 9.0 \end{pmatrix}$$

is represented as

$$\text{csrRowPtr} = (0 \ 2 \ 4 \ 8 \ 9)$$

```
csrColInd = (0 1 1 2 0 1 2 3 3)
```

```
csrVal = (1.0 3.0 4.0 6.0 2.0 5.0 7.0 8.0 9.0)
```

## 2.3.5. Matrix (CSC) Format

In CSC format the matrix is represented by the following parameters

parameter	type	size	Meaning
<code>n</code>	<code>(int)</code>		the number of rows (and columns) in the matrix.
<code>nnz</code>	<code>(int)</code>		the number of non-zero elements in the matrix.
<code>cscColPtr</code>	<code>(int *)</code>	<code>n+1</code>	the array of offsets corresponding to the start of each column in the arrays <code>cscRowInd</code> and <code>cscVal</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix.
<code>cscRowInd</code>	<code>(int *)</code>	<code>nnz</code>	the array of row indices corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by column and by row within each column.
<code>cscVal</code>	<code>(S D C Z) *</code>	<code>nnz</code>	the array of values corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by column and by row within each column.

Note that in our CSC format sparse matrices are assumed to be stored in column-major order, in other words, the index arrays are first sorted by column indices and then within each column by row indices. Also it is assumed that each pair of row and column indices appears only once.

For example, the **4x4** matrix

$$A = \begin{pmatrix} 1.0 & 3.0 & 0.0 & 0.0 \\ 0.0 & 4.0 & 6.0 & 0.0 \\ 2.0 & 5.0 & 7.0 & 8.0 \\ 0.0 & 0.0 & 0.0 & 9.0 \end{pmatrix}$$

is represented as

```
cscColPtr = (0 2 5 7 9)
```

```
cscRowInd = (0 2 0 1 2 1 2 2 3)
```

```
cscVal = (1.0 2.0 3.0 4.0 5.0 6.0 7.0 8.0 9.0)
```

## 2.4. cuSolverDN: dense LAPACK Function Reference

This chapter describes the API of cuSolverDN, which provides a subset of dense LAPACK functions.

### 2.4.1. cuSolverDN Helper Function Reference

The cuSolverDN helper functions are described in this section.

#### 2.4.1.1. cusolverDnCreate()

```
cusolverStatus_t
cusolverDnCreate(cusolverDnHandle_t *handle);
```

This function initializes the cuSolverDN library and creates a handle on the cuSolverDN context. It must be called before any other cuSolverDN API function is invoked. It allocates hardware resources necessary for accessing the GPU.

parameter	Memory	In/out	Meaning
handle	host	output	the pointer to the handle to the cuSolverDN context.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the initialization succeeded.
CUSOLVER_STATUS_NOT_INITIALIZED	the CUDA Runtime initialization failed.
CUSOLVER_STATUS_ALLOC_FAILED	the resources could not be allocated.
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.

#### 2.4.1.2. cusolverDnDestroy()

```
cusolverStatus_t
cusolverDnDestroy(cusolverDnHandle_t handle);
```

This function releases CPU-side resources used by the cuSolverDN library.

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the shutdown succeeded.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

### 2.4.1.3. cusolverDnSetStream()

```
cusolverStatus_t
cusolverDnSetStream(cusolverDnHandle_t handle, cudaStream_t streamId)
```

This function sets the stream to be used by the cuSolverDN library to execute its routines.

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
streamId	host	input	the stream to be used by the library.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the stream was set successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

### 2.4.1.4. cusolverDnGetStream()

```
cusolverStatus_t
cusolverDnGetStream(cusolverDnHandle_t handle, cudaStream_t *streamId)
```

This function sets the stream to be used by the cuSolverDN library to execute its routines.

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
streamId	host	output	the stream to be used by the library.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the stream was set successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

### 2.4.1.5. cusolverDnCreateSyevejInfo()

```
cusolverStatus_t
cusolverDnCreateSyevejInfo(
    syevjInfo_t *info);
```

This function creates and initializes the structure of **syevj**, **syevjBatched** and **sygvj** to default values.

parameter	Memory	In/out	Meaning
info	host	output	the pointer to the structure of syevj.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the structure was initialized successfully.
-------------------------	---------------------------------------------

CUSOLVER_STATUS_ALLOC_FAILED	the resources could not be allocated.
------------------------------	---------------------------------------

### 2.4.1.6. cusolverDnDestroySyevjInfo()

```
cusolverStatus_t
cusolverDnDestroySyevjInfo(
    syevjInfo_t info);
```

This function destroys and releases any memory required by the structure.

parameter	Memory	In/out	Meaning
info	host	input	the structure of syevj.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the resources are released successfully.
-------------------------	------------------------------------------

### 2.4.1.7. cusolverDnXsyevjSetTolerance()

```
cusolverStatus_t
cusolverDnXsyevjSetTolerance(
    syevjInfo_t info,
    double tolerance)
```

This function configures tolerance of **syevj**.

parameter	Memory	In/out	Meaning
info	host	in/out	the pointer to the structure of syevj.
tolerance	host	input	accuracy of numerical eigenvalues.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
-------------------------	---------------------------------------

### 2.4.1.8. cusolverDnXsyevjSetMaxSweeps()

```
cusolverStatus_t
cusolverDnXsyevjSetMaxSweeps(
    syevjInfo_t info,
    int max_sweeps)
```

This function configures maximum number of sweeps in **syevj**. The default value is 100.

parameter	Memory	In/out	Meaning
info	host	in/out	the pointer to the structure of syevj.
max_sweeps	host	input	maximum number of sweeps.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
-------------------------	---------------------------------------

### 2.4.1.9. cusolverDnXsyevjSetSortEig()

```
cusolverStatus_t
cusolverDnXsyevjSetSortEig(
    syevjInfo_t info,
    int sort_eig)
```

if **sort\_eig** is zero, the eigenvalues are not sorted. This function only works for **syevjBatched**. **syevj** and **sygvj** always sort eigenvalues in ascending order. By default, eigenvalues are always sorted in ascending order.

parameter	Memory	In/out	Meaning
<b>info</b>	host	in/out	the pointer to the structure of syevj.
<b>sort_eig</b>	host	input	if <b>sort_eig</b> is zero, the eigenvalues are not sorted.

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
--------------------------------	---------------------------------------

### 2.4.1.10. cusolverDnXsyevjGetResidual()

```
cusolverStatus_t
cusolverDnXsyevjGetResidual(
    cusolverDnHandle_t handle,
    syevjInfo_t info,
    double *residual)
```

This function reports residual of **syevj** or **sygvj**. It does not support **syevjBatched**. If the user calls this function after **syevjBatched**, the error **CUSOLVER\_STATUS\_NOT\_SUPPORTED** is returned.

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverDN library context.
<b>info</b>	host	input	the pointer to the structure of syevj.
<b>residual</b>	host	output	residual of syevj.

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_SUPPORTED</b>	does not support batched version

### 2.4.1.11. cusolverDnXsyevjGetSweeps()

```
cusolverStatus_t
cusolverDnXsyevjGetSweeps(
    cusolverDnHandle_t handle,
    syevjInfo_t info,
    int *executed_sweeps)
```

This function reports number of executed sweeps of **syevj** or **sygvj**. It does not support **syevjBatched**. If the user calls this function after **syevjBatched**, the error **CUSOLVER\_STATUS\_NOT\_SUPPORTED** is returned.

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverDN library context.
<b>info</b>	host	input	the pointer to the structure of <b>syevj</b> .
<b>executed_sweeps</b>	host	output	number of executed sweeps.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_SUPPORTED</b>	does not support batched version

## 2.4.1.12. cusolverDnCreateGesvdjInfo()

```
cusolverStatus_t
cusolverDnCreateGesvdjInfo(
    gesvdjInfo_t *info);
```

This function creates and initializes the structure of **gesvdj** and **gesvdjBatched** to default values.

parameter	Memory	In/out	Meaning
<b>info</b>	host	output	the pointer to the structure of <b>gesvdj</b> .

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the structure was initialized successfully.
<b>CUSOLVER_STATUS_ALLOC_FAILED</b>	the resources could not be allocated.

## 2.4.1.13. cusolverDnDestroyGesvdjInfo()

```
cusolverStatus_t
cusolverDnDestroyGesvdjInfo(
    gesvdjInfo_t info);
```

This function destroys and releases any memory required by the structure.

parameter	Memory	In/out	Meaning
<b>info</b>	host	input	the structure of <b>gesvdj</b> .

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the resources are released successfully.
--------------------------------	------------------------------------------

### 2.4.1.14. cusolverDnXgesvdjSetTolerance()

```
cusolverStatus_t
cusolverDnXgesvdjSetTolerance(
    gesvdjInfo_t info,
    double tolerance)
```

This function configures tolerance of **gesvdj**.

parameter	Memory	In/out	Meaning
<b>info</b>	host	in/out	the pointer to the structure of <b>gesvdj</b> .
<b>tolerance</b>	host	input	accuracy of numerical singular values.

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
--------------------------------	---------------------------------------

### 2.4.1.15. cusolverDnXgesvdjSetMaxSweeps()

```
cusolverStatus_t
cusolverDnXgesvdjSetMaxSweeps(
    gesvdjInfo_t info,
    int max_sweeps)
```

This function configures maximum number of sweeps in **gesvdj**. The default value is 100.

parameter	Memory	In/out	Meaning
<b>info</b>	host	in/out	the pointer to the structure of <b>gesvdj</b> .
<b>max_sweeps</b>	host	input	maximum number of sweeps.

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
--------------------------------	---------------------------------------

### 2.4.1.16. cusolverDnXgesvdjSetSortEig()

```
cusolverStatus_t
cusolverDnXgesvdjSetSortEig(
    gesvdjInfo_t info,
    int sort_svd)
```

if **sort\_svd** is zero, the singular values are not sorted. This function only works for **gesvdjBatched**. **gesvdj** always sorts singular values in descending order. By default, singular values are always sorted in descending order.

parameter	Memory	In/out	Meaning
<b>info</b>	host	in/out	the pointer to the structure of <b>gesvdj</b> .
<b>sort_svd</b>	host	input	if <b>sort_svd</b> is zero, the singular values are not sorted.

**Status Returned**

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
-------------------------	---------------------------------------

**2.4.1.17. cusolverDnXgesvdjGetResidual()**

```
cusolverStatus_t
cusolverDnXgesvdjGetResidual(
    cusolverDnHandle_t handle,
    gesvdjInfo_t info,
    double *residual)
```

This function reports residual of **gesvdj**. It does not support **gesvdjBatched**. If the user calls this function after **gesvdjBatched**, the error **CUSOLVER\_STATUS\_NOT\_SUPPORTED** is returned.

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
info	host	input	the pointer to the structure of gesvdj.
residual	host	output	residual of gesvdj.

**Status Returned**

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_SUPPORTED	does not support batched version

**2.4.1.18. cusolverDnXgesvdjGetSweeps()**

```
cusolverStatus_t
cusolverDnXgesvdjGetSweeps(
    cusolverDnHandle_t handle,
    gesvdjInfo_t info,
    int *executed_sweeps)
```

This function reports number of executed sweeps of **gesvdj**. It does not support **gesvdjBatched**. If the user calls this function after **gesvdjBatched**, the error **CUSOLVER\_STATUS\_NOT\_SUPPORTED** is returned.

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
info	host	input	the pointer to the structure of gesvdj.
executed_sweeps	host	output	number of executed sweeps.

**Status Returned**

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_SUPPORTED	does not support batched version

### 2.4.1.19. cusolverDnIRSPParamsCreate()

```
cusolverStatus_t
cusolverDnIRSPParamsCreate(cusolverDnIRSPParams_t *params);
```

This function creates and initializes the structure of parameters for an IRS solver such as the `cusolverDnIRSXgesv()` or the `cusolverDnIRSXgels()` functions to default values. The params structure created by this function can be used by one or more call to the same or to a different IRS solver. Note that, in CUDA-10.2, the behavior was different and a new params structure was needed to be created per each call to an IRS solver. Also note that, the user can also change configurations of the params then call a new IRS instance, but be careful that the previous call was done because any change to the configuration before that the previous call was done could affect it.

parameter	Memory	In/out	Meaning
params	host	output	Pointer to the cusolverDnIRSPParams_t Params structure

#### Status Returned

CUSOLVER_STATUS_SUCCESS	The structure was created and initialized successfully.
CUSOLVER_STATUS_ALLOC_FAILED	The resources could not be allocated.

### 2.4.1.20. cusolverDnIRSPParamsDestroy()

```
cusolverStatus_t
cusolverDnIRSPParamsDestroy(cusolverDnIRSPParams_t params);
```

This function destroys and releases any memory required by the Params structure.

parameter	Memory	In/out	Meaning
params	host	input	The cusolverDnIRSPParams_t Params structure

#### Status Returned

CUSOLVER_STATUS_SUCCESS	The resources are released successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.
CUSOLVER_STATUS_IRS_INFOS_NOT_DESTROYED	Not all the Infos structure associated with this Params structure have been destroyed yet.

### 2.4.1.21. cusolverDnIRSPParamsSetSolverPrecisions()

```
cusolverStatus_t
cusolverDnIRSPParamsSetSolverPrecisions(
    cusolverDnIRSPParams_t params,
    cusolverPrecType_t solver_main_precision,
    cusolverPrecType_t solver_lowest_precision );
```

This function set both, the main and the lowest precision for the Iterative Refinement Solver (IRS). By main precision, we mean the precision of the Input and Output datatype. By lowest precision, we mean the solver is allowed to use as lowest computational precision during the LU factorization process. Note that, the user has to set both the main and lowest precision before a first call to the IRS solver because they are NOT set by default with the params structure creation, as it depends on the Input Output data type and user request. It is a wrapper to both `cusolverDnIRSParamsSetSolverMainPrecision()` and `cusolverDnIRSParamsSetSolverLowestPrecision()`. All possible combinations of main/lowest precision are described in the table below. Usually the lowest precision defines the speedup that can be achieved. The ratio of the performance of the lowest precision over the main precision (e.g., Inputs/Outputs datatype) define somehow the upper bound of the speedup that could be obtained. More precisely, it depends on many factors, but for large matrices sizes, it is the ratio of the matrix-matrix rank-k product (e.g., GEMM where K is 256 and M=N=size of the matrix) that define the possible speedup. For instance, if the inout precision is real double precision CUSOLVER\_R\_64F and the lowest precision is CUSOLVER\_R\_32F, then we can expect a speedup of at most 2X for large problem sizes. If the lowest precision was CUSOLVER\_R\_16F, then we can expect 3X-4X. A reasonable strategy should take the number of right-hand sides, the size of the matrix as well as the convergence rate into account.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSParams_t Params structure
solver_main_precision	host	input	Allowed Inputs/Outputs datatype (for example CUSOLVER_R_FP64 for a real double precision data). See the table below for the supported precisions.
solver_lowest_precision	host	input	Allowed lowest compute type (for example CUSOLVER_R_16F for half precision computation). See the table below for the supported precisions.

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

### Supported Inputs/Outputs data type and lower precision for the IRS solver

Inputs/Outputs Data Type (e.g., main precision)	Supported values for the lowest precision
CUSOLVER_C_64F	CUSOLVER_C_64F, CUSOLVER_C_32F, CUSOLVER_C_16F, CUSOLVER_C_16BF, CUSOLVER_C_TF32
CUSOLVER_C_32F	CUSOLVER_C_32F, CUSOLVER_C_16F, CUSOLVER_C_16BF, CUSOLVER_C_TF32
CUSOLVER_R_64F	CUSOLVER_R_64F, CUSOLVER_R_32F, CUSOLVER_R_16F, CUSOLVER_R_16BF, CUSOLVER_R_TF32

CUSOLVER_R_32F	CUSOLVER_R_32F, CUSOLVER_R_16F, CUSOLVER_R_16BF, CUSOLVER_R_TF32
----------------	---------------------------------------------------------------------

### 2.4.1.22. cusolverDnIRSParamsSetSolverMainPrecision()

```
cusolverStatus_t
cusolverDnIRSParamsSetSolverMainPrecision(
    cusolverDnIRSParams_t params,
    cusolverPrecType_t solver_main_precision);
```

This function sets the main precision for the Iterative Refinement Solver (IRS). By main precision, we mean, the type of the Input and Output data. Note that, the user has to set both the main and lowest precision before a first call to the IRS solver because they are NOT set by default with the params structure creation, as it depends on the Input Output data type and user request. user can set it by either calling this function or by calling **cusolverDnIRSParamsSetSolverPrecisions()** which set both the main and the lowest precision together. All possible combinations of main/lowest precision are described in the table in the **cusolverDnIRSParamsSetSolverPrecisions()** section above.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSParams_t Params structure
solver_main_precision	host	input	Allowed Inputs/Outputs datatype (for example CUSOLVER_R_FP64 for a real double precision data). See the table in the <b>cusolverDnIRSParamsSetSolverPrecisions()</b> section above for the supported precisions.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

### 2.4.1.23. cusolverDnIRSParamsSetSolverLowestPrecision()

```
cusolverStatus_t
cusolverDnIRSParamsSetSolverLowestPrecision(
    cusolverDnIRSParams_t params,
    cusolverPrecType_t lowest_precision_type);
```

This function sets the lowest precision that will be used by Iterative Refinement Solver. By lowest precision, we mean the solver is allowed to use as lowest computational precision during the LU factorization process. Note that, the user has to set both the main and lowest precision before a first call to the IRS solver because they are NOT set by default with the params structure creation, as it depends on the Input Output data type and user request. Usually the lowest precision defines the speedup that can be achieved. The ratio of the performance of the lowest precision over the main precision (e.g., Inputs/Outputs datatype) define somehow the upper bound of the speedup that

could be obtained. More precisely, it depends on many factors, but for large matrices sizes, it is the ratio of the matrix-matrix rank-k product (e.g., GEMM where K is 256 and  $M=N$ =size of the matrix) that define the possible speedup. For instance, if the inout precision is real double precision CUSOLVER\_R\_64F and the lowest precision is CUSOLVER\_R\_32F, then we can expect a speedup of at most 2X for large problem sizes. If the lowest precision was CUSOLVER\_R\_16F, then we can expect 3X-4X. A reasonable strategy should take the number of right-hand sides, the size of the matrix as well as the convergence rate into account.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure
lowest_precision_typ	host	input	Allowed lowest compute type (for example CUSOLVER_R_16F for half precision computation). See the table in the <code>cusolverDnIRSPParamsSetSolverPrecisions()</code> section above for the supported precisions.

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

## 2.4.1.24. cusolverDnIRSPParamsSetRefinementSolver()

```
cusolverStatus_t
cusolverDnIRSPParamsSetRefinementSolver(
    cusolverDnIRSPParams_t params,
    cusolverIRSRefinement_t solver);
```

This function sets the refinement solver to be used in the Iterative Refinement Solver functions such as the `cusolverDnIRSXgesv()` or the `cusolverDnIRSXgels()` functions. Note that, the user has to set the refinement algorithm before a first call to the IRS solver because it is NOT set by default with the creating of params. Details about values that can be set to and theirs meaning are described in the table below.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure
solver	host	input	Type of the refinement solver to be used by the IRS solver such as <code>cusolverDnIRSXgesv()</code> or <code>cusolverDnIRSXgels</code>

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

Value	Meaning
CUSOLVER_IRS_REFINE_NOT_SET	Solver is not set, this value is what is set when creating the params structure. IRS solver will return error.
CUSOLVER_IRS_REFINE_NONE	No refinement solver, the IRS solver performs a factorization followed by a solve without any refinement. For example, if the IRS solver was <code>cusolverDnIRSXgesv()</code> , this is equivalent to a <code>Xgesv</code> routine without refinement and where the factorization is carried out in the lowest precision. If for example the main precision was CUSOLVER_R_64F and the lowest was CUSOLVER_R_32F as well, then this is equivalent to a call to <code>cusolverDnDgesv()</code> .
CUSOLVER_IRS_REFINE_CLASSICAL	Classical iterative refinement solver. Similar to the one used in LAPACK routines.
CUSOLVER_IRS_REFINE_GMRES	GMRES (Generalized Minimal Residual) based iterative refinement solver. In recent study, the GMRES method has drawn the scientific community attention for its ability to be used as refinement solver that outperforms the classical iterative refinement method. Based on our experimentation, we recommend this setting.
CUSOLVER_IRS_REFINE_CLASSICAL_GMRES	Classical iterative refinement solver that uses the GMRES (Generalized Minimal Residual) internally to solve the correction equation at each iteration. We call the <i>classical refinement iteration</i> the outer iteration while the <i>GMRES</i> is called inner iteration. Note that if the tolerance of the inner GMRES is set very low, let say to machine precision, then the outer <i>classical refinement iteration</i> will performs only one iteration and thus this option will behaves like CUSOLVER_IRS_REFINE_GMRES.
CUSOLVER_IRS_REFINE_GMRES_GMRES	Similar to CUSOLVER_IRS_REFINE_CLASSICAL_GMRES which consists of classical refinement process that uses GMRES to solve the inner correction system, here it is a GMRES (Generalized Minimal Residual) based iterative refinement solver that uses another GMRES internally to solve the preconditioned system.

### 2.4.1.25. cusolverDnIRSParamsSetTol()

```
cusolverStatus_t
cusolverDnIRSParamsSetTol(
    cusolverDnIRSParams_t params,
    double val );
```

This function sets the tolerance for the refinement solver. By default it is such that, all the RHS satisfy:

$$RNRM < \sqrt{N} * XNRM * ANRM * EPS * BWDMAX \quad \text{where}$$

- ▶ RNRM is the infinity-norm of the residual
- ▶ XNRM is the infinity-norm of the solution
- ▶ ANRM is the infinity-operator-norm of the matrix A
- ▶ EPS is the machine epsilon for the Inputs/Outputs datatype that matches LAPACK <X>LAMCH('Epsilon')
- ▶ BWDMAX, the value BWDMAX is fixed to 1.0

The user can use this function to change the tolerance to a lower or higher value. Our goal is to give the user more control such a way he can investigate and control every detail of the IRS solver. Note the, the tolerance value is always in *real double precision* whatever the Inputs/Outputs datatype is.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure
val	host	input	double precision real value to which the refinement tolerance will be set.

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

### 2.4.1.26. cusolverDnIRSPParamsSetTolInner()

```
cusolverStatus_t
cusolverDnIRSPParamsSetTolInner(
    cusolverDnIRSPParams_t params,
    double val );
```

This function sets the tolerance for the inner refinement solver when the refinement solver consists of two-levels solver (e.g., CUSOLVER\_IRS\_REFINE\_CLASSICAL\_GMRES or CUSOLVER\_IRS\_REFINE\_GMRES\_GMRES cases). It is not referenced in case of one level refinement solver such as CUSOLVER\_IRS\_REFINE\_CLASSICAL or CUSOLVER\_IRS\_REFINE\_GMRES. It is set to 1e-4 by default. This function set the tolerance for the inner solver (e.g. the inner GMRES). For example, if the Refinement Solver was set to CUSOLVER\_IRS\_REFINE\_CLASSICAL\_GMRES, setting this tolerance mean that the inner GMRES solver will converge to that tolerance at each outer iteration of the classical refinement solver. Our goal is to give the user more control such a way he can investigate and control every detail of the IRS solver. Note the, the tolerance value is always in *real double precision* whatever the Inputs/Outputs datatype is.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure
val	host	input	Double precision real value to which the tolerance of the inner refinement solver will be set.

**Status Returned**

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

**2.4.1.27. cusolverDnIRSPParamsSetMaxIters()**

```
cusolverStatus_t
cusolverDnIRSPParamsSetMaxIters(
    cusolverDnIRSPParams_t params,
    int max_iters);
```

This function sets the total number of allowed refinement iterations after which the solver will stop. Total means any iteration which means the sum of the outer and the inner iterations (inner is meaningful when two-levels refinement solver is set). Default value is set to 50. Our goal is to give the user more control such a way he can investigate and control every detail of the IRS solver.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure
max_iters	host	input	Maximum total number of iterations allowed for the refinement solver

**Status Returned**

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

**2.4.1.28. cusolverDnIRSPParamsSetMaxItersInner()**

```
cusolverStatus_t
cusolverDnIRSPParamsSetMaxItersInner(
    cusolverDnIRSPParams_t params,
    cusolver_int_t maxiters_inner );
```

This function sets the maximal number of iterations allowed for the inner refinement solver. It is not referenced in case of one level refinement solver such as CUSOLVER\_IRS\_REFINE\_CLASSICAL or CUSOLVER\_IRS\_REFINE\_GMRES. The inner refinement solver will stop after reaching either the inner tolerance or the MaxItersInner value. By default, it is set to 50. Note that this value could not be larger than the MaxIters since MaxIters is the total number of allowed iterations. Note that, if the user call cusolverDnIRSPParamsSetMaxIters after calling this function, the SetMaxIters has priority and will overwrite MaxItersInner to the minimum value of (MaxIters, MaxItersInner).

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure
maxiters_inner	host	input	Maximum number of allowed inner iterations for the inner refinement

			solver. Meaningful when the refinement solver is a two-levels solver such as CUSOLVER_IRS_REFINE_CLASSICAL_GMRES or CUSOLVER_IRS_REFINE_GMRES_GMRES. Value should be less or equal to MaxIters.
--	--	--	-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.
CUSOLVER_STATUS_IRS_PARAMS_INVALID	if the value was larger than MaxIters.

### 2.4.1.29. cusolverDnIRSPParamsEnableFallback()

```
cusolverStatus_t
cusolverDnIRSPParamsEnableFallback(
    cusolverDnIRSPParams_t params );
```

This function enable the fallback to the main precision in case the Iterative Refinement Solver (IRS) failed to converge. In other term, if the IRS solver failed to converge, the solver will return a no convergence code (e.g., niter < 0), but can either return the non-convergent solution as it is (e.g., disable fallback) or can fallback (e.g., enable fallback) to the main precision (which is the precision of the Inputs/Outputs data) and solve the problem from scratch returning the good solution. This is the behavior by default, and it will guarantee that the IRS solver always provide the good solution. this function is provided because we provided **cusolverDnIRSPParamsDisableFallback** which allow the user to disable the fallback and thus this function allow him to re-enable it.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

### 2.4.1.30. cusolverDnIRSPParamsDisableFallback()

```
cusolverStatus_t
cusolverDnIRSPParamsDisableFallback(
    cusolverDnIRSPParams_t params );
```

This function disables the fallback to the main precision in case the Iterative Refinement Solver (IRS) failed to converge. In other term, if the IRS solver failed to converge, the solver will return a no convergence code (e.g., niter < 0), but can either return the non-convergent solution as it is (e.g., disable fallback) or can fallback (e.g., enable fallback) to the main precision (which is the precision of the Inputs/Outputs data) and solve the problem from scratch returning the good solution. This function

disables the fallback and the returned solution is whatever the refinement solver was able to reach before it returns. Disabling fallback does not guarantee that the solution is the good one. However, for some users, who want to keep getting the solution of the lower precision, in case the IRS did not converge after certain number of iterations, they need to disable the fallback. The user can re-enable it by calling **cusolverDnIRSPParamsEnableFallback**.

parameter	Memory	In/out	Meaning
params	host	in/out	The cusolverDnIRSPParams_t Params structure

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

### 2.4.1.31. cusolverDnIRSPParamsGetMaxIters()

```
cusolverStatus_t
cusolverDnIRSPParamsGetMaxIters (
    cusolverDnIRSPParams_t params,
    cusolver_int_t *maxiters );
```

This function returns the current setting in the params structure for the maximal allowed number of iterations (e.g., either the default MaxIters, or the one set by the user in case he set it using **cusolverDnIRSPParamsSetMaxIters**). Note that, this function returns the current setting in the params configuration and not to be confused with the **cusolverDnIRSInfosGetMaxIters** which return the maximal allowed number of iterations for a particular call to an IRS solver. To be clearer, the params structure can be used for many calls to an IRS solver. A user can change the allowed MaxIters between calls while the Infos structure in **cusolverDnIRSInfosGetMaxIters** contains information about a particular call and cannot be reused for different call, and thus, **cusolverDnIRSInfosGetMaxIters** return the allowed MaxIters for that call.

parameter	Memory	In/out	Meaning
params	host	in	The cusolverDnIRSPParams_t Params structure
maxiters	host	output	The maximal number of iterations that is currently set

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE	The Params structure was not created.

### 2.4.1.32. cusolverDnIRSInfosCreate()

```
cusolverStatus_t
cusolverDnIRSInfosCreate(
    cusolverDnIRSInfos_t* infos )
```

This function creates and initializes the Infos structure that will hold the refinement information of an Iterative Refinement Solver (IRS) call. Such information includes the total number of iterations that was needed to converge (Niters), the outer number of iterations (meaningful when two-levels preconditioner such as CUSOLVER\_IRS\_REFINE\_CLASSICAL\_GMRES is used), the maximal number of iterations that was allowed for that call, and a pointer to the matrix of the convergence history residual norms. The Infos structure need to be created before a call to an IRS solver. The Infos structure is valid for only one call to an IRS solver, since it holds info about that solve and thus each solve will requires its own Infos structure.

parameter	Memory	In/out	Meaning
infos	host	output	Pointer to the cusolverDnIRSInfos_t Infos structure

#### Status Returned

CUSOLVER_STATUS_SUCCESS	The structure was initialized successfully.
CUSOLVER_STATUS_ALLOC_FAILED	The resources could not be allocated.

### 2.4.1.33. cusolverDnIRSInfosDestroy()

```
cusolverStatus_t
cusolverDnIRSInfosDestroy(
    cusolverDnIRSInfos_t infos );
```

This function destroys and releases any memory required by the Infos structure. This function destroys all the informations (e.g., Niters performed, OuterNiters performed, residual history etc) about a solver call, thus, a user is supposed to call it once he is done from the informations he need.

parameter	Memory	In/out	Meaning
infos	host	in/out	The cusolverDnIRSInfos_t Infos structure

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the resources are released successfully.
CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED	The Infos structure was not created.

### 2.4.1.34. cusolverDnIRSInfosGetMaxIters()

```
cusolverStatus_t
cusolverDnIRSInfosGetMaxIters(
    cusolverDnIRSInfos_t infos,
    cusolver_int_t *maxiters );
```

This function returns the maximal allowed number of iterations that was set for the corresponding call to the IRS solver. Note that, this function returns the setting that was set when that call happened and not to be confused with the `cusolverDnIRSParamsGetMaxIters` which return the current setting in the params configuration structure. To be clearer, the params structure can be used for many calls to an IRS solver. A user can change the allowed MaxIters between calls while the Infos structure in `cusolverDnIRSInfosGetMaxIters` contains information about a particular call and cannot be reused for different call, and thus `cusolverDnIRSInfosGetMaxIters` return the allowed MaxIters for that call.

parameter	Memory	In/out	Meaning
infos	host	in	The cusolverDnIRSInfos_t Infos structure
maxiters	host	output	The maximal number of iterations that is currently set

#### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED	The Infos structure was not created.

### 2.4.1.35. cusolverDnIRSInfosGetNiters()

```
cusolverStatus_t cusolverDnIRSInfosGetNiters (
    cusolverDnIRSInfos_t infos,
    cusolver_int_t *niters );
```

This function returns the total number of iterations performed by the IRS solver. If it was negative it means that the IRS solver did not converge and if the user did not disable the fallback to full precision, then the fallback to a full precision solution happened and solution is good. Please refer to the description of negative niters values in the corresponding IRS linear solver functions such as `cusolverDnXgesv()` or `cusolverDnXgels()`

parameter	Memory	In/out	Meaning
infos	host	in	The cusolverDnIRSInfos_t Infos structure
niters	host	output	The total number of iterations performed by the IRS solver

#### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED	The Infos structure was not created.

### 2.4.1.36. cusolverDnIRSInfosGetOuterNiters()

```
cusolverStatus_t
cusolverDnIRSInfosGetOuterNiters (
    cusolverDnIRSInfos_t infos,
    cusolver_int_t *outer_niters );
```

This function returns the number of iterations performed by the outer refinement loop of the IRS solver. When the refinement solver consists of a one level solver such as CUSOLVER\_IRS\_REFINE\_CLASSICAL or CUSOLVER\_IRS\_REFINE\_GMRES, it is the same as Niters. When the refinement solver consists of a two-levels solver such as CUSOLVER\_IRS\_REFINE\_CLASSICAL\_GMRES or CUSOLVER\_IRS\_REFINE\_GMRES\_GMRES, it is the number of iterations of the outer loop. See description of `cusolverIRSRefinementSolver_t` section for more details.

parameter	Memory	In/out	Meaning
<code>infos</code>	host	in	The <code>cusolverDnIRSInfos_t</code> Infos structure
<code>outer_niters</code>	host	output	The number of iterations of the outer refinement loop of the IRS solver

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	The operation completed successfully.
<code>CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED</code>	The Infos structure was not created.

## 2.4.1.37. `cusolverDnIRSInfosRequestResidual()`

```
cusolverStatus_t cusolverDnIRSInfosRequestResidual(
    cusolverDnIRSInfos_t infos );
```

This function, once called, tell the IRS solver to store the convergence history (residual norms) of the refinement phase in a matrix, that could be accessed via a pointer returned by the `cusolverDnIRSInfosGetResidualHistory()` function.

parameter	Memory	In/out	Meaning
<code>infos</code>	host	in	The <code>cusolverDnIRSInfos_t</code> Infos structure

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	The operation completed successfully.
<code>CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED</code>	The Infos structure was not created.

## 2.4.1.38. `cusolverDnIRSInfosGetResidualHistory()`

```
cusolverStatus_t
cusolverDnIRSInfosGetResidualHistory(
    cusolverDnIRSInfos_t infos,
    void **residual_history );
```

If the user called `cusolverDnIRSInfosRequestResidual()` before the call to the IRS function, then the IRS solver will store the convergence history (residual norms) of the refinement phase in a matrix, that could be accessed via a pointer returned by this function. The datatype of the residual norms depends on the input and output data type. If the Inputs/Outputs datatype is double precision real or complex (CUSOLVER\_R\_FP64 or CUSOLVER\_C\_FP64), this residual will be of type real double precision (FP64) *double*, otherwise if the Inputs/Outputs datatype is single precision real or complex

(CUSOLVER\_R\_FP32 or CUSOLVER\_C\_FP32), this residual will be real single precision FP32 *float*.

The residual history matrix consists of two columns (even for the multiple right-hand side case NRHS) of MaxIters+1 row, thus a matrix of size (MaxIters+1,2). Only the first OuterNeters+1 rows contains the residual norms the other (e.g., OuterNeters+2:Maxiters+1) are garbage. On the first column, each row "i" specify the total number of iterations happened till this outer iteration "i" and on the second columns the residual norm corresponding to this outer iteration "i". Thus, the first row (e.g., outer iteration "0") consists of the initial residual (e.g., the residual before the refinement loop start) then the consecutive rows are the residual obtained at each outer iteration of the refinement loop. Note, it only consists of the history of the outer loop.

If the refinement solver was CUSOLVER\_IRS\_REFINE\_CLASSICAL or CUSOLVER\_IRS\_REFINE\_GMRES, then OuterNeters=Neters (Neters is the total number of iterations performed) and there is Neters+1 rows of norms that correspond to the Neters outer iterations.

If the refinement solver was CUSOLVER\_IRS\_REFINE\_CLASSICAL\_GMRES or CUSOLVER\_IRS\_REFINE\_GMRES\_GMRES, then OuterNeters <= Neters corresponds to the outer iterations performed by the outer refinement loop. Thus, there is OuterNeters+1 residual norms where row "i" correspond to the outer iteration "i" and the first column specify the total number of iterations (outer and inner) that were performed till this step the second columns correspond to the residual norm at this step.

For example, let say the user specify CUSOLVER\_IRS\_REFINE\_CLASSICAL\_GMRES as a refinement solver and let say it needed 3 outer iterations to converge and 4,3,3 inner iterations at each outer respectively. This consists of 10 total iterations. Row 0 correspond to the first residual before the refinement start, so it has 0 in its first column. On row 1 which correspond to the outer iteration 1, it will be shown 4 (4 is the total number of iterations that were performed till now) on row 2, it will be 7 and on row 3 it will be 10.

As summary, let define ldh=Maxiters+1, the leading dimension of the residual matrix. then residual\_history[i] shows the total number of iterations performed at the outer iteration "i" and residual\_history[i+ldh] correspond to the norm of the residual at this outer iteration.

parameter	Memory	In/out	Meaning
infos	host	in	The cusolverDnIRSIInfos_t Infos structure
residual_history	host	output	Returns a void pointer to the matrix of the convergence history residual norms. See the description above for the relation between the residual norm datatype and the inout datatype.

### Status Returned

CUSOLVER_STATUS_SUCCESS	The operation completed successfully.
CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED	The Infos structure was not created.

<code>CUSOLVER_STATUS_INVALID_VALUE</code>	This function was called without calling <code>cusolverDnIRSInfosRequestResidual()</code> in advance.
--------------------------------------------	-------------------------------------------------------------------------------------------------------

### 2.4.1.39. `cusolverDnCreateParams()`

```
cusolverStatus_t
cusolverDnCreateParams (
    cusolverDnParams_t *params);
```

This function creates and initializes the structure of **64-bit API** to default values.

parameter	Memory	In/out	Meaning
<code>params</code>	host	output	the pointer to the structure of <b>64-bit API</b> .

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the structure was initialized successfully.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.

### 2.4.1.40. `cusolverDnDestroyParams()`

```
cusolverStatus_t
cusolverDnDestroyParams (
    cusolverDnParams_t params);
```

This function destroys and releases any memory required by the structure.

parameter	Memory	In/out	Meaning
<code>params</code>	host	input	the structure of <b>64-bit API</b> .

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the resources are released successfully.
--------------------------------------	------------------------------------------

### 2.4.1.41. `cusolverDnSetAdvOptions()`

```
cusolverStatus_t
cusolverDnSetAdvOptions (
    cusolverDnParams_t params,
    cusolverDnFunction_t function,
    cusolverAlgMode_t algo );
```

This function configures algorithm `algo` of `function`, a **64-bit API** routine.

parameter	Memory	In/out	Meaning
<code>params</code>	host	in/out	the pointer to the structure of <b>64-bit API</b> .
<code>function</code>	host	input	which routine to be configured.
<code>algo</code>	host	input	which algorithm to be configured.

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_INVALID_VALUE	wrong combination of <code>function</code> and <code>algo</code>

## 2.4.2. Dense Linear Solver Reference

This chapter describes linear solver API of cuSolverDN, including Cholesky factorization, LU with partial pivoting, QR factorization and Bunch-Kaufman (LDLT) factorization.

### 2.4.2.1. `cusolverDn<t>potrf()`

These helper functions calculate the necessary size of work buffers.

```
cusolverStatus_t
cusolverDnSpotrf_bufferSize(cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    int *Lwork );
```

```
cusolverStatus_t
cusolverDnDpotrf_bufferSize(cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    int *Lwork );
```

```
cusolverStatus_t
cusolverDnCpotrf_bufferSize(cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    int *Lwork );
```

```
cusolverStatus_t
cusolverDnZpotrf_bufferSize(cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    int *Lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSpotrf(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 float *A,
                 int lda,
                 float *Workspace,
                 int Lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnDpotrf(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 double *A,
                 int lda,
                 double *Workspace,
                 int Lwork,
                 int *devInfo );
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCpotrf(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 cuComplex *A,
                 int lda,
                 cuComplex *Workspace,
                 int Lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnZpotrf(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 cuDoubleComplex *A,
                 int lda,
                 cuDoubleComplex *Workspace,
                 int Lwork,
                 int *devInfo );
```

This function computes the Cholesky factorization of a Hermitian positive-definite matrix.

**A** is a  $n \times n$  Hermitian matrix, only lower or upper part is meaningful. The input parameter **uplo** indicates which part of the matrix is used. The function would leave other part untouched.

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, only lower triangular part of **A** is processed, and replaced by lower triangular Cholesky factor **L**.

$$A = L * L^H$$

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, only upper triangular part of **A** is processed, and replaced by upper triangular Cholesky factor **U**.

$$A = U^H * U$$

The user has to provide working space which is pointed by input parameter **Workspace**. The input parameter **Lwork** is size of the working space, and it is returned by **potrf\_bufferSize()**.

If Cholesky factorization failed, i.e. some leading minor of **A** is not positive definite, or equivalently some diagonal elements of **L** or **U** is not a real number. The output parameter **devInfo** would indicate smallest leading minor of **A** which is not positive definite.

If output parameter **devInfo** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle).

### API of potrf

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
uplo	host	input	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced.
n	host	input	number of rows and columns of matrix <b>A</b> .
<b>A</b>	device	in/out	<type> array of dimension <b>lda * n</b> with <b>lda</b> is not less than <b>max(1, n)</b> .
lda	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>Workspace</b>	device	in/out	working space, <type> array of size <b>Lwork</b> .
<b>Lwork</b>	host	input	size of <b>Workspace</b> , returned by <b>potrf_bufferSize</b> .
<b>devInfo</b>	device	output	if <b>devInfo</b> = 0, the Cholesky factorization is successful. if <b>devInfo</b> = <b>-i</b> , the <b>i-th</b> parameter is wrong (not counting handle). if <b>devInfo</b> = <b>i</b> , the leading minor of order <b>i</b> is not positive definite.

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( <b>n</b> <0 or <b>lda</b> < <b>max(1, n)</b> ).
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

### 2.4.2.2. cusolverDnPotrf()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnPotrf_bufferSize(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    cublasFillMode_t uplo,
    int64_t n,
    cudaDataType_t dataTypeA,
    const void *A,
    int64_t lda,
    cudaDataType_t computeType,
    size_t *workspaceInBytes )
```

The routine below

```
cusolverStatus_t
cusolverDnPotrf(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    cublasFillMode_t uplo,
    int64_t n,
    cudaDataType_t dataTypeA,
    void *A,
    int64_t lda,
    cudaDataType_t computeType,
    void *pBuffer,
    size_t workspaceInBytes,
    int *info )
```

computes the Cholesky factorization of a Hermitian positive-definite matrix using the generic API interface.

**A** is a  $n \times n$  Hermitian matrix, only lower or upper part is meaningful. The input parameter **uplo** indicates which part of the matrix is used. The function would leave other part untouched.

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, only lower triangular part of **A** is processed, and replaced by lower triangular Cholesky factor **L**.

$$A = L * L^H$$

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, only upper triangular part of **A** is processed, and replaced by upper triangular Cholesky factor **U**.

$$A = U^H * U$$

The user has to provide working space which is pointed by input parameter **pBuffer**. The input parameter **workspaceInBytes** is size in bytes of the working space, and it is returned by **cusolverDnPotrf\_bufferSize()**.

If Cholesky factorization failed, i.e. some leading minor of **A** is not positive definite, or equivalently some diagonal elements of **L** or **U** is not a real number. The output

parameter **info** would indicate smallest leading minor of **A** which is not positive definite.

If output parameter **info** = **-i** (less than zero), the **i**-th parameter is wrong (not counting handle).

Currently, **cusolverDnPotrf** supports only the default algorithm.

#### Table of algorithms supported by **cusolverDnPotrf**

CUSOLVER_ALG_0 or NULL	Default algorithm.
------------------------	--------------------

List of input arguments for **cusolverDnPotrf\_bufferSize** and **cusolverDnPotrf**:

#### API of **potrf**

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverDN library context.
<b>params</b>	host	input	structure with information collected by <b>cusolverDnSetAdvOptions</b> .
<b>uplo</b>	host	input	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced.
<b>n</b>	host	input	number of rows and columns of matrix <b>A</b> .
<b>dataTypeA</b>	host	in	data type of array <b>A</b> .
<b>A</b>	device	in/out	array of dimension $lda * n$ with $lda$ is not less than $\max(1, n)$ .
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>computeType</b>	host	in	data type of computation.
<b>pBuffer</b>	device	in/out	Working space. Array of type <b>void</b> of size <b>workspaceInBytes</b> bytes.
<b>workspaceInBytes</b>	host	input	size in bytes of <b>pBuffer</b> , returned by <b>cusolverDnPotrf_bufferSize</b> .
<b>info</b>	device	output	if <b>info</b> = 0, the Cholesky factorization is successful. if <b>info</b> = <b>-i</b> , the <b>i</b> -th parameter is wrong (not counting handle). if <b>info</b> = <b>i</b> , the leading minor of order <b>i</b> is not positive definite.

The generic API has two different types, **dataTypeA** is data type of the matrix **A**, **computeType** is compute type of the operation. **cusolverDnPotrf** only supports the following four combinations.

#### Valid combination of data type and compute type

DataTypeA	ComputeType	Meaning
CUDA_R_32F	CUDA_R_32F	SPOTRF
CUDA_R_64F	CUDA_R_64F	DPOTRF
CUDA_C_32F	CUDA_C_32F	CPOTRF

CUDA_C_64F	CUDA_C_64F	ZPOTRF
------------	------------	--------

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $n < 0$ or $lda < \max(1, n)$ ).
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

### 2.4.2.3. cusolverDn<t>potrs()

```

cusolverStatus_t
cusolverDnSpotrs(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 int nrhs,
                 const float *A,
                 int lda,
                 float *B,
                 int ldb,
                 int *devInfo);

cusolverStatus_t
cusolverDnDpotrs(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 int nrhs,
                 const double *A,
                 int lda,
                 double *B,
                 int ldb,
                 int *devInfo);

cusolverStatus_t
cusolverDnCpotrs(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 int nrhs,
                 const cuComplex *A,
                 int lda,
                 cuComplex *B,
                 int ldb,
                 int *devInfo);

cusolverStatus_t
cusolverDnZpotrs(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 int nrhs,
                 const cuDoubleComplex *A,
                 int lda,
                 cuDoubleComplex *B,
                 int ldb,
                 int *devInfo);

```

This function solves a system of linear equations

$$A * X = B$$

where **A** is a **n×n** Hermitian matrix, only lower or upper part is meaningful. The input parameter **uplo** indicates which part of the matrix is used. The function would leave other part untouched.

The user has to call **potrf** first to factorize matrix **A**. If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, **A** is lower triangular Cholesky factor **L** corresponding to

$A = L * L^H$ . If input parameter `uplo` is `CUBLAS_FILL_MODE_UPPER`, **A** is upper triangular Cholesky factor **U** corresponding to  $A = U^H * U$ .

The operation is in-place, i.e. matrix **X** overwrites matrix **B** with the same leading dimension `ldb`.

If output parameter `devInfo = -i` (less than zero), the `i-th` parameter is wrong (not counting handle).

### API of potrs

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolveDN library context.
<code>uplo</code>	host	input	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced.
<code>n</code>	host	input	number of rows and columns of matrix <b>A</b> .
<code>nrhs</code>	host	input	number of columns of matrix <b>x</b> and <b>B</b> .
<b>A</b>	device	input	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . <b>A</b> is either lower cholesky factor <b>L</b> or upper Cholesky factor <b>U</b> .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>B</b>	device	in/out	<type> array of dimension <code>ldb * nrhs</code> . <code>ldb</code> is not less than <code>max(1, n)</code> . As an input, <b>B</b> is right hand side matrix. As an output, <b>B</b> is the solution matrix.
<code>devInfo</code>	device	output	if <code>devInfo = 0</code> , the Cholesky factorization is successful. if <code>devInfo = -i</code> , the <code>i-th</code> parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n &lt; 0</code> , <code>nrhs &lt; 0</code> , <code>lda &lt; max(1, n)</code> or <code>ldb &lt; max(1, n)</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.2.4. cusolverDnPotrs()

```
cusolverStatus_t
cusolverDnPotrs(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    cublasFillMode_t uplo,
    int64_t n,
    int64_t nrhs,
    cudaDataType dataTypeA,
    const void *A,
    int64_t lda,
    cudaDataType dataTypeB,
    void *B,
    int64_t ldb,
    int *info)
```

This function solves a system of linear equations

$$A * X = B$$

where **A** is a  $n \times n$  Hermitian matrix, only lower or upper part is meaningful using the generic API interface. The input parameter **uplo** indicates which part of the matrix is used. The function would leave other part untouched.

The user has to call **cusolverDnPotrf** first to factorize matrix **A**. If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, **A** is lower triangular Cholesky factor **L** corresponding to  $A = L * L^H$ . If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, **A** is upper triangular Cholesky factor **U** corresponding to  $A = U^H * U$ .

The operation is in-place, i.e. matrix **X** overwrites matrix **B** with the same leading dimension **ldb**.

If output parameter **info** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle).

Currently, **cusolverDnPotrs** supports only the default algorithm.

#### Table of algorithms supported by cusolverDnPotrs

CUSOLVER_ALG_0 or NULL	Default algorithm.
------------------------	--------------------

List of input arguments for **cusolverDnPotrs**:

#### API of potrs

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolveDN library context.
<b>params</b>	host	input	structure with information collected by <b>cusolverDnSetAdvOptions</b> .
<b>uplo</b>	host	input	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced.

<b>n</b>	<b>host</b>	<b>input</b>	number of rows and columns of matrix <b>A</b> .
<b>nrhs</b>	<b>host</b>	<b>input</b>	number of columns of matrix <b>x</b> and <b>B</b> .
<b>dataTypeA</b>	<b>host</b>	<b>in</b>	data type of array <b>A</b> .
<b>A</b>	<b>device</b>	<b>input</b>	array of dimension $lda * n$ with $lda$ is not less than $\max(1, n)$ . <b>A</b> is either lower cholesky factor <b>L</b> or upper Cholesky factor <b>U</b> .
<b>lda</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>dataTypeB</b>	<b>host</b>	<b>in</b>	data type of array <b>B</b> .
<b>B</b>	<b>device</b>	<b>in/out</b>	array of dimension $ldb * nrhs$ . $ldb$ is not less than $\max(1, n)$ . As an input, <b>B</b> is right hand side matrix. As an output, <b>B</b> is the solution matrix.
<b>info</b>	<b>device</b>	<b>output</b>	if <b>info</b> = 0, the Cholesky factorization is successful. if <b>info</b> = - <i>i</i> , the <i>i</i> -th parameter is wrong (not counting handle).

The generic API has two different types, **dataTypeA** is data type of the matrix **A**, **dataTypeB** is data type of the matrix **B**. **cusolverDnPotrs** only supports the following four combinations.

#### Valid combination of data type and compute type

<b>dataTypeA</b>	<b>dataTypeB</b>	<b>Meaning</b>
CUDA_R_32F	CUDA_R_32F	SPOTRS
CUDA_R_64F	CUDA_R_64F	DPOTRS
CUDA_C_32F	CUDA_C_32F	CPOTRS
CUDA_C_64F	CUDA_C_64F	ZPOTRS

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $n < 0$ , $nrhs < 0$ , $lda < \max(1, n)$ or $ldb < \max(1, n)$ ).
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

### 2.4.2.5. cusolverDn<t>potri()

These helper functions calculate the necessary size of work buffers.

```

cusolverStatus_t
cusolverDnSpotri_bufferSize(cusolverDnHandle_t handle,
                           cublasFillMode_t uplo,
                           int n,
                           float *A,
                           int lda,
                           int *Lwork );

cusolverStatus_t
cusolverDnDpotri_bufferSize(cusolverDnHandle_t handle,
                           cublasFillMode_t uplo,
                           int n,
                           double *A,
                           int lda,
                           int *Lwork );

cusolverStatus_t
cusolverDnCpotri_bufferSize(cusolverDnHandle_t handle,
                           cublasFillMode_t uplo,
                           int n,
                           cuComplex *A,
                           int lda,
                           int *Lwork );

cusolverStatus_t
cusolverDnZpotri_bufferSize(cusolverDnHandle_t handle,
                           cublasFillMode_t uplo,
                           int n,
                           cuDoubleComplex *A,
                           int lda,
                           int *Lwork);

```

The S and D data types are real valued single and double precision, respectively.

```

cusolverStatus_t
cusolverDnSpotri(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 float *A,
                 int lda,
                 float *Workspace,
                 int Lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnDpotri(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 double *A,
                 int lda,
                 double *Workspace,
                 int Lwork,
                 int *devInfo );

```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCpotri(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 cuComplex *A,
                 int lda,
                 cuComplex *Workspace,
                 int Lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnZpotri(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 cuDoubleComplex *A,
                 int lda,
                 cuDoubleComplex *Workspace,
                 int Lwork,
                 int *devInfo );
```

This function computes the inverse of a positive-definite matrix **A** using the Cholesky factorization

$$A = L * L^H = U^H * U$$

computed by **potrf()**.

**A** is a **n×n** matrix containing the triangular factor **L** or **U** computed by the Cholesky factorization. Only lower or upper part is meaningful and the input parameter **uplo** indicates which part of the matrix is used. The function would leave the other part untouched.

If the input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, only lower triangular part of **A** is processed, and replaced by the lower triangular part of the inverse of **A**.

If the input parameter **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, only upper triangular part of **A** is processed, and replaced by the upper triangular part of the inverse of **A**.

The user has to provide the working space which is pointed to by input parameter **Workspace**. The input parameter **Lwork** is the size of the working space, returned by **potri\_bufferSize()**.

If the computation of the inverse fails, i.e. some leading minor of **L** or **U**, is null, the output parameter **devInfo** would indicate the smallest leading minor of **L** or **U** which is not positive definite.

If the output parameter **devInfo** = **-i** (less than zero), the **i-th** parameter is wrong (not counting the handle).

### API of potri

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.

<code>uplo</code>	<code>host</code>	<code>input</code>	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced.
<code>n</code>	<code>host</code>	<code>input</code>	number of rows and columns of matrix <b>A</b> .
<code>A</code>	<code>device</code>	<code>in/out</code>	<type> array of dimension <code>lda * n</code> where <code>lda</code> is not less than <code>max(1, n)</code> .
<code>lda</code>	<code>host</code>	<code>input</code>	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<code>Workspace</code>	<code>device</code>	<code>in/out</code>	working space, <type> array of size <code>Lwork</code> .
<code>Lwork</code>	<code>host</code>	<code>input</code>	size of <code>Workspace</code> , returned by <code>potri_bufferSize</code> .
<code>devInfo</code>	<code>device</code>	<code>output</code>	if <code>devInfo = 0</code> , the computation of the inverse is successful. if <code>devInfo = -i</code> , the <i>i</i> -th parameter is wrong (not counting handle). if <code>devInfo = i</code> , the leading minor of order <i>i</i> is zero.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n &lt; 0</code> or <code>lda &lt; max(1, n)</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.2.6. cusolverDn<t>getrf()

These helper functions calculate the size of work buffers needed.

```

cusolverStatus_t
cusolverDnSgetrf_bufferSize(cusolverDnHandle_t handle,
                            int m,
                            int n,
                            float *A,
                            int lda,
                            int *Lwork );

cusolverStatus_t
cusolverDnDgetrf_bufferSize(cusolverDnHandle_t handle,
                            int m,
                            int n,
                            double *A,
                            int lda,
                            int *Lwork );

cusolverStatus_t
cusolverDnCgetrf_bufferSize(cusolverDnHandle_t handle,
                            int m,
                            int n,
                            cuComplex *A,
                            int lda,
                            int *Lwork );

cusolverStatus_t
cusolverDnZgetrf_bufferSize(cusolverDnHandle_t handle,
                            int m,
                            int n,
                            cuDoubleComplex *A,
                            int lda,
                            int *Lwork );

```

The S and D data types are real single and double precision, respectively.

```

cusolverStatus_t
cusolverDnSgetrf(cusolverDnHandle_t handle,
                int m,
                int n,
                float *A,
                int lda,
                float *Workspace,
                int *devIpivot,
                int *devInfo );

cusolverStatus_t
cusolverDnDgetrf(cusolverDnHandle_t handle,
                int m,
                int n,
                double *A,
                int lda,
                double *Workspace,
                int *devIpivot,
                int *devInfo );

```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCgetrf(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 cuComplex *A,
                 int lda,
                 cuComplex *Workspace,
                 int *devIpiv,
                 int *devInfo );

cusolverStatus_t
cusolverDnZgetrf(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 cuDoubleComplex *A,
                 int lda,
                 cuDoubleComplex *Workspace,
                 int *devIpiv,
                 int *devInfo );
```

This function computes the LU factorization of a  $m \times n$  matrix

$$P * A = L * U$$

where  $\mathbf{A}$  is a  $m \times n$  matrix,  $\mathbf{P}$  is a permutation matrix,  $\mathbf{L}$  is a lower triangular matrix with unit diagonal, and  $\mathbf{U}$  is an upper triangular matrix.

The user has to provide working space which is pointed by input parameter **Workspace**. The input parameter **Lwork** is size of the working space, and it is returned by **getrf\_bufferSize()**.

If LU factorization failed, i.e. matrix  $\mathbf{A}$  ( $\mathbf{U}$ ) is singular, The output parameter **devInfo=i** indicates  $\mathbf{U}(i, i) = 0$ .

If output parameter **devInfo = -i** (less than zero), the **i-th** parameter is wrong (not counting handle).

If **devIpiv** is null, no pivoting is performed. The factorization is  $\mathbf{A} = \mathbf{L} * \mathbf{U}$ , which is not numerically stable.

No matter LU factorization failed or not, the output parameter **devIpiv** contains pivoting sequence, row **i** is interchanged with row **devIpiv(i)**.

The user can combine **getrf** and **getrs** to complete a linear solver. Please refer to appendix D.1.

Remark: **getrf** uses fastest implementation with large workspace of size  $m * n$ . The user can choose the legacy implementation with minimal workspace by **Getrf** and **cusolverDnSetAdvOptions(params, CUSOLVERDN\_GETRF, CUSOLVER\_ALG\_1)**.

#### API of getrf

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
m	host	input	number of rows of matrix $\mathbf{A}$ .

<b>n</b>	<b>host</b>	<b>input</b>	number of columns of matrix <b>A</b> .
<b>A</b>	<b>device</b>	<b>in/out</b>	<type> array of dimension <b>lda * n</b> with <b>lda</b> is not less than <b>max(1,m)</b> .
<b>lda</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>Workspace</b>	<b>device</b>	<b>in/out</b>	working space, <type> array of size <b>Lwork</b> .
<b>devI piv</b>	<b>device</b>	<b>output</b>	array of size at least <b>min(m,n)</b> , containing pivot indices.
<b>devInfo</b>	<b>device</b>	<b>output</b>	if <b>devInfo = 0</b> , the LU factorization is successful. if <b>devInfo = -i</b> , the <b>i</b> -th parameter is wrong (not counting handle). if <b>devInfo = i</b> , the $U(i,i) = 0$ .

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( <b>m,n</b> <0 or <b>lda</b> < <b>max(1,m)</b> ).
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.

### 2.4.2.7. cusolverDnGetrf()

The helper function below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnGetrf_bufferSize(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    int64_t m,
    int64_t n,
    cudaDataType dataTypeA,
    const void *A,
    int64_t lda,
    cudaDataType computeType,
    size_t *workspaceInBytes )
```

The function below

```
cusolverStatus_t
cusolverDnGetrf(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    int64_t m,
    int64_t n,
    cudaDataType_t dataTypeA,
    void *A,
    int64_t lda,
    int64_t *ipiv,
    cudaDataType_t computeType,
    void *pBuffer,
    size_t workspaceInBytes,
    int *info )
```

computes the LU factorization of a  $m \times n$  matrix

$$P * A = L * U$$

where  $\mathbf{A}$  is a  $m \times n$  matrix,  $\mathbf{P}$  is a permutation matrix,  $\mathbf{L}$  is a lower triangular matrix with unit diagonal, and  $\mathbf{U}$  is an upper triangular matrix using the generic API interface.

If LU factorization failed, i.e. matrix  $\mathbf{A}$  ( $\mathbf{U}$ ) is singular, The output parameter **info=i** indicates  $\mathbf{U}(i, i) = 0$ .

If output parameter **info = -i** (less than zero), the **i-th** parameter is wrong (not counting handle).

If **ipiv** is null, no pivoting is performed. The factorization is  $\mathbf{A} = \mathbf{L} * \mathbf{U}$ , which is not numerically stable.

No matter LU factorization failed or not, the output parameter **ipiv** contains pivoting sequence, row **i** is interchanged with row **ipiv(i)**.

The user has to provide working space which is pointed by input parameter **pBuffer**. The input parameter **workspaceInBytes** is size in bytes of the working space, and it is returned by **cusolverDnGetrf\_bufferSize()**.

The user can combine **cusolverDnGetrf** and **cusolverDnGetrs** to complete a linear solver.

Currently, **cusolverDnGetrf** supports two algorithms. To select legacy implementation, the user has to call **cusolverDnSetAdvOptions**.

#### Table of algorithms supported by cusolverDnGetrf

CUSOLVER_ALG_0 or NULL	Default algorithm. The fastest, requires a large workspace of $m * n$ elements.
CUSOLVER_ALG_1	Legacy implementation

List of input arguments for **cusolverDnGetrf\_bufferSize** and **cusolverDnGetrf**:

#### API of cusolverDnGetrf

parameter	Memory	In/out	Meaning
-----------	--------	--------	---------

<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>params</code>	host	input	structure with information collected by <code>cusolverDnSetAdvOptions</code> .
<code>m</code>	host	input	number of rows of matrix <b>A</b> .
<code>n</code>	host	input	number of columns of matrix <b>A</b> .
<code>dataTypeA</code>	host	in	data type of array <b>A</b> .
<code>A</code>	device	in/out	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, m)</code> .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<code>ipiv</code>	device	output	array of size at least <code>min(m, n)</code> , containing pivot indices.
<code>computeType</code>	host	in	data type of computation.
<code>pBuffer</code>	device	in/out	Working space. Array of type <code>void</code> of size <code>workspaceInBytes</code> bytes.
<code>workspaceInBytes</code>	host	input	size in bytes of <code>pBuffer</code> , returned by <code>cusolverDnGetrf_bufferSize</code> .
<code>info</code>	device	output	if <code>info = 0</code> , the LU factorization is successful. if <code>info = -i</code> , the <i>i</i> -th parameter is wrong (not counting <code>handle</code> ). if <code>info = i</code> , the $U(i, i) = 0$ .

The generic API has two different types, `dataTypeA` is data type of the matrix **A**, `computeType` is compute type of the operation. `cusolverDnGetrf` only supports the following four combinations.

#### valid combination of data type and compute type

<code>DataTypeA</code>	<code>ComputeType</code>	Meaning
<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	<code>SGETRF</code>
<code>CUDA_R_64F</code>	<code>CUDA_R_64F</code>	<code>DGETRF</code>
<code>CUDA_C_32F</code>	<code>CUDA_C_32F</code>	<code>CGETRF</code>
<code>CUDA_C_64F</code>	<code>CUDA_C_64F</code>	<code>ZGETRF</code>

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>m, n &lt; 0</code> or <code>lda &lt; max(1, m)</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.2.8. cusolverDn<t>getrs()

```
cusolverStatus_t
cusolverDnSgetrs(cusolverDnHandle_t handle,
                cublasOperation_t trans,
                int n,
                int nrhs,
                const float *A,
                int lda,
                const int *devIpiv,
                float *B,
                int ldb,
                int *devInfo );
```

```
cusolverStatus_t
cusolverDnDgetrs(cusolverDnHandle_t handle,
                cublasOperation_t trans,
                int n,
                int nrhs,
                const double *A,
                int lda,
                const int *devIpiv,
                double *B,
                int ldb,
                int *devInfo );
```

```
cusolverStatus_t
cusolverDnCgetrs(cusolverDnHandle_t handle,
                cublasOperation_t trans,
                int n,
                int nrhs,
                const cuComplex *A,
                int lda,
                const int *devIpiv,
                cuComplex *B,
                int ldb,
                int *devInfo );
```

```
cusolverStatus_t
cusolverDnZgetrs(cusolverDnHandle_t handle,
                cublasOperation_t trans,
                int n,
                int nrhs,
                const cuDoubleComplex *A,
                int lda,
                const int *devIpiv,
                cuDoubleComplex *B,
                int ldb,
                int *devInfo );
```

This function solves a linear system of multiple right-hand sides

$$\text{op}(A) * X = B$$

where  $\mathbf{A}$  is a  $n \times n$  matrix, and was LU-factored by `getrf`, that is, lower triangular part of  $\mathbf{A}$  is  $\mathbf{L}$ , and upper triangular part (including diagonal elements) of  $\mathbf{A}$  is  $\mathbf{U}$ .  $\mathbf{B}$  is a  $n \times nrhs$  right-hand side matrix.

The input parameter `trans` is defined by

$$\text{op}(\mathbf{A}) = \begin{cases} \mathbf{A} & \text{if trans == CUBLAS_OP_N} \\ \mathbf{A}^T & \text{if trans == CUBLAS_OP_T} \\ \mathbf{A}^H & \text{if trans == CUBLAS_OP_C} \end{cases}$$

The input parameter `devI piv` is an output of `getrf`. It contains pivot indices, which are used to permute right-hand sides.

If output parameter `devInfo = -i` (less than zero), the  $i$ -th parameter is wrong (not counting handle).

The user can combine `getrf` and `getrs` to complete a linear solver. Please refer to appendix D.1.

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>trans</code>	host	input	operation $\text{op}(\mathbf{A})$ that is non- or (conj.) transpose.
<code>n</code>	host	input	number of rows and columns of matrix $\mathbf{A}$ .
<code>nrhs</code>	host	input	number of right-hand sides.
<code>A</code>	device	input	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix $\mathbf{A}$ .
<code>devI piv</code>	device	input	array of size at least <code>n</code> , containing pivot indices.
<code>B</code>	device	output	<type> array of dimension <code>ldb * nrhs</code> with <code>ldb</code> is not less than <code>max(1, n)</code> .
<code>ldb</code>	host	input	leading dimension of two-dimensional array used to store matrix $\mathbf{B}$ .
<code>devInfo</code>	device	output	if <code>devInfo = 0</code> , the operation is successful. if <code>devInfo = -i</code> , the $i$ -th parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n &lt; 0</code> or <code>lda &lt; max(1, n)</code> or <code>ldb &lt; max(1, n)</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.2.9. cusolverDnGetrs()

```
cusolverStatus_t
cusolverDnGetrs(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    cublasOperation_t trans,
    int64_t n,
    int64_t nrhs,
    cudaDataType_t dataTypeA,
    const void *A,
    int64_t lda,
    const int64_t *ipiv,
    cudaDataType_t dataTypeB,
    void *B,
    int64_t ldb,
    int *info )
```

This function solves a linear system of multiple right-hand sides

$$\text{op}(\mathbf{A}) * \mathbf{X} = \mathbf{B}$$

where  $\mathbf{A}$  is a  $n \times n$  matrix, and was LU-factored by `cusolverDnGetrf`, that is, lower triangular part of  $\mathbf{A}$  is  $\mathbf{L}$ , and upper triangular part (including diagonal elements) of  $\mathbf{A}$  is  $\mathbf{U}$ .  $\mathbf{B}$  is a  $n \times \text{nrhs}$  right-hand side matrix using the generic API interface.

The input parameter `trans` is defined by

$$\text{op}(\mathbf{A}) = \begin{cases} \mathbf{A} & \text{if trans == CUBLAS_OP_N} \\ \mathbf{A}^T & \text{if trans == CUBLAS_OP_T} \\ \mathbf{A}^H & \text{if trans == CUBLAS_OP_C} \end{cases}$$

The input parameter `ipiv` is an output of `cusolverDnGetrf`. It contains pivot indices, which are used to permute right-hand sides.

If output parameter `info = -i` (less than zero), the  $i$ -th parameter is wrong (not counting handle).

The user can combine `cusolverDnGetrf` and `cusolverDnGetrs` to complete a linear solver.

Currently, `cusolverDnGetrs` supports only the default algorithm.

#### Table of algorithms supported by cusolverDnGetrs

CUSOLVER_ALG_0 or NULL	Default algorithm.
------------------------	--------------------

List of input arguments for `cusolverDnGetrs`:

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>params</code>	host	input	structure with information collected by <code>cusolverDnSetAdvOptions</code> .
<code>trans</code>	host	input	operation <code>op</code> ( $\mathbf{A}$ ) that is non- or (conj.) transpose.

<b>n</b>	host	input	number of rows and columns of matrix <b>A</b> .
<b>nrhs</b>	host	input	number of right-hand sides.
<b>dataTypeA</b>	host	in	data type of array <b>A</b> .
<b>A</b>	device	input	array of dimension $lda * n$ with $lda$ is not less than $\max(1, n)$ .
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>ipiv</b>	device	input	array of size at least $n$ , containing pivot indices.
<b>dataTypeB</b>	host	in	data type of array <b>B</b> .
<b>B</b>	device	output	<type> array of dimension $ldb * nrhs$ with $ldb$ is not less than $\max(1, n)$ .
<b>ldb</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>B</b> .
<b>info</b>	device	output	if <b>info</b> = 0, the operation is successful. if <b>info</b> = - <i>i</i> , the <i>i</i> -th parameter is wrong (not counting handle).

The generic API has two different types, **dataTypeA** is data type of the matrix **A** and **dataTypeB** is data type of the matrix **B**. **cusolverDnGetrs** only supports the following four combinations.

#### Valid combination of data type and compute type

DataTypeA	dataTypeB	Meaning
CUDA_R_32F	CUDA_R_32F	SGETRS
CUDA_R_64F	CUDA_R_64F	DGETRS
CUDA_C_32F	CUDA_C_32F	CGETRS
CUDA_C_64F	CUDA_C_64F	ZGETRS

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $n < 0$ or $lda < \max(1, n)$ or $ldb < \max(1, n)$ ).
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

#### 2.4.2.10. cusolverDn<t1><t2>gesv()

These functions are modelled after functions DSGESV and ZCGESV from LAPACK. They compute the solution of a system of linear equations with one or multiple right hand sides using mixed precision iterative refinement techniques based on the LU factorization Xgesv. These functions are similar in term of functionalities to the full

precision LU solver (Xgesv, where X denotes Z,C,D,S) but it uses lower precision internally in order to provide faster time to solution, from here comes the name mixed precision. Mixed precision iterative refinement techniques means that the solver compute an LU factorization in lower precision and then iteratively refine the solution to achieve the accuracy of the Inputs/Outputs datatype precision. The <t1> corresponds to the Inputs/Outputs datatype precision while <t2> represent the internal lower precision at which the factorization will be carried on.

$$A \times X = B$$

Where **A** is **n-by-n** matrix and **X** and **B** are **n-by-nrhs** matrices.

Functions API are designed to be as close as possible to LAPACK API to be considered as a quick and easy drop-in replacement. Parameters and behavior are mostly the same as LAPACK counterparts. Description of these functions and differences from LAPACK is given below. <t1><t2>gesv() functions are designated by two floating point precisions. The <t1> corresponds to the main precision (e.g., Inputs/Outputs datatype precision) and the <t2> represent the internal lower precision at which the factorization will be carried on. cusolver<t1><t2>gesv() first attempts to factorize the matrix in lower precision and use this factorization within an iterative refinement procedure to obtain a solution with same normwise backward error as the main precision <t1>. If the approach fails to converge, then the method fallback to the main precision factorization and solve (Xgesv) such a way that there is always a good solution at the output of these functions. If <t2> is equal to <t1>, then it is not a mixed precision process but rather a full one precision factorisation, solve and refinement within the same main precision.

The iterative refinement process is stopped if

$$\mathbf{ITER} > \mathbf{ITERMAX}$$

or for all the RHS we have:

$$\mathbf{RNRM} < \mathbf{SQRT(N)} * \mathbf{XNRM} * \mathbf{ANRM} * \mathbf{EPS} * \mathbf{BWDMAX}$$

where

- ▶ ITER is the number of the current iteration in the iterative refinement process
- ▶ RNRM is the infinity-norm of the residual
- ▶ XNRM is the infinity-norm of the solution
- ▶ ANRM is the infinity-operator-norm of the matrix A
- ▶ EPS is the machine epsilon that matches LAPACK <t1>LAMCH('Epsilon')

The value ITERMAX and BWDMAX are fixed to 50 and 1.0 respectively.

The function returns value describes the results of the solving process. A CUSOLVER\_STATUS\_SUCCESS indicates that the function finished with success otherwise, it indicates if one of the API arguments is incorrect, or if the function did not finish with success. More details about the error will be in the **niters** and the **dinfo** API parameters. See their description below for more details. User should provide the required workspace allocated on device memory. The amount of bytes required can be queried by calling the respective function <t1><t2>gesv\_bufferSize().

Note that in addition to the two mixed precision functions available in LAPACK (e.g., `dsgesv` and `zcgsv`), we provide a large set of mixed precision functions that include half, bfloat and tensorfloat as a lower precision as well as same precision functions (e.g., main and lowest precision are equal <t2> is equal to <t1>). The following table specifies which precisions will be used for which interface function.

Tensor Float (TF32), introduced with NVIDIA Ampere Architecture GPUs, is the most robust tensor core accelerated compute mode for the iterative refinement solver. It is able to solve the widest range of problems in HPC arising from different applications and provides up to 4X and 5X speedup for real and complex systems, respectively. On Volta and Turing architecture GPUs, half precision tensor core acceleration is recommended. In cases where the iterative refinement solver fails to converge to the desired accuracy (main precision, INOUT data precision), it is recommended to use main precision as internal lowest precision (i.e., `cusolverDn[DD,ZZ]gesv` for the FP64 case).

### Supported combinations of floating point precisions for `cusolver <t1><t2>gesv()` functions

Interface function	Main precision (matrix, rhs and solution datatype)	Lowest precision allowed to be used internally
<code>cusolverDnZZgesv</code>	<code>cuDoubleComplex</code>	double complex
<code>cusolverDnZCgesv</code> *has LAPACK counterparts	<code>cuDoubleComplex</code>	single complex
<code>cusolverDnZKgesv</code>	<code>cuDoubleComplex</code>	half complex
<code>cusolverDnZEgesv</code>	<code>cuDoubleComplex</code>	bfloat complex
<code>cusolverDnZYgesv</code>	<code>cuDoubleComplex</code>	tensorfloat complex
<code>cusolverDnCCgesv</code>	<code>cuComplex</code>	single complex
<code>cusolverDnCKgesv</code>	<code>cuComplex</code>	half complex
<code>cusolverDnCEgesv</code>	<code>cuComplex</code>	bfloat complex
<code>cusolverDnCYgesv</code>	<code>cuComplex</code>	tensorfloat complex
<code>cusolverDnDDgesv</code>	double	double
<code>cusolverDnDSgesv</code> *has LAPACK counterparts	double	single
<code>cusolverDnDHgesv</code>	double	half
<code>cusolverDnDBgesv</code>	double	bfloat
<code>cusolverDnDXgesv</code>	double	tensorfloat
<code>cusolverDnSSgesv</code>	float	single
<code>cusolverDnSHgesv</code>	float	half
<code>cusolverDnSBgesv</code>	float	bfloat
<code>cusolverDnSXgesv</code>	float	tensorfloat

`cusolverDn<t1><t2>gesv_bufferSize()` functions will return workspace buffer size in bytes required for the corresponding `cusolverDn<t1><t2>gesv()` function.

```
cusolverStatus_t
cusolverDnZZgesv_bufferSize(
    cusolverHandle_t      handle,
    int                   n,
    int                   nrhs,
    cuDoubleComplex      *dA,
    int                   ldda,
    int                   *dipiv,
    cuDoubleComplex      *dB,
    int                   lddb,
    cuDoubleComplex      *dX,
    int                   lddx,
    void                  *dwork,
    size_t                *lwork_bytes);
```

```
cusolverStatus_t
cusolverDnZCgesv_bufferSize(
    cusolverHandle_t      handle,
    int                   n,
    int                   nrhs,
    cuDoubleComplex      *dA,
    int                   ldda,
    int                   *dipiv,
    cuDoubleComplex      *dB,
    int                   lddb,
    cuDoubleComplex      *dX,
    int                   lddx,
    void                  *dwork,
    size_t                *lwork_bytes);
```

```
cusolverStatus_t
cusolverDnZKgesv_bufferSize(
    cusolverHandle_t      handle,
    int                   n,
    int                   nrhs,
    cuDoubleComplex      *dA,
    int                   ldda,
    int                   *dipiv,
    cuDoubleComplex      *dB,
    int                   lddb,
    cuDoubleComplex      *dX,
    int                   lddx,
    void                  *dwork,
    size_t                *lwork_bytes);
```

```
cusolverStatus_t
cusolverDnZEgesv_bufferSize(
    cusolverHandle_t      handle,
    int                   n,
    int                   nrhs,
    cuDoubleComplex      *dA,
    int                   ldda,
    int                   *dipiv,
    cuDoubleComplex      *dB,
    int                   lddb,
    cuDoubleComplex      *dX,
    int                   lddx,
    void                  *dwork,
    size_t                *lwork_bytes);
```

```
cusolverStatus_t
cusolverDnZYgesv_bufferSize(
    cusolverHandle_t      handle,
    int                   n,
    int                   nrhs,
```

Parameters of `cusolverDn<T1><T2>gesv_bufferSize()` functions

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	Handle to the <code>cusolverDN</code> library context.
<code>n</code>	host	input	Number of rows and columns of square matrix <b>A</b> . Should be non-negative.
<code>nrhs</code>	host	input	Number of right hand sides to solve. Should be non-negative.
<code>dA</code>	device	None	Matrix <b>A</b> with size <code>n-by-n</code> . Can be <code>NULL</code> .
<code>ldda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> . <code>ldda &gt;= n</code> .
<code>dipiv</code>	device	None	Pivoting sequence. Not used and can be <code>NULL</code> .
<code>dB</code>	device	None	Set of right hand sides <b>B</b> of size <code>n-by-nrhs</code> . Can be <code>NULL</code> .
<code>lddb</code>	host	input	leading dimension of two-dimensional array used to store matrix of right hand sides <b>B</b> . <code>lddb &gt;= n</code> .
<code>dX</code>	device	None	Set of solution vectors <b>X</b> of size <code>n-by-nrhs</code> . Can be <code>NULL</code> .
<code>lddx</code>	host	input	leading dimension of two-dimensional array used to store matrix of solution vectors <b>X</b> . <code>ldx &gt;= n</code> .
<code>dwork</code>	device	none	Pointer to device workspace. Not used and can be <code>NULL</code> .

<b>lwork_bytes</b>	<b>host</b>	<b>output</b>	Pointer to a variable where required size of temporary workspace in bytes will be stored. Can't be NULL.
--------------------	-------------	---------------	----------------------------------------------------------------------------------------------------------

```
cusolverStatus_t cusolverDnZZgesv(
    cusolverDnHandle_t handle,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    int * dipiv,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

```
cusolverStatus_t cusolverDnZCgesv(
    cusolverDnHandle_t handle,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    int * dipiv,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

```
cusolverStatus_t cusolverDnZKgesv(
    cusolverDnHandle_t handle,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    int * dipiv,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

```
cusolverStatus_t cusolverDnZEgesv(
    cusolverDnHandle_t handle,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    int * dipiv,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

Parameters of `cusolverDn<T1><T2>gesv()` functions

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	Handle to the <code>cusolverDN</code> library context.
<code>n</code>	host	input	Number of rows and columns of square matrix <b>A</b> . Should be non-negative.
<code>nrhs</code>	host	input	Number of right hand sides to solve. Should be non-negative.
<code>dA</code>	device	in/out	Matrix <b>A</b> with size <code>n-by-n</code> . Can't be <code>NULL</code> . On return - unchanged if the iterative refinement process converged. If not - will contains the factorization of the matrix <b>A</b> in the main precision <code>&lt;T1&gt;</code> ( $A = P * L * U$ , where <b>P</b> - permutation matrix defined by vector <code>ipiv</code> , <b>L</b> and <b>U</b> - lower and upper triangular matrices).
<code>ldda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> . <code>ldda &gt;= n</code> .
<code>dipiv</code>	device	output	Vector that defines permutation for the factorization - row <code>i</code> was interchanged with row <code>ipiv[i]</code>
<code>dB</code>	device	input	Set of right hand sides <b>B</b> of size <code>n-by-nrhs</code> . Can't be <code>NULL</code> .
<code>lddb</code>	host	input	leading dimension of two-dimensional array used to store matrix of right hand sides <b>B</b> . <code>lddb &gt;= n</code> .
<code>dX</code>	device	output	Set of solution vectors <b>X</b> of size <code>n-by-nrhs</code> . Can't be <code>NULL</code> .
<code>lddx</code>	host	input	leading dimension of two-dimensional array used to store matrix of solution vectors <b>X</b> . <code>lddx &gt;= n</code> .
<code>dWorkspace</code>	device	input	Pointer to an allocated workspace in device memory of size <code>lwork_bytes</code> .
<code>lwork_bytes</code>	host	input	Size of the allocated device workspace. Should be at least what was returned by <code>cusolverDn&lt;T1&gt;&lt;T2&gt;gesv_bufferSize()</code> function
<code>niters</code>	host	output	If iter is <ul style="list-style-type: none"> <li>▶ <code>&lt;0</code> : iterative refinement has failed, main precision (Inputs/Outputs precision) factorization has been performed.</li> <li>▶ <code>-1</code> : taking into account machine parameters, <code>n</code>, <code>nrhs</code>, it is a priori not worth working in lower precision</li> <li>▶ <code>-2</code> : overflow of an entry when moving from main to lower precision</li> <li>▶ <code>-3</code> : failure during the factorization</li> <li>▶ <code>-5</code> : overflow occurred during computation</li> </ul>

			<ul style="list-style-type: none"> <li>▶ -50: solver stopped the iterative refinement after reaching maximum allowed iterations.</li> <li>▶ &gt;0 : iter is a number of iterations solver performed to reach convergence criteria</li> </ul>
<b>dinfo</b>	<b>device</b>	<b>output</b>	Status of the IRS solver on the return. If 0 - solve was successful. If dinfo = -i then i-th argument is not valid. If dinfo = i, then $\sigma(i, i)$ computed in main precision is exactly zero. The factorization has been completed, but the factor U is exactly singular, so the solution could not be computed.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed, for example: <ul style="list-style-type: none"> <li>▶ <math>n &lt; 0</math></li> <li>▶ <math>lda &lt; \max(1, n)</math></li> <li>▶ <math>ldb &lt; \max(1, n)</math></li> <li>▶ <math>ldx &lt; \max(1, n)</math></li> </ul>
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the IRS solver supports compute capability 7.0 and above. The lowest precision options CUSOLVER_[CR]_16BF and CUSOLVER_[CR]_TF32 are only available on compute capability 8.0 and above.
<b>CUSOLVER_STATUS_INVALID_WORKSPACE</b>	lwork_bytes is smaller than the required workspace.
<b>CUSOLVER_STATUS_IRS_OUT_OF_RANGE</b>	Numerical error related to niters <0, seeniters description for more details.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal error occurred, check the <b>dinfo</b> and the <b>niters</b> arguments for more details.

#### 2.4.2.11. cusolverDnIRSxgesv()

This function is designed to perform same functionality as **cusolverDn<T1><T2>gesv()** functions, but wrapped in a more generic and expert interface that gives user more control to parametrize the function as well as it provides more informations on output. **cusolverDnIRSxgesv()** allows additional control of the solver parameters such as setting:

- ▶ the main precision (Inputs/Outputs precision) of the solver,
- ▶ the lowest precision to be used internally by the solver,
- ▶ the refinement solver type,
- ▶ the maximum allowed number of iterations in the refinement phase,
- ▶ the tolerance of the refinement solver,

- ▶ the fallback to main precision,
- ▶ and more other.

through the configuration parameters structure `gesv_irs_params` and its helper functions. For more details about what configuration can be set and its meaning please refer to all the functions in the `cuSolverDN Helper Function Section` that start with `cusolverDnIRSParamsxxxx()`. Moreover, `cusolverDnIRSXgesv()` provides additional informations on the output such as the convergence history (e.g., the residual norms) at each iteration and the number of iterations needed to converge. For more details about what informations can be retrieved and its meaning please refer to all the functions in the `cuSolverDN Helper Function Section` that start with `cusolverDnIRSInfosxxxx()`

The function returns value describes the results of the solving process. A `CUSOLVER_STATUS_SUCCESS` indicates that the function finished with success otherwise, it indicates if one of the API arguments is incorrect, or if the configurations of `params/infos` structure is incorrect or if the function did not finish with success. More details about the error can be found by checking the `niters` and the `dinfo` API parameters. See their description below for further details. User should provide the required workspace allocated on device for the `cusolverDnIRSXgesv()` function. The amount of bytes required for the function can be queried by calling the respective function `cusolverDnIRSXgesv_bufferSize()`. Note that, if the user would like a particular configuration to be set via the `params` structure, it should be set before the call to `cusolverDnIRSXgesv_bufferSize()` to get the size of the required workspace.

Tensor Float (TF32), introduced with NVIDIA Ampere Architecture GPUs, is the most robust tensor core accelerated compute mode for the iterative refinement solver. It is able to solve the widest range of problems in HPC arising from different applications and provides up to 4X and 5X speedup for real and complex systems, respectively. On Volta and Turing architecture GPUs, half precision tensor core acceleration is recommended. In cases where the iterative refinement solver fails to converge to the desired accuracy (main precision, INOUT data precision), it is recommended to use main precision as internal lowest precision.

The following table provides all possible combinations values for the lowest precision corresponding to the Inputs/Outputs data type. Note that if the lowest precision matches the Inputs/Outputs datatype, then the main precision factorization will be used.

#### Supported Inputs/Outputs data type and lower precision for the IRS solver

Inputs/Outputs Data Type (e.g., main precision)	Supported values for the lowest precision
<code>CUSOLVER_C_64F</code>	<code>CUSOLVER_C_64F</code> , <code>CUSOLVER_C_32F</code> , <code>CUSOLVER_C_16F</code> , <code>CUSOLVER_C_16BF</code> , <code>CUSOLVER_C_TF32</code>
<code>CUSOLVER_C_32F</code>	<code>CUSOLVER_C_32F</code> , <code>CUSOLVER_C_16F</code> , <code>CUSOLVER_C_16BF</code> , <code>CUSOLVER_C_TF32</code>
<code>CUSOLVER_R_64F</code>	<code>CUSOLVER_R_64F</code> , <code>CUSOLVER_R_32F</code> , <code>CUSOLVER_R_16F</code> , <code>CUSOLVER_R_16BF</code> , <code>CUSOLVER_R_TF32</code>
<code>CUSOLVER_R_32F</code>	<code>CUSOLVER_R_32F</code> , <code>CUSOLVER_R_16F</code> , <code>CUSOLVER_R_16BF</code> , <code>CUSOLVER_R_TF32</code>

The `cusolverDnIRSXgesv_bufferSize()` function returns the required workspace buffer size in bytes for the corresponding `cusolverDnXgesv()` call with the given `gesv_irs_params` configuration.

```
cusolverStatus_t
cusolverDnIRSXgesv_bufferSize(
    cusolverDnHandle_t      handle,
    cusolverDnIRSParams_t  gesv_irs_params,
    cusolver_int_t          n,
    cusolver_int_t          nrhs,
    size_t                  * lwork_bytes);
```

### Parameters of `cusolverDnIRSXgesv_bufferSize()` functions

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	Handle to the <code>cusolverDn</code> library context.
<code>params</code>	host	input	Xgesv configuration parameters
<code>n</code>	host	input	Number of rows and columns of the square matrix <b>A</b> . Should be non-negative.
<code>nrhs</code>	host	input	Number of right hand sides to solve. Should be non-negative. Note that, <code>nrhs</code> is limited to 1 if the selected IRS refinement solver is <code>CUSOLVER_IRS_REFINE_GMRES</code> , <code>CUSOLVER_IRS_REFINE_GMRES_GMRES</code> , <code>CUSOLVER_IRS_REFINE_CLASSICAL_GMRES</code> .
<code>lwork_bytes</code>	host	out	Pointer to a variable, where the required size in bytes, of the workspace will be stored after a call to <code>cusolverDnIRSXgesv_bufferSize</code> . Can't be NULL.

```
cusolverStatus_t cusolverDnIRSXgesv(
    cusolverDnHandle_t      handle,
    cusolverDnIRSParams_t  gesv_irs_params,
    cusolverDnIRSInfos_t   gesv_irs_infos,
    int                     n,
    int                     nrhs,
    void                    * dA,
    int                     * ldda,
    void                    * dB,
    int                     * lddb,
    void                    * dX,
    int                     * lddx,
    void                    * dWorkspace,
    size_t                  lwork_bytes,
    int                     * dinfo);
```

### Parameters of `cusolverDnIRSXgesv()` functions

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	Handle to the <code>cusolverDn</code> library context.

<code>gesv_irs_params</code>	<code>host</code>	<code>input</code>	Configuration parameters structure, can serve one or more calls to any IRS solver
<code>gesv_irs_infos</code>	<code>host</code>	<code>in/out</code>	Info structure, where information about a particular solve will be stored. The <code>gesv_irs_infos</code> structure correspond to a particular call. Thus different calls requires different <code>gesv_irs_infos</code> structure otherwise, it will be overwritten.
<code>n</code>	<code>host</code>	<code>input</code>	Number of rows and columns of square matrix <b>A</b> . Should be non-negative.
<code>nrhs</code>	<code>host</code>	<code>input</code>	Number of right hand sides to solve. Should be non-negative. Note that, <code>nrhs</code> is limited to 1 if the selected IRS refinement solver is <code>CUSOLVER_IRS_REFINE_GMRES</code> , <code>CUSOLVER_IRS_REFINE_GMRES_GMRES</code> , <code>CUSOLVER_IRS_REFINE_CLASSICAL_GMRES</code> .
<code>dA</code>	<code>device</code>	<code>in/out</code>	Matrix <b>A</b> with size <code>n-by-n</code> . Can't be <code>NULL</code> . On return - will contain the factorization of the matrix <b>A</b> in the main precision ( $\mathbf{A} = \mathbf{P} * \mathbf{L} * \mathbf{U}$ , where <b>P</b> - permutation matrix defined by vector <code>ipiv</code> , <b>L</b> and <b>U</b> - lower and upper triangular matrices) if the iterative refinement solver was set to <code>CUSOLVER_IRS_REFINE_NONE</code> and the lowest precision is equal to the main precision (Inputs/Outputs datatype), or if the iterative refinement solver did not converge and the fallback to main precision was enabled (fallback enabled is the default setting); - unchanged otherwise.
<code>ldda</code>	<code>host</code>	<code>input</code>	leading dimension of two-dimensional array used to store matrix <b>A</b> . <code>ldda</code> $\geq$ <code>n</code> .
<code>dB</code>	<code>device</code>	<code>input</code>	Set of right hand sides <b>B</b> of size <code>n-by-nrhs</code> . Can't be <code>NULL</code> .
<code>lddb</code>	<code>host</code>	<code>input</code>	leading dimension of two-dimensional array used to store matrix of right hand sides <b>B</b> . <code>ldb</code> $\geq$ <code>n</code> .
<code>dX</code>	<code>device</code>	<code>output</code>	Set of solution vectors <b>X</b> of size <code>n-by-nrhs</code> . Can't be <code>NULL</code> .
<code>lddx</code>	<code>host</code>	<code>input</code>	leading dimension of two-dimensional array used to store matrix of solution vectors <b>X</b> . <code>ldx</code> $\geq$ <code>n</code> .
<code>dWorkspace</code>	<code>device</code>	<code>input</code>	Pointer to an allocated workspace in device memory of size <code>lwork_bytes</code> .
<code>lwork_bytes</code>	<code>host</code>	<code>input</code>	Size of the allocated device workspace. Should be at least what was returned by <code>cusolverDnIRSXgesv_bufferSize()</code> function
<code>niters</code>	<code>host</code>	<code>output</code>	If iter is

			<ul style="list-style-type: none"> <li>▶ &lt;0 : iterative refinement has failed, main precision (Inputs/Outputs precision) factorization has been performed if fallback is enabled.</li> <li>▶ -1 : taking into account machine parameters, n, nrhs, it is a priori not worth working in lower precision</li> <li>▶ -2 : overflow of an entry when moving from main to lower precision</li> <li>▶ -3 : failure during the factorization</li> <li>▶ -5 : overflow occurred during computation</li> <li>▶ -maxiter: solver stopped the iterative refinement after reaching maximum allowed iterations.</li> <li>▶ &gt;0 : iter is a number of iterations solver performed to reach convergence criteria</li> </ul>
dinfo	device	output	Status of the IRS solver on the return. If 0 - solve was successful. If dinfo = -i then i-th argument is not valid. If dinfo = i, then $\sigma(i, i)$ computed in main precision is exactly zero. The factorization has been completed, but the factor U is exactly singular, so the solution could not be computed.

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed, for example: <ul style="list-style-type: none"> <li>▶ <math>n &lt; 0</math></li> <li>▶ <math>lda &lt; \max(1, n)</math></li> <li>▶ <math>ldb &lt; \max(1, n)</math></li> <li>▶ <math>ldx &lt; \max(1, n)</math></li> </ul>
CUSOLVER_STATUS_ARCH_MISMATCH	the IRS solver supports compute capability 7.0 and above. The lowest precision options CUSOLVER_[CR]_16BF and CUSOLVER_[CR]_TF32 are only available on compute capability 8.0 and above.
CUSOLVER_STATUS_INVALID_WORKSPACE	lwork_bytes is smaller than the required workspace. Could happen if the users called <code>cusolverDnIRSXgesv_bufferSize()</code> function, then changed some of the configurations setting such as the lowest precision.
CUSOLVER_STATUS_IRS_OUT_OF_RANGE	Numerical error related to niters <0, see niters description for more details.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal error occurred, check the dinfo and the niters arguments for more details.

<code>CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE</code>	The configuration parameter <code>gesv_irs_params</code> structure was not created.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID</code>	One of the configuration parameter in the <code>gesv_irs_params</code> structure is not valid.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID_PREC</code>	The main and/or the lowest precision configuration parameter in the <code>gesv_irs_params</code> structure is not valid, check the table above for the supported combinations.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID_MAXITER</code>	The maxiter configuration parameter in the <code>gesv_irs_params</code> structure is not valid.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID_REFINE</code>	The refinement solver configuration parameter in the <code>gesv_irs_params</code> structure is not valid.
<code>CUSOLVER_STATUS_IRS_NOT_SUPPORTED</code>	One of the configuration parameter in the <code>gesv_irs_params</code> structure is not supported. For example if <code>nrhs &gt; 1</code> , and refinement solver was set to <code>CUSOLVER_IRS_REFINE_GMRES</code>
<code>CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED</code>	The information structure <code>gesv_irs_infos</code> was not created.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	CPU memory allocation failed, most likely during the allocation of the residual array that store the residual norms.

### 2.4.2.12. cusolverDn<t>geqrf()

These helper functions calculate the size of work buffers needed.

```
cusolverStatus_t
cusolverDnSgeqrf_bufferSize(cusolverDnHandle_t handle,
                           int m,
                           int n,
                           float *A,
                           int lda,
                           int *Lwork );

cusolverStatus_t
cusolverDnDgeqrf_bufferSize(cusolverDnHandle_t handle,
                           int m,
                           int n,
                           double *A,
                           int lda,
                           int *Lwork );

cusolverStatus_t
cusolverDnCgeqrf_bufferSize(cusolverDnHandle_t handle,
                            int m,
                            int n,
                            cuComplex *A,
                            int lda,
                            int *Lwork );

cusolverStatus_t
cusolverDnZgeqrf_bufferSize(cusolverDnHandle_t handle,
                            int m,
                            int n,
                            cuDoubleComplex *A,
                            int lda,
                            int *Lwork );
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSgeqrf(cusolverDnHandle_t handle,
                int m,
                int n,
                float *A,
                int lda,
                float *TAU,
                float *Workspace,
                int Lwork,
                int *devInfo );

cusolverStatus_t
cusolverDnDgeqrf(cusolverDnHandle_t handle,
                int m,
                int n,
                double *A,
                int lda,
                double *TAU,
                double *Workspace,
                int Lwork,
                int *devInfo );
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCgeqrf(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 cuComplex *A,
                 int lda,
                 cuComplex *TAU,
                 cuComplex *Workspace,
                 int Lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnZgeqrf(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 cuDoubleComplex *A,
                 int lda,
                 cuDoubleComplex *TAU,
                 cuDoubleComplex *Workspace,
                 int Lwork,
                 int *devInfo );
```

This function computes the QR factorization of a  $m \times n$  matrix

$$A = Q * R$$

where  $A$  is a  $m \times n$  matrix,  $Q$  is a  $m \times n$  matrix, and  $R$  is a  $n \times n$  upper triangular matrix.

The user has to provide working space which is pointed by input parameter **Workspace**. The input parameter **Lwork** is size of the working space, and it is returned by **geqrf\_bufferSize()**.

The matrix  $R$  is overwritten in upper triangular part of  $A$ , including diagonal elements.

The matrix  $Q$  is not formed explicitly, instead, a sequence of householder vectors are stored in lower triangular part of  $A$ . The leading nonzero element of householder vector is assumed to be 1 such that output parameter **TAU** contains the scaling factor  $\tau$ . If  $v$  is original householder vector,  $q$  is the new householder vector corresponding to  $\tau$ , satisfying the following relation

$$I - 2 * v * v^H = I - \tau * q * q^H$$

If output parameter **devInfo** =  $-i$  (less than zero), the  $i$ -th parameter is wrong (not counting handle).

### API of geqrf

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
m	host	input	number of rows of matrix A.
n	host	input	number of columns of matrix A.
A	device	in/out	<type> array of dimension $lda * n$ with $lda$ is not less than $\max(1, m)$ .

<b>lda</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>TAU</b>	<b>device</b>	<b>output</b>	<type> array of dimension at least $\min(m, n)$ .
<b>Workspace</b>	<b>device</b>	<b>in/out</b>	working space, <type> array of size <b>Lwork</b> .
<b>Lwork</b>	<b>host</b>	<b>input</b>	size of working array <b>Workspace</b> .
<b>devInfo</b>	<b>device</b>	<b>output</b>	if <b>devInfo</b> = 0, the LU factorization is successful. if <b>devInfo</b> = -i, the i-th parameter is wrong (not counting handle).

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( $m, n < 0$ or $lda < \max(1, m)$ ).
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.

### 2.4.2.13. cusolverDnGexrf()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnGexrf_bufferSize(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    int64_t m,
    int64_t n,
    cudaDataType dataTypeA,
    const void *A,
    int64_t lda,
    cudaDataType dataTypeTau,
    const void *tau,
    cudaDataType computeType,
    size_t *workspaceInBytes )
```

The routine bellow

```
cusolverStatus_t
cusolverDnGeqrf(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    int64_t m,
    int64_t n,
    cudaDataType_t dataTypeA,
    void *A,
    int64_t lda,
    cudaDataType_t dataTypeTau,
    void *tau,
    cudaDataType_t computeType,
    void *pBuffer,
    size_t workspaceInBytes,
    int *info )
```

computes the QR factorization of a  $m \times n$  matrix

$$A = Q * R$$

where  $\mathbf{A}$  is a  $m \times n$  matrix,  $\mathbf{Q}$  is a  $m \times n$  matrix, and  $\mathbf{R}$  is a  $n \times n$  upper triangular matrix using the generic API interface.

The user has to provide working space which is pointed by input parameter **pBuffer**. The input parameter **workspaceInBytes** is size in bytes of the working space, and it is returned by **cusolverDnGeqrf\_bufferSize()**.

The matrix  $\mathbf{R}$  is overwritten in upper triangular part of  $\mathbf{A}$ , including diagonal elements.

The matrix  $\mathbf{Q}$  is not formed explicitly, instead, a sequence of householder vectors are stored in lower triangular part of  $\mathbf{A}$ . The leading nonzero element of householder vector is assumed to be 1 such that output parameter **TAU** contains the scaling factor  $\tau$ . If  $\mathbf{v}$  is original householder vector,  $\mathbf{q}$  is the new householder vector corresponding to  $\tau$ , satisfying the following relation

$$I - 2 * \mathbf{v} * \mathbf{v}^H = I - \tau * \mathbf{q} * \mathbf{q}^H$$

If output parameter **info** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle).

Currently, **cusolverDnGeqrf** supports only the default algorithm.

#### Table of algorithms supported by cusolverDnGeqrf

CUSOLVER_ALG_0 or NULL	Default algorithm.
------------------------	--------------------

List of input arguments for **cusolverDnGeqrf\_bufferSize** and **cusolverDnGeqrf**:

#### API of geqrf

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
params	host	input	structure with information collected by <b>cusolverDnSetAdvOptions</b> .
m	host	input	number of rows of matrix $\mathbf{A}$ .

<b>n</b>	host	input	number of columns of matrix <b>A</b> .
<b>dataTypeA</b>	host	in	data type of array <b>A</b> .
<b>A</b>	device	in/out	array of dimension $lda * n$ with $lda$ is not less than $\max(1, m)$ .
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>TAU</b>	device	output	array of dimension at least $\min(m, n)$ .
<b>computeType</b>	host	in	data type of computation.
<b>pBuffer</b>	device	in/out	Working space. Array of type <code>void</code> of size <code>workspaceInBytes</code> bytes.
<b>workspaceInBytes</b>	host	input	size in bytes of working array <code>pBuffer</code> .
<b>info</b>	device	output	if <code>info = 0</code> , the LU factorization is successful. if <code>info = -i</code> , the <i>i</i> -th parameter is wrong (not counting handle).

The generic API has two different types, `dataTypeA` is data type of the matrix **A** and array `tau` and `computeType` is compute type of the operation. `cusolverDnGexrf` only supports the following four combinations.

#### Valid combination of data type and compute type

<b>DataTypeA</b>	<b>ComputeType</b>	<b>Meaning</b>
CUDA_R_32F	CUDA_R_32F	SGEQRF
CUDA_R_64F	CUDA_R_64F	DGEQRF
CUDA_C_32F	CUDA_C_32F	CGEQRF
CUDA_C_64F	CUDA_C_64F	ZGEQRF

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $m, n < 0$ or $lda < \max(1, m)$ ).
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

#### 2.4.2.14. cusolverDn<t1><t2>gels()

These functions compute the solution of a system of linear equations with one or multiple right hand sides using mixed precision iterative refinement techniques based on the QR factorization `Xgels`. These functions are similar in term of functionalities to the full precision LAPACK QR (least squares) solver (`Xgels`, where *X* denotes *Z, C, D, S*) but it uses lower precision internally in order to provide faster time to solution, from here comes the name mixed precision. Mixed precision iterative refinement techniques means that the solver compute an QR factorization in lower precision and then iteratively refine

the solution to achieve the accuracy of the Inputs/Outputs datatype precision. The  $\langle t1 \rangle$  corresponds to the Inputs/Outputs datatype precision while  $\langle t2 \rangle$  represent the internal lower precision at which the factorization will be carried on.

$$A \times X = B$$

Where **A** is **m-by-n** matrix and **X** is **n-by-nrhs** and **B** is **m-by-nrhs** matrices.

Functions API are designed to be as close as possible to LAPACK API to be considered as a quick and easy drop-in replacement. Description of these functions is given below.  $\langle t1 \rangle \langle t2 \rangle$ gels() functions are designated by two floating point precisions. The  $\langle t1 \rangle$  corresponds to the main precision (e.g., Inputs/Outputs datatype precision) and the  $\langle t2 \rangle$  represent the internal lower precision at which the factorization will be carried on. cusolver $\langle t1 \rangle \langle t2 \rangle$ gels() first attempts to factorize the matrix in lower precision and use this factorization within an iterative refinement procedure to obtain a solution with same normwise backward error as the main precision  $\langle t1 \rangle$ . If the approach fails to converge, then the method fallback to the main precision factorization and solve (Xgels) such a way that there is always a good solution at the output of these functions. If  $\langle t2 \rangle$  is equal to  $\langle t1 \rangle$ , then it is not a mixed precision process but rather a full one precision factorisation, solve and refinement within the same main precision.

The iterative refinement process is stopped if

$$\mathbf{ITER} > \mathbf{ITERMAX}$$

or for all the RHS we have:

$$\mathbf{RNRM} < \mathbf{SQRT(N)} * \mathbf{XNRM} * \mathbf{ANRM} * \mathbf{EPS} * \mathbf{BWDMAX}$$

where

- ▶ ITER is the number of the current iteration in the iterative refinement process
- ▶ RNRM is the infinity-norm of the residual
- ▶ XNRM is the infinity-norm of the solution
- ▶ ANRM is the infinity-operator-norm of the matrix A
- ▶ EPS is the machine epsilon that matches LAPACK  $\langle t1 \rangle$ LAMCH('Epsilon')

The value ITERMAX and BWDMAX are fixed to 50 and 1.0 respectively.

The function returns value describes the results of the solving process. A CUSOLVER\_STATUS\_SUCCESS indicates that the function finished with success otherwise, it indicates if one of the API arguments is incorrect, or if the function did not finish with success. More details about the error will be in the **niters** and the **dinfo** API parameters. See their description below for more details. User should provide the required workspace allocated on device memory. The amount of bytes required can be queried by calling the respective function  $\langle t1 \rangle \langle t2 \rangle$ gels\_bufferSize().

We provide a large set of mixed precision functions that include half, bfloat and tensorfloat as a lower precision as well as same precision functions (e.g., main and lowest precision are equal  $\langle t2 \rangle$  is equal to  $\langle t1 \rangle$ ). The following table specifies which precisions will be used for which interface function:

Tensor Float (TF32), introduced with NVIDIA Ampere Architecture GPUs, is the most robust tensor core accelerated compute mode for the iterative refinement solver. It is able to solve the widest range of problems in HPC arising from different applications and provides up to 4X and 5X speedup for real and complex systems, respectively. On Volta and Turing architecture GPUs, half precision tensor core acceleration is recommended. In cases where the iterative refinement solver fails to converge to the desired accuracy (main precision, INOUT data precision), it is recommended to use main precision as internal lowest precision (i.e., `cusolverDn[DD,ZZ]gels` for the FP64 case).

### Supported combinations of floating point precisions for `cusolver<t1><t2>gels()` functions

Interface function	Main precision (matrix, rhs and solution datatype)	Lowest precision allowed to be used internally
<code>cusolverDnZZgels</code>	<code>cuDoubleComplex</code>	double complex
<code>cusolverDnZCgels</code>	<code>cuDoubleComplex</code>	single complex
<code>cusolverDnZKgels</code>	<code>cuDoubleComplex</code>	half complex
<code>cusolverDnZegels</code>	<code>cuDoubleComplex</code>	bfloat complex
<code>cusolverDnZYgels</code>	<code>cuDoubleComplex</code>	tensorfloat complex
<code>cusolverDnCCgels</code>	<code>cuComplex</code>	single complex
<code>cusolverDnCKgels</code>	<code>cuComplex</code>	half complex
<code>cusolverDnCEgels</code>	<code>cuComplex</code>	bfloat complex
<code>cusolverDnCYgels</code>	<code>cuComplex</code>	tensorfloat complex
<code>cusolverDnDDgels</code>	double	double
<code>cusolverDnDSgels</code>	double	single
<code>cusolverDnDHgels</code>	double	half
<code>cusolverDnDBGels</code>	double	bfloat
<code>cusolverDnDXgels</code>	double	tensorfloat
<code>cusolverDnSSgels</code>	float	single
<code>cusolverDnSHgels</code>	float	half
<code>cusolverDnSBgels</code>	float	bfloat
<code>cusolverDnSXgels</code>	float	tensorfloat



Parameters of `cusolverDn<T1><T2>gels_bufferSize()` functions

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	Handle to the <code>cusolverDN</code> library context.
<code>m</code>	host	input	Number of rows of the matrix <b>A</b> . Should be non-negative and $n \leq m$
<code>n</code>	host	input	Number of columns of the matrix <b>A</b> . Should be non-negative and $n \leq m$ .
<code>nrhs</code>	host	input	Number of right hand sides to solve. Should be non-negative.
<code>dA</code>	device	None	Matrix <b>A</b> with size $m$ -by- $n$ . Can be <code>NULL</code> .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> . $lda \geq m$ .
<code>dB</code>	device	None	Set of right hand sides <b>B</b> of size $m$ -by- $nrhs$ . Can be <code>NULL</code> .
<code>ldb</code>	host	input	leading dimension of two-dimensional array used to store matrix of right hand sides <b>B</b> . $ldb \geq m$ .
<code>dX</code>	device	None	Set of solution vectors <b>X</b> of size $n$ -by- $nrhs$ . Can be <code>NULL</code> .
<code>ldx</code>	host	input	leading dimension of two-dimensional array used to store matrix of solution vectors <b>X</b> . $ldx \geq n$ .
<code>dwork</code>	device	none	Pointer to device workspace. Not used and can be <code>NULL</code> .

<b>lwork_bytes</b>	<b>host</b>	<b>output</b>	Pointer to a variable where required size of temporary workspace in bytes will be stored. Can't be NULL.
--------------------	-------------	---------------	----------------------------------------------------------------------------------------------------------

```
cusolverStatus_t cusolverDnZZgels(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

```
cusolverStatus_t cusolverDnZCgels(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

```
cusolverStatus_t cusolverDnZKgels(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

```
cusolverStatus_t cusolverDnZEgels(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int nrhs,
    cuDoubleComplex * dA,
    int ldda,
    cuDoubleComplex * dB,
    int lddb,
    cuDoubleComplex * dX,
    int lddx,
    void * dWorkspace,
    size_t lwork_bytes,
    int * niter,
    int * dinfo);
```

Parameters of `cusolverDn<T1><T2>gels()` functions

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	Handle to the <code>cusolverDN</code> library context.
<code>m</code>	host	input	Number of rows of the matrix <b>A</b> . Should be non-negative and $n \leq m$
<code>n</code>	host	input	Number of columns of the matrix <b>A</b> . Should be non-negative and $n \leq m$ .
<code>nrhs</code>	host	input	Number of right hand sides to solve. Should be non-negative.
<code>dA</code>	device	in/out	Matrix <b>A</b> with size $m$ -by- $n$ . Can't be <code>NULL</code> . On return - unchanged if the lowest precision is not equal to the main precision and the iterative refinement solver converged, - garbage otherwise.
<code>ldda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> . $ldda \geq m$ .
<code>dB</code>	device	input	Set of right hand sides <b>B</b> of size $m$ -by- $nrhs$ . Can't be <code>NULL</code> .
<code>lddb</code>	host	input	leading dimension of two-dimensional array used to store matrix of right hand sides <b>B</b> . $lddb \geq m$ .
<code>dX</code>	device	output	Set of solution vectors <b>X</b> of size $n$ -by- $nrhs$ . Can't be <code>NULL</code> .
<code>lddx</code>	host	input	leading dimension of two-dimensional array used to store matrix of solution vectors <b>X</b> . $lddx \geq n$ .
<code>dWorkspace</code>	device	input	Pointer to an allocated workspace in device memory of size <code>lwork_bytes</code> .
<code>lwork_bytes</code>	host	input	Size of the allocated device workspace. Should be at least what was returned by <code>cusolverDn&lt;T1&gt;&lt;T2&gt;gels_bufferSize()</code> function
<code>niters</code>	host	output	If iter is <ul style="list-style-type: none"> <li>▶ <code>&lt;0</code> : iterative refinement has failed, main precision (Inputs/Outputs precision) factorization has been performed.</li> <li>▶ <code>-1</code> : taking into account machine parameters, <math>n</math>, <math>nrhs</math>, it is a priori not worth working in lower precision</li> <li>▶ <code>-2</code> : overflow of an entry when moving from main to lower precision</li> <li>▶ <code>-3</code> : failure during the factorization</li> <li>▶ <code>-5</code> : overflow occurred during computation</li> <li>▶ <code>-50</code>: solver stopped the iterative refinement after reaching maximum allowed iterations.</li> </ul>

			<ul style="list-style-type: none"> <li>▶ <code>&gt;0</code> : iter is a number of iterations solver performed to reach convergence criteria</li> </ul>
<code>dinfo</code>	<code>device</code>	<code>output</code>	Status of the IRS solver on the return. If 0 - solve was successful. If <code>dinfo = -i</code> then <code>i</code> -th argument is not valid.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed, for example: <ul style="list-style-type: none"> <li>▶ <code>n &lt; 0</code></li> <li>▶ <code>lda &lt; max(1, m)</code></li> <li>▶ <code>ldb &lt; max(1, m)</code></li> <li>▶ <code>ldx &lt; max(1, n)</code></li> </ul>
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the IRS solver supports compute capability 7.0 and above. The lowest precision options <code>CUSOLVER_[CR]_16BF</code> and <code>CUSOLVER_[CR]_TF32</code> are only available on compute capability 8.0 and above.
<code>CUSOLVER_STATUS_INVALID_WORKSPACE</code>	<code>lwork_bytes</code> is smaller than the required workspace.
<code>CUSOLVER_STATUS_IRS_OUT_OF_RANGE</code>	Numerical error related to <code>niters &lt; 0</code> , see <code>niters</code> description for more details.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal error occurred, check the <code>dinfo</code> and the <code>niters</code> arguments for more details.

### 2.4.2.15. `cusolverDnIRSXgels()`

This function is designed to perform same functionality as `cusolverDn<T1><T2>gels()` functions, but wrapped in a more generic and expert interface that gives user more control to parametrize the function as well as it provides more informations on output. `cusolverDnIRSXgels()` allows additional control of the solver parameters such as setting:

- ▶ the main precision (Inputs/Outputs precision) of the solver,
- ▶ the lowest precision to be used internally by the solver,
- ▶ the refinement solver type,
- ▶ the maximum allowed number of iterations in the refinement phase,
- ▶ the tolerance of the refinement solver,
- ▶ the fallback to main precision,
- ▶ and more other.

through the configuration parameters structure `gels_irs_params` and its helper functions. For more details about what configuration can be set and its meaning please refer to all the functions in the `cuSolverDN Helper Function Section` that start with `cusolverDnIRSParamsxxxx()`. Moreover, `cusolverDnIRSXgels()` provides additional

informations on the output such as the convergence history (e.g., the residual norms) at each iteration and the number of iterations needed to converge. For more details about what informations can be retrieved and its meaning please refer to all the functions in the cuSolverDN Helper Function Section that start with `cusolverDnIRSInfosxxxx()`

The function returns value describes the results of the solving process. A `CUSOLVER_STATUS_SUCCESS` indicates that the function finished with success otherwise, it indicates if one of the API arguments is incorrect, or if the configurations of params/infos structure is incorrect or if the function did not finish with success. More details about the error can be found by checking the `niters` and the `dinfo` API parameters. See their description below for further details. User should provide the required workspace allocated on device for the `cusolverDnIRSXgels()` function. The amount of bytes required for the function can be queried by calling the respective function `cusolverDnIRSXgels_bufferSize()`. Note that, if the user would like a particular configuration to be set via the params structure, it should be set before the call to `cusolverDnIRSXgels_bufferSize()` to get the size of the required workspace.

The following table provides all possible combinations values for the lowest precision corresponding to the Inputs/Outputs data type. Note that if the lowest precision matches the Inputs/Outputs datatype, then main precision factorization will be used

Tensor Float (TF32), introduced with NVIDIA Ampere Architecture GPUs, is the most robust tensor core accelerated compute mode for the iterative refinement solver. It is able to solve the widest range of problems in HPC arising from different applications and provides up to 4X and 5X speedup for real and complex systems, respectively. On Volta and Turing architecture GPUs, half precision tensor core acceleration is recommended. In cases where the iterative refinement solver fails to converge to the desired accuracy (main precision, INOUT data precision), it is recommended to use main precision as internal lowest precision.

#### Supported Inputs/Outputs data type and lower precision for the IRS solver

Inputs/Outputs Data Type (e.g., main precision)	Supported values for the lowest precision
<code>CUSOLVER_C_64F</code>	<code>CUSOLVER_C_64F</code> , <code>CUSOLVER_C_32F</code> , <code>CUSOLVER_C_16F</code> , <code>CUSOLVER_C_16BF</code> , <code>CUSOLVER_C_TF32</code>
<code>CUSOLVER_C_32F</code>	<code>CUSOLVER_C_32F</code> , <code>CUSOLVER_C_16F</code> , <code>CUSOLVER_C_16BF</code> , <code>CUSOLVER_C_TF32</code>
<code>CUSOLVER_R_64F</code>	<code>CUSOLVER_R_64F</code> , <code>CUSOLVER_R_32F</code> , <code>CUSOLVER_R_16F</code> , <code>CUSOLVER_R_16BF</code> , <code>CUSOLVER_R_TF32</code>
<code>CUSOLVER_R_32F</code>	<code>CUSOLVER_R_32F</code> , <code>CUSOLVER_R_16F</code> , <code>CUSOLVER_R_16BF</code> , <code>CUSOLVER_R_TF32</code>

The `cusolverDnIRSXgels_bufferSize()` function return the required workspace buffer size in bytes for the corresponding `cusolverDnXgels()` call with given `gels_irs_params` configuration.

```
cusolverStatus_t
cusolverDnIRSXgels_bufferSize(
    cusolverDnHandle_t      handle,
    cusolverDnIRSParams_t   gels_irs_params,
    cusolver_int_t          m,
    cusolver_int_t          n,
    cusolver_int_t          nrhs,
    size_t                  * lwork_bytes);
```

### Parameters of `cusolverDnIRSXgels_bufferSize()` functions

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	Handle to the <code>cusolverDn</code> library context.
<code>params</code>	host	input	Xgels configuration parameters
<code>m</code>	host	input	Number of rows of the matrix <b>A</b> . Should be non-negative and $n \leq m$
<code>n</code>	host	input	Number of columns of the matrix <b>A</b> . Should be non-negative and $n \leq m$ .
<code>nrhs</code>	host	input	Number of right hand sides to solve. Should be non-negative. Note that, <code>nrhs</code> is limited to 1 if the selected IRS refinement solver is <code>CUSOLVER_IRS_REFINE_GMRES</code> , <code>CUSOLVER_IRS_REFINE_GMRES_GMRES</code> , <code>CUSOLVER_IRS_REFINE_CLASSICAL_GMRES</code> .
<code>lwork_bytes</code>	host	out	Pointer to a variable, where the required size in bytes, of the workspace will be stored after a call to <code>cusolverDnIRSXgels_bufferSize</code> . Can't be NULL.

```
cusolverStatus_t cusolverDnIRSXgels(
    cusolverDnHandle_t      handle,
    cusolverDnIRSParams_t   gels_irs_params,
    cusolverDnIRSInfos_t    gels_irs_infos,
    int                     m,
    int                     n,
    int                     nrhs,
    void                    * dA,
    int                     * ldda,
    void                    * dB,
    int                     * lddb,
    void                    * dX,
    int                     * lddx,
    void                    * dWorkspace,
    size_t                  lwork_bytes,
    int                     * dinfo);
```

## Parameters of cusolverDnIRSXgels() functions

parameter	Memory	In/out	Meaning
handle	host	input	Handle to the cusolverDn library context.
gels_irs_params	host	input	Configuration parameters structure, can serve one or more calls to any IRS solver
gels_irs_infos	host	in/out	Info structure, where information about a particular solve will be stored. The gels_irs_infos structure correspond to a particular call. Thus different calls requires different gels_irs_infos structure otherwise, it will be overwritten.
m	host	input	Number of rows of the matrix <b>A</b> . Should be non-negative and $n \leq m$
n	host	input	Number of columns of the matrix <b>A</b> . Should be non-negative and $n \leq m$ .
nrhs	host	input	Number of right hand sides to solve. Should be non-negative. Note that, <b>nrhs</b> is limited to 1 if the selected IRS refinement solver is CUSOLVER_IRS_REFINE_GMRES, CUSOLVER_IRS_REFINE_GMRES_GMRES, CUSOLVER_IRS_REFINE_CLASSICAL_GMRES.
dA	device	in/out	Matrix <b>A</b> with size <b>m-by-n</b> . Can't be <b>NULL</b> . On return - unchanged if the lowest precision is not equal to the main precision and the iterative refinement solver converged, - garbage otherwise.
ldda	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> . $ldda \geq m$ .
dB	device	input	Set of right hand sides <b>B</b> of size <b>m-by-nrhs</b> . Can't be <b>NULL</b> .
lddb	host	input	leading dimension of two-dimensional array used to store matrix of right hand sides <b>B</b> . $lddb \geq m$ .
dX	device	output	Set of solution vectors <b>X</b> of size <b>n-by-nrhs</b> . Can't be <b>NULL</b> .
lddx	host	input	leading dimension of two-dimensional array used to store matrix of solution vectors <b>X</b> . $lddx \geq n$ .
dWorkspace	device	input	Pointer to an allocated workspace in device memory of size <b>lwork_bytes</b> .
lwork_bytes	host	input	Size of the allocated device workspace. Should be at least what was returned by <code>cusolverDnIRSXgels_bufferSize()</code> function
niters	host	output	If iter is <ul style="list-style-type: none"> <li>▶ <math>&lt; 0</math> : iterative refinement has failed, main precision (Inputs/Outputs)</li> </ul>

			<p>precision) factorization has been performed if fallback is enabled.</p> <ul style="list-style-type: none"> <li>▶ -1 : taking into account machine parameters, n, nrhs, it is a priori not worth working in lower precision</li> <li>▶ -2 : overflow of an entry when moving from main to lower precision</li> <li>▶ -3 : failure during the factorization</li> <li>▶ -5 : overflow occurred during computation</li> <li>▶ -maxiter: solver stopped the iterative refinement after reaching maximum allowed iterations.</li> <li>▶ &gt;0 : iter is a number of iterations solver performed to reach convergence criteria</li> </ul>
<code>dinfo</code>	<code>device</code>	<code>output</code>	Status of the IRS solver on the return. If 0 - solve was successful. If <code>dinfo = -i</code> then <code>i</code> -th argument is not valid.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	<p>invalid parameters were passed, for example:</p> <ul style="list-style-type: none"> <li>▶ <code>n &lt; 0</code></li> <li>▶ <code>lda &lt; max(1, m)</code></li> <li>▶ <code>ldb &lt; max(1, m)</code></li> <li>▶ <code>ldx &lt; max(1, n)</code></li> </ul>
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the IRS solver supports compute capability 7.0 and above. The lowest precision options <code>CUSOLVER_[CR]_16BF</code> and <code>CUSOLVER_[CR]_TF32</code> are only available on compute capability 8.0 and above.
<code>CUSOLVER_STATUS_INVALID_WORKSPACE</code>	<code>lwork_bytes</code> is smaller than the required workspace. Could happen if the users called <code>cusolverDnIRSxgels_bufferSize()</code> function, then changed some of the configurations setting such as the lowest precision.
<code>CUSOLVER_STATUS_IRS_OUT_OF_RANGE</code>	Numerical error related to <code>niters &lt; 0</code> , see <code>niters</code> description for more details.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal error occurred, check the <code>dinfo</code> and the <code>niters</code> arguments for more details.
<code>CUSOLVER_STATUS_IRS_PARAMS_NOT_INITIALIZE</code>	The configuration parameter <code>gels_irs_params</code> structure was not created.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID</code>	One of the configuration parameter in the <code>gels_irs_params</code> structure is not valid.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID_PREC</code>	The main and/or the lowest precision configuration parameter in the <code>gels_irs_params</code>

	structure is not valid, check the table above for the supported combinations.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID_MAXITER</code>	The maxiter configuration parameter in the <code>gels_irs_params</code> structure is not valid.
<code>CUSOLVER_STATUS_IRS_PARAMS_INVALID_REFINE</code>	The refinement solver configuration parameter in the <code>gels_irs_params</code> structure is not valid.
<code>CUSOLVER_STATUS_IRS_NOT_SUPPORTED</code>	One of the configuration parameter in the <code>gels_irs_params</code> structure is not supported. For example if <code>nrhs &gt; 1</code> , and refinement solver was set to <code>CUSOLVER_IRS_REFINE_GMRES</code>
<code>CUSOLVER_STATUS_IRS_INFOS_NOT_INITIALIZED</code>	The information structure <code>gels_irs_infos</code> was not created.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	CPU memory allocation failed, most likely during the allocation of the residual array that store the residual norms.

### 2.4.2.16. cusolverDn<t>ormqr()

These helper functions calculate the size of work buffers needed.

```

cusolverStatus_t
cusolverDnSormqr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const float *A,
    int lda,
    const float *tau,
    const float *C,
    int ldc,
    int *lwork);

cusolverStatus_t
cusolverDnDormqr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const double *A,
    int lda,
    const double *tau,
    const double *C,
    int ldc,
    int *lwork);

cusolverStatus_t
cusolverDnCunmqr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const cuComplex *A,
    int lda,
    const cuComplex *tau,
    const cuComplex *C,
    int ldc,
    int *lwork);

cusolverStatus_t
cusolverDnZunmqr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    const cuDoubleComplex *C,
    int ldc,
    int *lwork);

```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSormqr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const float *A,
    int lda,
    const float *tau,
    float *C,
    int ldc,
    float *work,
    int lwork,
    int *devInfo);
```

```
cusolverStatus_t
cusolverDnDormqr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const double *A,
    int lda,
    const double *tau,
    double *C,
    int ldc,
    double *work,
    int lwork,
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCumqr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const cuComplex *A,
    int lda,
    const cuComplex *tau,
    cuComplex *C,
    int ldc,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZunmqr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasOperation_t trans,
    int m,
    int n,
    int k,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    cuDoubleComplex *C,
    int ldc,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function overwrites  $m \times n$  matrix **C** by

$$C = \begin{cases} \text{op}(Q) * C & \text{if side} == \text{CUBLAS\_SIDE\_LEFT} \\ C * \text{op}(Q) & \text{if side} == \text{CUBLAS\_SIDE\_RIGHT} \end{cases}$$

The operation of **Q** is defined by

$$\text{op}(Q) = \begin{cases} Q & \text{if transa} == \text{CUBLAS\_OP\_N} \\ Q^T & \text{if transa} == \text{CUBLAS\_OP\_T} \\ Q^H & \text{if transa} == \text{CUBLAS\_OP\_C} \end{cases}$$

**Q** is a unitary matrix formed by a sequence of elementary reflection vectors from QR factorization (**geqrf**) of **A**.

**Q**=**H**(1) **H**(2) ... **H**(**k**)

**Q** is of order **m** if **side** = **CUBLAS\_SIDE\_LEFT** and of order **n** if **side** = **CUBLAS\_SIDE\_RIGHT**.

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is size of the working space, and it is returned by **geqrf\_bufferSize()** or **ormqr\_bufferSize()**.

If output parameter **devInfo = -i** (less than zero), the **i-th** parameter is wrong (not counting handle).

The user can combine **geqrf**, **ormqr** and **trsm** to complete a linear solver or a least-square solver. Please refer to appendix C.1.

### API of **ormqr**

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	Handle to the cuSolverDn library context.
<b>side</b>	host	input	Indicates if matrix <b>Q</b> is on the left or right of <b>C</b> .
<b>trans</b>	host	input	Operation <b>op</b> ( <b>Q</b> ) that is non- or (conj.) transpose.
<b>m</b>	host	input	Number of columns of matrix <b>C</b> .
<b>n</b>	host	input	Number of rows of matrix <b>C</b> .
<b>k</b>	host	input	Number of elementary reflections whose product defines the matrix <b>Q</b> .
<b>A</b>	device	in/out	<type> array of dimension <b>lda * k</b> with <b>lda</b> is not less than <b>max(1,m)</b> . The matrix <b>A</b> is from <b>geqrf</b> , so <b>i-th</b> column contains elementary reflection vector.
<b>lda</b>	host	input	Leading dimension of two-dimensional array used to store matrix <b>A</b> . if <b>side</b> is <b>CUBLAS_SIDE_LEFT</b> , <b>lda</b> $\geq$ <b>max(1,m)</b> ; if <b>side</b> is <b>CUBLAS_SIDE_RIGHT</b> , <b>lda</b> $\geq$ <b>max(1,n)</b> .
<b>tau</b>	device	output	<type> array of dimension at least <b>min(m,n)</b> . The vector <b>tau</b> is from <b>geqrf</b> , so <b>tau(i)</b> is the scalar of <b>i-th</b> elementary reflection vector.
<b>C</b>	device	in/out	<type> array of size <b>ldc * n</b> . On exit, <b>C</b> is overwritten by <b>op(Q) * C</b> .
<b>ldc</b>	host	input	Leading dimension of two-dimensional array of matrix <b>C</b> . <b>ldc</b> $\geq$ <b>max(1,m)</b> .
<b>work</b>	device	in/out	Working space, <type> array of size <b>lwork</b> .
<b>lwork</b>	host	input	Size of working array <b>work</b> .
<b>devInfo</b>	device	output	If <b>devInfo = 0</b> , the <b>ormqr</b> is successful. If <b>devInfo = -i</b> , the <b>i-th</b> parameter is wrong (not counting handle).

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	The operation completed successfully.
--------------------------------	---------------------------------------

<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	The library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	Invalid parameters were passed ( <code>m, n &lt; 0</code> or wrong <code>lda</code> or <code>ldc</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	The device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	An internal operation failed.

### 2.4.2.17. `cusolverDn<t>orgqr()`

These helper functions calculate the size of work buffers needed.

```
cusolverStatus_t
cusolverDnSorgqr_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    const float *A,
    int lda,
    const float *tau,
    int *lwork);

cusolverStatus_t
cusolverDnDorgqr_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    const double *A,
    int lda,
    const double *tau,
    int *lwork);

cusolverStatus_t
cusolverDnCungqr_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    const cuComplex *A,
    int lda,
    const cuComplex *tau,
    int *lwork);

cusolverStatus_t
cusolverDnZungqr_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSorgqr(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    float *A,
    int lda,
    const float *tau,
    float *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnDorgqr(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    double *A,
    int lda,
    const double *tau,
    double *work,
    int lwork,
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCungqr(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    cuComplex *A,
    int lda,
    const cuComplex *tau,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZungqr(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int k,
    cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function overwrites  $m \times n$  matrix **A** by

$$Q = H(1) * H(2) * \dots * H(k)$$

where  $Q$  is a unitary matrix formed by a sequence of elementary reflection vectors stored in  $A$ .

The user has to provide working space which is pointed by input parameter `work`. The input parameter `lwork` is size of the working space, and it is returned by `orgqr_bufferSize()`.

If output parameter `devInfo = -i` (less than zero), the  $i$ -th parameter is wrong (not counting handle).

The user can combine `geqrf`, `orgqr` to complete orthogonalization. Please refer to appendix C.2.

### API of `ormqr`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>m</code>	host	input	number of rows of matrix $Q$ . $m \geq 0$ ;
<code>n</code>	host	input	number of columns of matrix $Q$ . $m \geq n \geq 0$ ;
<code>k</code>	host	input	number of elementary reflections whose product defines the matrix $Q$ . $n \geq k \geq 0$ ;
<code>A</code>	device	in/out	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, m)</code> . $i$ -th column of $A$ contains elementary reflection vector.
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix $A$ . <code>lda</code> $\geq$ <code>max(1, m)</code> .
<code>tau</code>	device	output	<type> array of dimension <code>k</code> . <code>tau(i)</code> is the scalar of $i$ -th elementary reflection vector.
<code>work</code>	device	in/out	working space, <type> array of size <code>lwork</code> .
<code>lwork</code>	host	input	size of working array <code>work</code> .
<code>devInfo</code>	device	output	if <code>info = 0</code> , the <code>orgqr</code> is successful. if <code>info = -i</code> , the $i$ -th parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $m, n, k < 0, n > m, k > n$ or <code>lda &lt; m</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.

<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.
---------------------------------------------	-------------------------------

### 2.4.2.18. `cusolverDn<t>sytrf()`

These helper functions calculate the size of the needed buffers.

```
cusolverStatus_t
cusolverDnSsytrf_bufferSize(cusolverDnHandle_t handle,
                            int n,
                            float *A,
                            int lda,
                            int *Lwork );

cusolverStatus_t
cusolverDnDsytrf_bufferSize(cusolverDnHandle_t handle,
                            int n,
                            double *A,
                            int lda,
                            int *Lwork );

cusolverStatus_t
cusolverDnCsytrf_bufferSize(cusolverDnHandle_t handle,
                            int n,
                            cuComplex *A,
                            int lda,
                            int *Lwork );

cusolverStatus_t
cusolverDnZsytrf_bufferSize(cusolverDnHandle_t handle,
                            int n,
                            cuDoubleComplex *A,
                            int lda,
                            int *Lwork );
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSsytrf(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 float *A,
                 int lda,
                 int *ipiv,
                 float *work,
                 int lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnDsytrf(cusolverDnHandle_t handle,
                 cublasFillMode_t uplo,
                 int n,
                 double *A,
                 int lda,
                 int *ipiv,
                 double *work,
                 int lwork,
                 int *devInfo );
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCsyrtrf(cusolverDnHandle_t handle,
                  cublasFillMode_t uplo,
                  int n,
                  cuComplex *A,
                  int lda,
                  int *ipiv,
                  cuComplex *work,
                  int lwork,
                  int *devInfo );

cusolverStatus_t
cusolverDnZsyrtrf(cusolverDnHandle_t handle,
                  cublasFillMode_t uplo,
                  int n,
                  cuDoubleComplex *A,
                  int lda,
                  int *ipiv,
                  cuDoubleComplex *work,
                  int lwork,
                  int *devInfo );
```

This function computes the Bunch-Kaufman factorization of a  $n \times n$  symmetric indefinite matrix

**A** is a  $n \times n$  symmetric matrix, only lower or upper part is meaningful. The input parameter **uplo** which part of the matrix is used. The function would leave other part untouched.

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, only lower triangular part of **A** is processed, and replaced by lower triangular factor **L** and block diagonal matrix **D**. Each block of **D** is either 1x1 or 2x2 block, depending on pivoting.

$$P * A * P^T = L * D * L^T$$

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, only upper triangular part of **A** is processed, and replaced by upper triangular factor **U** and block diagonal matrix **D**.

$$P * A * P^T = U * D * U^T$$

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is size of the working space, and it is returned by **sytrf\_bufferSize()**.

If Bunch-Kaufman factorization failed, i.e. **A** is singular. The output parameter **devInfo = i** would indicate **D(i, i) = 0**.

If output parameter **devInfo = -i** (less than zero), the **i-th** parameter is wrong (not counting handle).

The output parameter **devI piv** contains pivoting sequence. If **devI piv(i) = k > 0**, **D(i, i)** is 1x1 block, and **i-th** row/column of **A** is interchanged with **k-th** row/column of **A**. If **uplo** is **CUBLAS\_FILL\_MODE\_UPPER** and **devI piv(i-1) = devI piv(i) = -m < 0**, **D(i-1:i, i-1:i)** is a 2x2 block, and **(i-1)-th** row/column is interchanged

with  $m$ -th row/column. If `uplo` is `CUBLAS_FILL_MODE_LOWER` and `devI piv(i+1) = devI piv(i) = -m < 0`,  $D(i:i+1, i:i+1)$  is a 2x2 block, and  $(i+1)$ -th row/column is interchanged with  $m$ -th row/column.

### API of `sytrf`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the <code>cuSolverDN</code> library context.
<code>uplo</code>	host	input	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced.
<code>n</code>	host	input	number of rows and columns of matrix <b>A</b> .
<b>A</b>	device	in/out	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<code>ipiv</code>	device	output	array of size at least <code>n</code> , containing pivot indices.
<code>work</code>	device	in/out	working space, <type> array of size <code>lwork</code> .
<code>lwork</code>	host	input	size of working space <code>work</code> .
<code>devInfo</code>	device	output	if <code>devInfo = 0</code> , the LU factorization is successful. if <code>devInfo = -i</code> , the $i$ -th parameter is wrong (not counting <code>handle</code> ). if <code>devInfo = i</code> , the $D(i, i) = 0$ .

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n &lt; 0</code> or <code>lda &lt; max(1, n)</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.2.19. cusolverDn<t>potrfBatched()

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSpotrfBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    float *Aarray[],
    int lda,
    int *infoArray,
    int batchSize);

cusolverStatus_t
cusolverDnDpotrfBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    double *Aarray[],
    int lda,
    int *infoArray,
    int batchSize);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCpotrfBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuComplex *Aarray[],
    int lda,
    int *infoArray,
    int batchSize);

cusolverStatus_t
cusolverDnZpotrfBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *Aarray[],
    int lda,
    int *infoArray,
    int batchSize);
```

This function computes the Cholesky factorization of a sequence of Hermitian positive-definite matrices.

Each **Aarray[i]** for **i=0,1,..., batchSize-1** is a **n×n** Hermitian matrix, only lower or upper part is meaningful. The input parameter **uplo** indicates which part of the matrix is used.

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, only lower triangular part of **A** is processed, and replaced by lower triangular Cholesky factor **L**.

$$A = L * L^H$$

If input parameter `uplo` is `CUBLAS_FILL_MODE_UPPER`, only upper triangular part of **A** is processed, and replaced by upper triangular Cholesky factor **U**.

$$A = U^H * U$$

If Cholesky factorization failed, i.e. some leading minor of **A** is not positive definite, or equivalently some diagonal elements of **L** or **U** is not a real number. The output parameter `infoArray` would indicate smallest leading minor of **A** which is not positive definite.

`infoArray` is an integer array of size `batchsize`. If `potrfBatched` returns `CUSOLVER_STATUS_INVALID_VALUE`, `infoArray[0] = -i` (less than zero), meaning that the *i*-th parameter is wrong (not counting handle). If `potrfBatched` returns `CUSOLVER_STATUS_SUCCESS` but `infoArray[i] = k` is positive, then *i*-th matrix is not positive definite and the Cholesky factorization failed at row **k**.

Remark: the other part of **A** is used as a workspace. For example, if `uplo` is `CUBLAS_FILL_MODE_UPPER`, upper triangle of **A** contains cholesky factor **U** and lower triangle of **A** is destroyed after `potrfBatched`.

### API of `potrfBatched`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>uplo</code>	host	input	indicates if lower or upper part is stored, the other part is used as a workspace.
<code>n</code>	host	input	number of rows and columns of matrix <b>A</b> .
<code>Aarray</code>	device	in/out	array of pointers to <type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store each matrix <code>Aarray[i]</code> .
<code>infoArray</code>	device	output	array of size <code>batchSize</code> . <code>infoArray[i]</code> contains information of factorization of <code>Aarray[i]</code> . if <code>potrfBatched</code> returns <code>CUSOLVER_STATUS_INVALID_VALUE</code> , <code>infoArray[0] = -i</code> (less than zero) means the <i>i</i> -th parameter is wrong (not counting handle). if <code>potrfBatched</code> returns <code>CUSOLVER_STATUS_SUCCESS</code> , <code>infoArray[i] = 0</code> means the Cholesky factorization of <i>i</i> -th matrix is successful, and <code>infoArray[i] = k</code> means the leading submatrix of order <i>k</i> of <i>i</i> -th matrix is not positive definite.
<code>batchSize</code>	host	input	number of pointers in <code>Aarray</code> .

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.

<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n&lt;0</code> or <code>lda&lt;max(1,n)</code> or <code>batchSize&lt;1</code> ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.2.20. `cusolverDn<t>potrsBatched()`

```
cusolverStatus_t
cusolverDnSpotrsBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    int nrhs,
    float *Aarray[],
    int lda,
    float *Barray[],
    int ldb,
    int *info,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnDpotrsBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    int nrhs,
    double *Aarray[],
    int lda,
    double *Barray[],
    int ldb,
    int *info,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnCpotrsBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    int nrhs,
    cuComplex *Aarray[],
    int lda,
    cuComplex *Barray[],
    int ldb,
    int *info,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnZpotrsBatched(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    int nrhs,
    cuDoubleComplex *Aarray[],
    int lda,
    cuDoubleComplex *Barray[],
    int ldb,
    int *info,
    int batchSize);
```

This function solves a sequence of linear systems

$$A[i]*X[i]=B[i]$$

where each **Aarray[i]** for  $i=0,1,\dots, \text{batchSize}-1$  is a  $n \times n$  Hermitian matrix, only lower or upper part is meaningful. The input parameter **uplo** indicates which part of the matrix is used.

The user has to call **potrfBatched** first to factorize matrix **Aarray[i]**. If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, **A** is lower triangular Cholesky factor **L** corresponding to  $A = L * L^H$ . If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, **A** is upper triangular Cholesky factor **U** corresponding to  $A = U^H * U$ .

The operation is in-place, i.e. matrix **X** overwrites matrix **B** with the same leading dimension **ldb**.

The output parameter **info** is a scalar. If **info** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle).

Remark 1: only **nrhs=1** is supported.

Remark 2: **infoArray** from **potrfBatched** indicates if the matrix is positive definite. **info** from **potrsBatched** only shows which input parameter is wrong (not counting handle).

Remark 3: the other part of **A** is used as a workspace. For example, if **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, upper triangle of **A** contains cholesky factor **U** and lower triangle of **A** is destroyed after **potrsBatched**.

### API of potrsBatched

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolveDN library context.
uplo	host	input	indicates if matrix <b>A</b> lower or upper part is stored.
n	host	input	number of rows and columns of matrix <b>A</b> .
nrhs	host	input	number of columns of matrix <b>x</b> and <b>B</b> .
Aarray	device	in/out	array of pointers to <type> array of dimension $lda * n$ with $lda$ is not less than $\max(1, n)$ . <b>Aarray[i]</b> is either lower cholesky factor <b>L</b> or upper Cholesky factor <b>U</b> .
lda	host	input	leading dimension of two-dimensional array used to store each matrix <b>Aarray[i]</b> .
Barray	device	in/out	array of pointers to <type> array of dimension $ldb * nrhs$ . $ldb$ is not less than $\max(1, n)$ . As an input, <b>Barray[i]</b> is right hand side matrix. As an output, <b>Barray[i]</b> is the solution matrix.

<b>ldb</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store each matrix <b>Aarray[i]</b> .
<b>info</b>	<b>device</b>	<b>output</b>	if <b>info</b> = 0, all parameters are correct. if <b>info</b> = -i, the i-th parameter is wrong (not counting handle).
<b>batchSize</b>	<b>host</b>	<b>input</b>	number of pointers in <b>Aarray</b> .

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( <b>n</b> <0, <b>nrhs</b> <0, <b>lda</b> < <b>max</b> (1, <b>n</b> ), <b>ldb</b> < <b>max</b> (1, <b>n</b> ) or <b>batchSize</b> <0).
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.

## 2.4.3. Dense Eigenvalue Solver Reference

This chapter describes eigenvalue solver API of cuSolverDN, including bidiagonalization and SVD.

### 2.4.3.1. cusolverDn<t>gebrd()

These helper functions calculate the size of work buffers needed.

```

cusolverStatus_t
cusolverDnSgebrd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );

cusolverStatus_t
cusolverDnDgebrd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );

cusolverStatus_t
cusolverDnCgebrd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );

cusolverStatus_t
cusolverDnZgebrd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );

```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSgebrd(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 float *A,
                 int lda,
                 float *D,
                 float *E,
                 float *TAUQ,
                 float *TAUP,
                 float *Work,
                 int Lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnDgebrd(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 double *A,
                 int lda,
                 double *D,
                 double *E,
                 double *TAUQ,
                 double *TAUP,
                 double *Work,
                 int Lwork,
                 int *devInfo );
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCgebrd(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 cuComplex *A,
                 int lda,
                 float *D,
                 float *E,
                 cuComplex *TAUQ,
                 cuComplex *TAUP,
                 cuComplex *Work,
                 int Lwork,
                 int *devInfo );

cusolverStatus_t
cusolverDnZgebrd(cusolverDnHandle_t handle,
                 int m,
                 int n,
                 cuDoubleComplex *A,
                 int lda,
                 double *D,
                 double *E,
                 cuDoubleComplex *TAUQ,
                 cuDoubleComplex *TAUP,
                 cuDoubleComplex *Work,
                 int Lwork,
                 int *devInfo );
```

This function reduces a general  $m \times n$  matrix  $A$  to a real upper or lower bidiagonal form  $B$  by an orthogonal transformation:  $Q^H * A * P = B$

If  $m \geq n$ ,  $B$  is upper bidiagonal; if  $m < n$ ,  $B$  is lower bidiagonal.

The matrix  $Q$  and  $P$  are overwritten into matrix  $A$  in the following sense:

if  $m \geq n$ , the diagonal and the first superdiagonal are overwritten with the upper bidiagonal matrix  $B$ ; the elements below the diagonal, with the array **TAUQ**, represent the orthogonal matrix  $Q$  as a product of elementary reflectors, and the elements above the first superdiagonal, with the array **TAUP**, represent the orthogonal matrix  $P$  as a product of elementary reflectors.

if  $m < n$ , the diagonal and the first subdiagonal are overwritten with the lower bidiagonal matrix  $B$ ; the elements below the first subdiagonal, with the array **TAUQ**, represent the orthogonal matrix  $Q$  as a product of elementary reflectors, and the elements above the diagonal, with the array **TAUP**, represent the orthogonal matrix  $P$  as a product of elementary reflectors.

The user has to provide working space which is pointed by input parameter **Work**.

The input parameter **Lwork** is size of the working space, and it is returned by **gebrd\_bufferSize()**.

If output parameter **devInfo** =  $-i$  (less than zero), the  $i$ -th parameter is wrong (not counting handle).

Remark: **gebrd** only supports  $m \geq n$ .

### API of gebrd

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
m	host	input	number of rows of matrix $A$ .
n	host	input	number of columns of matrix $A$ .
A	device	in/out	<type> array of dimension $lda * n$ with $lda$ is not less than $\max(1, n)$ .
lda	host	input	leading dimension of two-dimensional array used to store matrix $A$ .
D	device	output	real array of dimension $\min(m, n)$ . The diagonal elements of the bidiagonal matrix $B$ : $D(i) = A(i, i)$ .
E	device	output	real array of dimension $\min(m, n)$ . The off-diagonal elements of the bidiagonal matrix $B$ : if $m \geq n$ , $E(i) = A(i, i+1)$ for $i = 1, 2, \dots, n-1$ ; if $m < n$ , $E(i) = A(i+1, i)$ for $i = 1, 2, \dots, m-1$ .
TAUQ	device	output	<type> array of dimension $\min(m, n)$ . The scalar factors of the elementary reflectors which represent the orthogonal matrix $Q$ .
TAUP	device	output	<type> array of dimension $\min(m, n)$ . The scalar factors of the elementary reflectors which represent the orthogonal matrix $P$ .

<code>Work</code>	<code>device</code>	<code>in/out</code>	working space, <type> array of size <code>Lwork</code> .
<code>Lwork</code>	<code>host</code>	<code>input</code>	size of <code>Work</code> , returned by <code>gebrd_bufferSize</code> .
<code>devInfo</code>	<code>device</code>	<code>output</code>	if <code>devInfo</code> = 0, the operation is successful. if <code>devInfo</code> = -i, the i-th parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>m, n &lt; 0</code> , or <code>lda &lt; max(1, m)</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.2. cusolverDn<t>orgbr()

These helper functions calculate the size of work buffers needed.

```
cusolverStatus_t
cusolverDnSorgbr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    int m,
    int n,
    int k,
    const float *A,
    int lda,
    const float *tau,
    int *lwork);

cusolverStatus_t
cusolverDnDorgbr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    int m,
    int n,
    int k,
    const double *A,
    int lda,
    const double *tau,
    int *lwork);

cusolverStatus_t
cusolverDnCungbr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    int m,
    int n,
    int k,
    const cuComplex *A,
    int lda,
    const cuComplex *tau,
    int *lwork);

cusolverStatus_t
cusolverDnZungbr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    int m,
    int n,
    int k,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t  
cusolverDnSorgbr(  
    cusolverDnHandle_t handle,  
    cublasSideMode_t side,  
    int m,  
    int n,  
    int k,  
    float *A,  
    int lda,  
    const float *tau,  
    float *work,  
    int lwork,  
    int *devInfo);
```

```
cusolverStatus_t  
cusolverDnDorgbr(  
    cusolverDnHandle_t handle,  
    cublasSideMode_t side,  
    int m,  
    int n,  
    int k,  
    double *A,  
    int lda,  
    const double *tau,  
    double *work,  
    int lwork,  
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCungbr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    int m,
    int n,
    int k,
    cuComplex *A,
    int lda,
    const cuComplex *tau,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZungbr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    int m,
    int n,
    int k,
    cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function generates one of the unitary matrices  $Q$  or  $P^{*}H$  determined by `gebrd` when reducing a matrix  $A$  to bidiagonal form:  $Q^{H} * A * P = B$

$Q$  and  $P^{*}H$  are defined as products of elementary reflectors  $H(i)$  or  $G(i)$  respectively.

The user has to provide working space which is pointed by input parameter `work`. The input parameter `lwork` is size of the working space, and it is returned by `orgbr_bufferSize()`.

If output parameter `devInfo = -i` (less than zero), the `i-th` parameter is wrong (not counting handle).

### API of orgbr

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>side</code>	host	input	if <code>side = CUBLAS_SIDE_LEFT</code> , generate $Q$ . if <code>side = CUBLAS_SIDE_RIGHT</code> , generate $P^{*}T$ .
<code>m</code>	host	input	number of rows of matrix $Q$ or $P^{*}T$ .
<code>n</code>	host	input	if <code>side = CUBLAS_SIDE_LEFT</code> , $m \geq n \geq \min(m,k)$ . if <code>side = CUBLAS_SIDE_RIGHT</code> , $n \geq m \geq \min(n,k)$ .
<code>k</code>	host	input	if <code>side = CUBLAS_SIDE_LEFT</code> , the number of columns in the original $m$ -by- $k$ matrix reduced by <code>gebrd</code> . if <code>side</code>

			= CUBLAS_SIDE_RIGHT, the number of rows in the original k-by-n matrix reduced by <code>gebrd</code> .
<b>A</b>	device	in/out	<type> array of dimension <code>lda * n</code> On entry, the vectors which define the elementary reflectors, as returned by <code>gebrd</code> . On exit, the m-by-n matrix <code>Q</code> or <code>P**T</code> .
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix <code>A</code> . <code>lda</code> $\geq$ <code>max(1,m)</code> ;
<b>tau</b>	device	output	<type> array of dimension <code>min(m,k)</code> if <code>side</code> is <code>CUBLAS_SIDE_LEFT</code> ; of dimension <code>min(n,k)</code> if <code>side</code> is <code>CUBLAS_SIDE_RIGHT</code> ; <code>tau(i)</code> must contain the scalar factor of the elementary reflector <code>H(i)</code> or <code>G(i)</code> , which determines <code>Q</code> or <code>P**T</code> , as returned by <code>gebrd</code> in its array argument <code>TAUQ</code> or <code>TAUP</code> .
<b>work</b>	device	in/out	working space, <type> array of size <code>lwork</code> .
<b>lwork</b>	host	input	size of working array <code>work</code> .
<b>devInfo</b>	device	output	if <code>info = 0</code> , the <code>ormqr</code> is successful. if <code>info = -i</code> , the <code>i</code> -th parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>m,n</code> <0 or wrong <code>lda</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.3. cusolverDn<t>sytrd()

These helper functions calculate the size of work buffers needed.

```
cusolverStatus_t
cusolverDnSsytrd_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *d,
    const float *e,
    const float *tau,
    int *lwork);

cusolverStatus_t
cusolverDnDsytrd_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *d,
    const double *e,
    const double *tau,
    int *lwork);

cusolverStatus_t
cusolverDnChetrd_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const float *d,
    const float *e,
    const cuComplex *tau,
    int *lwork);

cusolverStatus_t
cusolverDnZhetrd_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const double *d,
    const double *e,
    const cuDoubleComplex *tau,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSsytrd(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    float *d,
    float *e,
    float *tau,
    float *work,
    int lwork,
    int *devInfo);
```

```
cusolverStatus_t
cusolverDnDsytrd(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    double *d,
    double *e,
    double *tau,
    double *work,
    int lwork,
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnChetrd(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    float *d,
    float *e,
    cuComplex *tau,
    cuComplex *work,
    int lwork,
    int *devInfo);
```

```
cusolverStatus_t CUDENSEAPI cusolverDnZhetrd(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    double *d,
    double *e,
    cuDoubleComplex *tau,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function reduces a general symmetric (Hermitian)  $n \times n$  matrix  $\mathbf{A}$  to real symmetric tridiagonal form  $\mathbf{T}$  by an orthogonal transformation:  $Q^H * A * Q = T$

As an output,  $\mathbf{A}$  contains  $\mathbf{T}$  and householder reflection vectors. If `uplo = CUBLAS_FILL_MODE_UPPER`, the diagonal and first superdiagonal of  $\mathbf{A}$  are overwritten by the corresponding elements of the tridiagonal matrix  $\mathbf{T}$ , and the elements above the first superdiagonal, with the array `tau`, represent the orthogonal matrix  $\mathbf{Q}$  as a product of elementary reflectors; If `uplo = CUBLAS_FILL_MODE_LOWER`, the diagonal and first subdiagonal of  $\mathbf{A}$  are overwritten by the corresponding elements of the tridiagonal matrix  $\mathbf{T}$ , and the elements below the first subdiagonal, with the array `tau`, represent the orthogonal matrix  $\mathbf{Q}$  as a product of elementary reflectors.

The user has to provide working space which is pointed by input parameter `work`. The input parameter `lwork` is size of the working space, and it is returned by `sytrd_bufferSize()`.

If output parameter `devInfo = -i` (less than zero), the `i-th` parameter is wrong (not counting handle).

### API of `sytrd`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>uplo</code>	host	input	specifies which part of $\mathbf{A}$ is stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of $\mathbf{A}$ is stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of $\mathbf{A}$ is stored.
<code>n</code>	host	input	number of rows (columns) of matrix $\mathbf{A}$ .
<code>A</code>	device	in/out	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <code>n</code> -by- <code>n</code> upper triangular part of $\mathbf{A}$ contains the upper triangular part of the matrix $\mathbf{A}$ , and the strictly lower triangular part of $\mathbf{A}$ is not referenced. If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <code>n</code> -by- <code>n</code> lower triangular part of $\mathbf{A}$ contains the lower triangular part of the matrix $\mathbf{A}$ , and the strictly upper triangular part of $\mathbf{A}$ is not referenced. On exit, $\mathbf{A}$ is overwritten by $\mathbf{T}$ and householder reflection vectors.
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix $\mathbf{A}$ . <code>lda</code> $\geq$ <code>max(1, n)</code> .
<code>D</code>	device	output	real array of dimension <code>n</code> . The diagonal elements of the tridiagonal matrix $\mathbf{T}$ : <code>D(i) = A(i, i)</code> .
<code>E</code>	device	output	real array of dimension <code>(n-1)</code> . The off-diagonal elements of the tridiagonal matrix $\mathbf{T}$ : if <code>uplo = CUBLAS_FILL_MODE_UPPER</code> ,

			$E(i) = A(i, i+1)$ . if <code>uplo = CUBLAS_FILL_MODE_LOWER</code> $E(i) = A(i+1, i)$ .
<code>tau</code>	<code>device</code>	<code>output</code>	<type> array of dimension $(n-1)$ . The scalar factors of the elementary reflectors which represent the orthogonal matrix $Q$ .
<code>work</code>	<code>device</code>	<code>in/out</code>	working space, <type> array of size <code>lwork</code> .
<code>lwork</code>	<code>host</code>	<code>input</code>	size of <code>work</code> , returned by <code>sytrd_bufferSize</code> .
<code>devInfo</code>	<code>device</code>	<code>output</code>	if <code>devInfo = 0</code> , the operation is successful. if <code>devInfo = -i</code> , the $i$ -th parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $n < 0$ , or $lda < \max(1, n)$ , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.4. cusolverDn<t>ormtr()

These helper functions calculate the size of work buffers needed.

```
cusolverStatus_t
cusolverDnSormtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasFillMode_t uplo,
    cublasOperation_t trans,
    int m,
    int n,
    const float *A,
    int lda,
    const float *tau,
    const float *C,
    int ldc,
    int *lwork);

cusolverStatus_t
cusolverDnDormtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasFillMode_t uplo,
    cublasOperation_t trans,
    int m,
    int n,
    const double *A,
    int lda,
    const double *tau,
    const double *C,
    int ldc,
    int *lwork);

cusolverStatus_t
cusolverDnCunmtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasFillMode_t uplo,
    cublasOperation_t trans,
    int m,
    int n,
    const cuComplex *A,
    int lda,
    const cuComplex *tau,
    const cuComplex *C,
    int ldc,
    int *lwork);

cusolverStatus_t
cusolverDnZunmtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasFillMode_t uplo,
    cublasOperation_t trans,
    int m,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    const cuDoubleComplex *C,
    int ldc,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t  
cusolverDnSormtr(  
    cusolverDnHandle_t handle,  
    cublasSideMode_t side,  
    cublasFillMode_t uplo,  
    cublasOperation_t trans,  
    int m,  
    int n,  
    float *A,  
    int lda,  
    float *tau,  
    float *C,  
    int ldc,  
    float *work,  
    int lwork,  
    int *devInfo);
```

```
cusolverStatus_t  
cusolverDnDormtr(  
    cusolverDnHandle_t handle,  
    cublasSideMode_t side,  
    cublasFillMode_t uplo,  
    cublasOperation_t trans,  
    int m,  
    int n,  
    double *A,  
    int lda,  
    double *tau,  
    double *C,  
    int ldc,  
    double *work,  
    int lwork,  
    int *devInfo);
```

The **C** and **Z** data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCumtr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasFillMode_t uplo,
    cublasOperation_t trans,
    int m,
    int n,
    cuComplex *A,
    int lda,
    cuComplex *tau,
    cuComplex *C,
    int ldc,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZumtr(
    cusolverDnHandle_t handle,
    cublasSideMode_t side,
    cublasFillMode_t uplo,
    cublasOperation_t trans,
    int m,
    int n,
    cuDoubleComplex *A,
    int lda,
    cuDoubleComplex *tau,
    cuDoubleComplex *C,
    int ldc,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function overwrites  $m \times n$  matrix **C** by

$$C = \begin{cases} \text{op}(Q) * C & \text{if side == CUBLAS_SIDE_LEFT} \\ C * \text{op}(Q) & \text{if side == CUBLAS_SIDE_RIGHT} \end{cases}$$

where **Q** is a unitary matrix formed by a sequence of elementary reflection vectors from **sytrd**.

The operation on **Q** is defined by

$$\text{op}(Q) = \begin{cases} Q & \text{if transa == CUBLAS_OP_N} \\ Q^T & \text{if transa == CUBLAS_OP_T} \\ Q^H & \text{if transa == CUBLAS_OP_C} \end{cases}$$

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is size of the working space, and it is returned by **ormtr\_bufferSize()**.

If output parameter **devInfo** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle).

## API of ormtr

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverDN library context.
side	host	input	side = CUBLAS_SIDE_LEFT, apply Q or Q**T from the Left; side = CUBLAS_SIDE_RIGHT, apply Q or Q**T from the Right.
uplo	host	input	uplo = CUBLAS_FILL_MODE_LOWER: Lower triangle of A contains elementary reflectors from sytrd. uplo = CUBLAS_FILL_MODE_UPPER: Upper triangle of A contains elementary reflectors from sytrd.
trans	host	input	operation op(Q) that is non- or (conj.) transpose.
m	host	input	number of rows of matrix c.
n	host	input	number of columns of matrix c.
A	device	in/out	<type> array of dimension lda * m if side = CUBLAS_SIDE_LEFT; lda * n if side = CUBLAS_SIDE_RIGHT. The matrix A from sytrd contains the elementary reflectors.
lda	host	input	leading dimension of two-dimensional array used to store matrix A. if side is CUBLAS_SIDE_LEFT, lda >= max(1,m); if side is CUBLAS_SIDE_RIGHT, lda >= max(1,n).
tau	device	output	<type> array of dimension (m-1) if side is CUBLAS_SIDE_LEFT; of dimension (n-1) if side is CUBLAS_SIDE_RIGHT; The vector tau is from sytrd, so tau(i) is the scalar of i-th elementary reflection vector.
C	device	in/out	<type> array of size ldc * n. On exit, C is overwritten by op(Q) * C or C * op(Q).
ldc	host	input	leading dimension of two-dimensional array of matrix c. ldc >= max(1,m).
work	device	in/out	working space, <type> array of size lwork.
lwork	host	input	size of working array work.
devInfo	device	output	if devInfo = 0, the ormqr is successful. if devInfo = -i, the i-th parameter is wrong (not counting handle).

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>m, n &lt; 0</code> or wrong <code>lda</code> or <code>ldc</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.5. `cusolverDn<t>orgtr()`

These helper functions calculate the size of work buffers needed.

```
cusolverStatus_t
cusolverDnSorgtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *tau,
    int *lwork);

cusolverStatus_t
cusolverDnDorgtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *tau,
    int *lwork);

cusolverStatus_t
cusolverDnCungtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const cuComplex *tau,
    int *lwork);

cusolverStatus_t
cusolverDnZungtr_bufferSize(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSorgtr(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    const float *tau,
    float *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnDorgtr(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    const double *tau,
    double *work,
    int lwork,
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCungtr(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    const cuComplex *tau,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZungtr(
    cusolverDnHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *tau,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function generates a unitary matrix  $Q$  which is defined as the product of  $n-1$  elementary reflectors of order  $n$ , as returned by **sytrd**:

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is size of the working space, and it is returned by **orgtr\_bufferSize()**.

If output parameter `devInfo = -i` (less than zero), the `i-th` parameter is wrong (not counting handle).

### API of `orgtr`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the <code>cuSolverDN</code> library context.
<code>uplo</code>	host	input	<code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of <code>A</code> contains elementary reflectors from <code>sytrd</code> . <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of <code>A</code> contains elementary reflectors from <code>sytrd</code> .
<code>n</code>	host	input	number of rows (columns) of matrix <code>Q</code> .
<code>A</code>	device	in/out	<type> array of dimension <code>lda * n</code> On entry, matrix <code>A</code> from <code>sytrd</code> contains the elementary reflectors. On exit, matrix <code>A</code> contains the <code>n</code> -by- <code>n</code> orthogonal matrix <code>Q</code> .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <code>A</code> . <code>lda &gt;= max(1,n)</code> .
<code>tau</code>	device	output	<type> array of dimension <code>(n-1)</code> <code>tau(i)</code> is the scalar of <code>i</code> -th elementary reflection vector.
<code>work</code>	device	in/out	working space, <type> array of size <code>lwork</code> .
<code>lwork</code>	host	input	size of working array <code>work</code> .
<code>devInfo</code>	device	output	if <code>devInfo = 0</code> , the <code>orgtr</code> is successful. if <code>devInfo = -i</code> , the <code>i-th</code> parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n &lt; 0</code> or wrong <code>lda</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.6. cusolverDn<t>gesvd()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSgesvd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );

cusolverStatus_t
cusolverDnDgesvd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );

cusolverStatus_t
cusolverDnCgesvd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );

cusolverStatus_t
cusolverDnZgesvd_bufferSize(
    cusolverDnHandle_t handle,
    int m,
    int n,
    int *lwork );
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSgesvd (
    cusolverDnHandle_t handle,
    signed char jobu,
    signed char jobvt,
    int m,
    int n,
    float *A,
    int lda,
    float *S,
    float *U,
    int ldu,
    float *VT,
    int ldvt,
    float *work,
    int lwork,
    float *rwork,
    int *devInfo);

cusolverStatus_t
cusolverDnDgesvd (
    cusolverDnHandle_t handle,
    signed char jobu,
    signed char jobvt,
    int m,
    int n,
    double *A,
    int lda,
    double *S,
    double *U,
    int ldu,
    double *VT,
    int ldvt,
    double *work,
    int lwork,
    double *rwork,
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCgesvd (
    cusolverDnHandle_t handle,
    signed char jobu,
    signed char jobvt,
    int m,
    int n,
    cuComplex *A,
    int lda,
    float *S,
    cuComplex *U,
    int ldu,
    cuComplex *VT,
    int ldvt,
    cuComplex *work,
    int lwork,
    float *rwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZgesvd (
    cusolverDnHandle_t handle,
    signed char jobu,
    signed char jobvt,
    int m,
    int n,
    cuDoubleComplex *A,
    int lda,
    double *S,
    cuDoubleComplex *U,
    int ldu,
    cuDoubleComplex *VT,
    int ldvt,
    cuDoubleComplex *work,
    int lwork,
    double *rwork,
    int *devInfo);
```

This function computes the singular value decomposition (SVD) of a  $m \times n$  matrix  $\mathbf{A}$  and corresponding the left and/or right singular vectors. The SVD is written

$$\mathbf{A} = \mathbf{U} * \mathbf{\Sigma} * \mathbf{V}^H$$

where  $\mathbf{\Sigma}$  is an  $m \times n$  matrix which is zero except for its  $\min(m, n)$  diagonal elements,  $\mathbf{U}$  is an  $m \times m$  unitary matrix, and  $\mathbf{V}$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\mathbf{\Sigma}$  are the singular values of  $\mathbf{A}$ ; they are real and non-negative, and are returned in descending order. The first  $\min(m, n)$  columns of  $\mathbf{U}$  and  $\mathbf{V}$  are the left and right singular vectors of  $\mathbf{A}$ .

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is size of the working space, and it is returned by **gesvd\_bufferSize()**.

If output parameter **devInfo** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle). if **bdsqr** did not converge, **devInfo** specifies how many superdiagonals of an intermediate bidiagonal form did not converge to zero.

The **rwork** is real array of dimension  $(\min(m,n)-1)$ . If **devInfo**>0 and **rwork** is not nil, **rwork** contains the unconverged superdiagonal elements of an upper bidiagonal matrix. This is slightly different from LAPACK which puts unconverged superdiagonal elements in **work** if type is **real**; in **rwork** if type is **complex**. **rwork** can be a NULL pointer if the user does not want the information from superdiagonal.

Appendix G.1 provides a simple example of **gesvd**.

Remark 1: **gesvd** only supports  $m \geq n$ .

Remark 2: the routine returns  $V^H$ , not  $v$ .

### API of gesvd

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverDN library context.
<b>jobu</b>	host	input	specifies options for computing all or part of the matrix $u$ : = 'A': all $m$ columns of $U$ are returned in array $U$ ; = 'S': the first $\min(m,n)$ columns of $U$ (the left singular vectors) are returned in the array $U$ ; = 'O': the first $\min(m,n)$ columns of $U$ (the left singular vectors) are overwritten on the array $A$ ; = 'N': no columns of $U$ (no left singular vectors) are computed.
<b>jobvt</b>	host	input	specifies options for computing all or part of the matrix $V^{*T}$ : = 'A': all $N$ rows of $V^{*T}$ are returned in the array $VT$ ; = 'S': the first $\min(m,n)$ rows of $V^{*T}$ (the right singular vectors) are returned in the array $VT$ ; = 'O': the first $\min(m,n)$ rows of $V^{*T}$ (the right singular vectors) are overwritten on the array $A$ ; = 'N': no rows of $V^{*T}$ (no right singular vectors) are computed.
<b>m</b>	host	input	number of rows of matrix $A$ .
<b>n</b>	host	input	number of columns of matrix $A$ .
<b>A</b>	device	in/out	<type> array of dimension $lda * n$ with $lda$ is not less than $\max(1, m)$ . On exit, the contents of $A$ are destroyed.
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix $A$ .
<b>s</b>	device	output	real array of dimension $\min(m, n)$ . The singular values of $A$ , sorted so that $s(i) \geq s(i+1)$ .
<b>U</b>	device	output	<type> array of dimension $ldu * m$ with $ldu$ is not less than $\max(1, m)$ . $U$ contains the $m \times m$ unitary matrix $u$ .
<b>ldu</b>	host	input	leading dimension of two-dimensional array used to store matrix $u$ .

<b>VT</b>	<b>device</b>	<b>output</b>	<type> array of dimension $ldvt * n$ with $ldvt$ is not less than $\max(1, n)$ . <b>VT</b> contains the $n \times n$ unitary matrix $V^{**T}$ .
<b>ldvt</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>vt</b> .
<b>work</b>	<b>device</b>	<b>in/out</b>	working space, <type> array of size <b>lwork</b> .
<b>lwork</b>	<b>host</b>	<b>input</b>	size of <b>work</b> , returned by <b>gesvd_bufferSize</b> .
<b>rwork</b>	<b>device</b>	<b>input</b>	real array of dimension $\min(m,n)-1$ . It contains the unconverged superdiagonal elements of an upper bidiagonal matrix if <b>devInfo</b> > 0.
<b>devInfo</b>	<b>device</b>	<b>output</b>	if <b>devInfo</b> = 0, the operation is successful. if <b>devInfo</b> = -i, the i-th parameter is wrong (not counting handle). if <b>devInfo</b> > 0, <b>devInfo</b> indicates how many superdiagonals of an intermediate bidiagonal form did not converge to zero.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( $m, n < 0$ or $lda < \max(1, m)$ or $ldu < \max(1, m)$ or $ldvt < \max(1, n)$ ).
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.

### 2.4.3.7. cusolverDnGesvd()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t cusolverDnGesvd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    signed char jobu,
    signed char jobvt,
    int64_t m,
    int64_t n,
    cudaDataType dataTypeA,
    const void *A,
    int64_t lda,
    cudaDataType dataTypeS,
    const void *S,
    cudaDataType dataTypeU,
    const void *U,
    int64_t ldu,
    cudaDataType dataTypeVT,
    const void *VT,
    int64_t ldvt,
    cudaDataType computeType,
    size_t *workspaceInBytes);
```

The routine below

```
cusolverStatus_t CUSOLVERAPI cusolverDnGesvd(
    cusolverDnHandle_t handle,
    cusolverDnParams_t params,
    signed char jobu,
    signed char jobvt,
    int64_t m,
    int64_t n,
    cudaDataType dataTypeA,
    void *A,
    int64_t lda,
    cudaDataType dataTypeS,
    void *S,
    cudaDataType dataTypeU,
    void *U,
    int64_t ldu,
    cudaDataType dataTypeVT,
    void *VT,
    int64_t ldvt,
    cudaDataType computeType,
    void *pBuffer,
    size_t workspaceInBytes,
    int *info);
```

This function computes the singular value decomposition (SVD) of a  $m \times n$  matrix  $\mathbf{A}$  and corresponding the left and/or right singular vectors. The SVD is written

$$A = U \Sigma V^H$$

where  $\Sigma$  is an  $m \times n$  matrix which is zero except for its  $\min(m, n)$  diagonal elements,  $\mathbf{U}$  is an  $m \times m$  unitary matrix, and  $\mathbf{V}$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\Sigma$  are the singular values of  $\mathbf{A}$ ; they are real and non-negative, and are returned in descending order. The first  $\min(m, n)$  columns of  $\mathbf{U}$  and  $\mathbf{V}$  are the left and right singular vectors of  $\mathbf{A}$ .

The user has to provide working space which is pointed by input parameter `pBuffer`. The input parameter `workspaceInBytes` is size in bytes of the working space, and it is returned by `cusolverDnGesvd_bufferSize()`.

If output parameter `info = -i` (less than zero), the `i-th` parameter is wrong (not counting handle). if `bdsqr` did not converge, `info` specifies how many superdiagonals of an intermediate bidiagonal form did not converge to zero.

Currently, `cusolverDnGesvd` supports only the default algorithm.

#### Table of algorithms supported by `cusolverDnGesvd`

<code>CUSOLVER_ALG_0</code> or <code>NULL</code>	Default algorithm.
--------------------------------------------------	--------------------

Appendix G.5 provides a simple example of `cusolverDnGesvd`.

Remark 1: `gesvd` only supports  $m \geq n$ .

Remark 2: the routine returns  $V^H$ , not  $v$ .

List of input arguments for `cusolverDnGesvd_bufferSize` and `cusolverDnGesvd`:

#### API of `cusolverDnGesvd`

parameter	Memory	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>input</code>	handle to the <code>cuSolverDN</code> library context.
<code>params</code>	<code>host</code>	<code>input</code>	structure with information collected by <code>cusolverDnSetAdvOptions</code> .
<code>jobu</code>	<code>host</code>	<code>input</code>	specifies options for computing all or part of the matrix $u$ : = 'A': all $m$ columns of $U$ are returned in array $U$ ; = 'S': the first $\min(m,n)$ columns of $U$ (the left singular vectors) are returned in the array $U$ ; = 'O': the first $\min(m,n)$ columns of $U$ (the left singular vectors) are overwritten on the array $A$ ; = 'N': no columns of $U$ (no left singular vectors) are computed.
<code>jobvt</code>	<code>host</code>	<code>input</code>	specifies options for computing all or part of the matrix $V^{**T}$ : = 'A': all $N$ rows of $V^{**T}$ are returned in the array $VT$ ; = 'S': the first $\min(m,n)$ rows of $V^{**T}$ (the right singular vectors) are returned in the array $VT$ ; = 'O': the first $\min(m,n)$ rows of $V^{**T}$ (the right singular vectors) are overwritten on the array $A$ ; = 'N': no rows of $V^{**T}$ (no right singular vectors) are computed.
<code>m</code>	<code>host</code>	<code>input</code>	number of rows of matrix $A$ .
<code>n</code>	<code>host</code>	<code>input</code>	number of columns of matrix $A$ .
<code>dataTypeA</code>	<code>host</code>	<code>input</code>	data type of array $A$ .
<code>A</code>	<code>device</code>	<code>in/out</code>	array of dimension $lda * n$ with $lda$ is not less than $\max(1, m)$ . On exit, the contents of $A$ are destroyed.

<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<code>dataTypeS</code>	host	input	data type of array <b>s</b> .
<code>S</code>	device	output	real array of dimension $\min(m, n)$ . The singular values of <b>A</b> , sorted so that $s(i) \geq s(i+1)$ .
<code>dataTypeU</code>	host	input	data type of array <b>U</b> .
<code>U</code>	device	output	array of dimension <code>ldu * m</code> with <code>ldu</code> is not less than $\max(1, m)$ . <b>U</b> contains the $m \times m$ unitary matrix <b>U</b> .
<code>ldu</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>U</b> .
<code>dataTypeVT</code>	host	input	data type of array <b>VT</b> .
<code>VT</code>	device	output	array of dimension <code>ldvt * n</code> with <code>ldvt</code> is not less than $\max(1, n)$ . <b>VT</b> contains the $n \times n$ unitary matrix $V^*T$ .
<code>ldvt</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>VT</b> .
<code>computeType</code>	host	input	data type of computation.
<code>pBuffer</code>	device	in/out	Working space. Array of type <code>void</code> of size <code>workspaceInBytes</code> bytes.
<code>workspaceInBytes</code>	host	input	Size in bytes of <code>pBuffer</code> , returned by <code>cusolverDnGesvd_bufferSize</code> .
<code>info</code>	device	output	if <code>info = 0</code> , the operation is successful. if <code>info = -i</code> , the <i>i</i> -th parameter is wrong (not counting handle). if <code>info &gt; 0</code> , <code>info</code> indicates how many superdiagonals of an intermediate bidiagonal form did not converge to zero.

The generic API has three different types, `dataTypeA` is data type of the matrix **A**, `dataTypeS` is data type of the vector **S** and `dataTypeU` is data type of the matrix **U**, `dataTypeVT` is data type of the matrix **VT**, `computeType` is compute type of the operation. `cusolverDnGesvd` only supports the following four combinations.

#### Valid combination of data type and compute type

<code>DataTypeA</code>	<code>DataTypeS</code>	<code>DataTypeU</code>	<code>DataTypeVT</code>	<code>ComputeType</code>	Meaning
<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	SGESVD
<code>CUDA_R_64F</code>	<code>CUDA_R_64F</code>	<code>CUDA_R_64F</code>	<code>CUDA_R_64F</code>	<code>CUDA_R_64F</code>	DGESVD
<code>CUDA_C_32F</code>	<code>CUDA_C_32F</code>	<code>CUDA_C_32F</code>	<code>CUDA_C_32F</code>	<code>CUDA_C_32F</code>	CGESVD
<code>CUDA_C_64F</code>	<code>CUDA_C_64F</code>	<code>CUDA_C_64F</code>	<code>CUDA_C_64F</code>	<code>CUDA_C_64F</code>	ZGESVD

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.

<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $m, n < 0$ or $lda < \max(1, m)$ or $ldu < \max(1, m)$ or $ldvt < \max(1, n)$ ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.8. cusolverDn<t>gesvdj()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSgesvdj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    const float *A,
    int lda,
    const float *S,
    const float *U,
    int ldu,
    const float *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params);
```

```
cusolverStatus_t
cusolverDnDgesvdj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    const double *A,
    int lda,
    const double *S,
    const double *U,
    int ldu,
    const double *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params);
```

```
cusolverStatus_t
cusolverDnCgesvdj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    const cuComplex *A,
    int lda,
    const float *S,
    const cuComplex *U,
    int ldu,
    const cuComplex *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params);
```

```
cusolverStatus_t
cusolverDnZgesvdj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const double *S,
    const cuDoubleComplex *U,
    int ldu,
    const cuDoubleComplex *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSgesvdj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    float *A,
    int lda,
    float *S,
    float *U,
    int ldu,
    float *V,
    int ldv,
    float *work,
    int lwork,
    int *info,
    gesvdjInfo_t params);
```

```
cusolverStatus_t
cusolverDnDgesvdj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    double *A,
    int lda,
    double *S,
    double *U,
    int ldu,
    double *V,
    int ldv,
    double *work,
    int lwork,
    int *info,
    gesvdjInfo_t params);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCgesvdj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    cuComplex *A,
    int lda,
    float *S,
    cuComplex *U,
    int ldu,
    cuComplex *V,
    int ldv,
    cuComplex *work,
    int lwork,
    int *info,
    gesvdjInfo_t params);

cusolverStatus_t
cusolverDnZgesvdj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int econ,
    int m,
    int n,
    cuDoubleComplex *A,
    int lda,
    double *S,
    cuDoubleComplex *U,
    int ldu,
    cuDoubleComplex *V,
    int ldv,
    cuDoubleComplex *work,
    int lwork,
    int *info,
    gesvdjInfo_t params);
```

This function computes the singular value decomposition (SVD) of a  $m \times n$  matrix  $\mathbf{A}$  and corresponding the left and/or right singular vectors. The SVD is written

$$\mathbf{A} = \mathbf{U} * \mathbf{\Sigma} * \mathbf{V}^H$$

where  $\mathbf{\Sigma}$  is an  $m \times n$  matrix which is zero except for its  $\min(m, n)$  diagonal elements,  $\mathbf{U}$  is an  $m \times m$  unitary matrix, and  $\mathbf{V}$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\mathbf{\Sigma}$  are the singular values of  $\mathbf{A}$ ; they are real and non-negative, and are returned in descending order. The first  $\min(m, n)$  columns of  $\mathbf{U}$  and  $\mathbf{V}$  are the left and right singular vectors of  $\mathbf{A}$ .

**gesvdj** has the same functionality as **gesvd**. The difference is that **gesvd** uses QR algorithm and **gesvdj** uses Jacobi method. The parallelism of Jacobi method gives GPU better performance on small and medium size matrices. Moreover the user can configure **gesvdj** to perform approximation up to certain accuracy.

**gesvdj** iteratively generates a sequence of unitary matrices to transform matrix **A** to the following form

$$U^H * A * V = S + E$$

where **S** is diagonal and diagonal of **E** is zero.

During the iterations, the Frobenius norm of **E** decreases monotonically. As **E** goes down to zero, **S** is the set of singular values. In practice, Jacobi method stops if

$$\|E\|_F \leq \text{eps} * \|A\|_F$$

where **eps** is given tolerance.

**gesvdj** has two parameters to control the accuracy. First parameter is tolerance (**eps**). The default value is machine accuracy but The user can use function **cusolverDnXgesvdjSetTolerance** to set a priori tolerance. The second parameter is maximum number of sweeps which controls number of iterations of Jacobi method. The default value is 100 but the user can use function **cusolverDnXgesvdjSetMaxSweeps** to set a proper bound. The experimentis show 15 sweeps are good enough to converge to machine accuracy. **gesvdj** stops either tolerance is met or maximum number of sweeps is met.

Jacobi method has quadratic convergence, so the accuracy is not proportional to number of sweeps. To guarantee certain accuracy, the user should configure tolerance only.

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is the size of the working space, and it is returned by **gesvdj\_bufferSize()**.

If output parameter **info = -i** (less than zero), the **i-th** parameter is wrong (not counting handle). If **info = min(m,n)+1**, **gesvdj** does not converge under given tolerance and maximum sweeps.

If the user sets an improper tolerance, **gesvdj** may not converge. For example, tolerance should not be smaller than machine accuracy.

Appendix G.2 provides a simple example of **gesvdj**.

Remark 1: **gesvdj** supports any combination of **m** and **n**.

Remark 2: the routine returns **v**, not  $V^H$ . This is different from **gesvd**.

### API of **gesvdj**

parameter	Memory	In/out	Meaning
<b>handle</b>	<b>host</b>	<b>input</b>	handle to the cuSolverDN library context.
<b>jobz</b>	<b>host</b>	<b>input</b>	specifies options to either compute singular value only or singular vectors as well: <b>jobz = CUSOLVER_EIG_MODE_NOVECTOR</b> : Compute singular values only; <b>jobz = CUSOLVER_EIG_MODE_VECTOR</b> : Compute singular values and singular vectors.
<b>econ</b>	<b>host</b>	<b>input</b>	econ = 1 for economy size for <b>u</b> and <b>v</b> .

<b>m</b>	host	input	number of rows of matrix <b>A</b> .
<b>n</b>	host	input	number of columns of matrix <b>A</b> .
<b>A</b>	device	in/out	<type> array of dimension $lda * n$ with $lda$ is not less than $\max(1, m)$ . On exit, the contents of <b>A</b> are destroyed.
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>S</b>	device	output	real array of dimension $\min(m, n)$ . The singular values of <b>A</b> , sorted so that $s(i) \geq s(i+1)$ .
<b>U</b>	device	output	<type> array of dimension $ldu * m$ if <b>econ</b> is zero. If <b>econ</b> is nonzero, the dimension is $ldu * \min(m, n)$ . <b>U</b> contains the left singular vectors.
<b>ldu</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>U</b> . $ldu$ is not less than $\max(1, m)$ .
<b>V</b>	device	output	<type> array of dimension $ldv * n$ if <b>econ</b> is zero. If <b>econ</b> is nonzero, the dimension is $ldv * \min(m, n)$ . <b>V</b> contains the right singular vectors.
<b>ldv</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>V</b> . $ldv$ is not less than $\max(1, n)$ .
<b>work</b>	device	in/out	<type> array of size $lwork$ , working space.
<b>lwork</b>	host	input	size of <b>work</b> , returned by <code>gesvdj_bufferSize</code> .
<b>info</b>	device	output	if <b>info</b> = 0, the operation is successful. if <b>info</b> = -i, the i-th parameter is wrong (not counting handle). if <b>info</b> = $\min(m, n) + 1$ , <code>gesvdj</code> dose not converge under given tolerance and maximum sweeps.
<b>params</b>	host	in/out	structure filled with parameters of Jacobi algorithm and results of <code>gesvdj</code> .

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $m, n < 0$ or $lda < \max(1, m)$ or $ldu < \max(1, m)$ or $ldv < \max(1, n)$ or <b>jobz</b> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.9. cusolverDn<t>gesvdjBatched()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSgesvdjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int m,
    int n,
    const float *A,
    int lda,
    const float *S,
    const float *U,
    int ldu,
    const float *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnDgesvdjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int m,
    int n,
    const double *A,
    int lda,
    const double *S,
    const double *U,
    int ldu,
    const double *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnCgesvdjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int m,
    int n,
    const cuComplex *A,
    int lda,
    const float *S,
    const cuComplex *U,
    int ldu,
    const cuComplex *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnZgesvdjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int m,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const double *S,
    const cuDoubleComplex *U,
    int ldu,
    const cuDoubleComplex *V,
    int ldv,
    int *lwork,
    gesvdjInfo_t params,
    int batchSize);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t  
cusolverDnSgesvdjBatched(  
    cusolverDnHandle_t handle,  
    cusolverEigMode_t jobz,  
    int m,  
    int n,  
    float *A,  
    int lda,  
    float *S,  
    float *U,  
    int ldu,  
    float *V,  
    int ldv,  
    float *work,  
    int lwork,  
    int *info,  
    gesvdjInfo_t params,  
    int batchSize);
```

```
cusolverStatus_t  
cusolverDnDgesvdjBatched(  
    cusolverDnHandle_t handle,  
    cusolverEigMode_t jobz,  
    int m,  
    int n,  
    double *A,  
    int lda,  
    double *S,  
    double *U,  
    int ldu,  
    double *V,  
    int ldv,  
    double *work,  
    int lwork,  
    int *info,  
    gesvdjInfo_t params,  
    int batchSize);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCgesvdjBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int m,
    int n,
    cuComplex *A,
    int lda,
    float *S,
    cuComplex *U,
    int ldu,
    cuComplex *V,
    int ldv,
    cuComplex *work,
    int lwork,
    int *info,
    gesvdjInfo_t params,
    int batchSize);

cusolverStatus_t
cusolverDnZgesvdjBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int m,
    int n,
    cuDoubleComplex *A,
    int lda,
    double *S,
    cuDoubleComplex *U,
    int ldu,
    cuDoubleComplex *V,
    int ldv,
    cuDoubleComplex *work,
    int lwork,
    int *info,
    gesvdjInfo_t params,
    int batchSize);
```

This function computes singular values and singular vectors of a sequence of general  $m \times n$  matrices

$$A_j = U_j \Sigma_j V_j^H$$

where  $\Sigma_j$  is a real  $m \times n$  diagonal matrix which is zero except for its  $\min(m, n)$  diagonal elements.  $U_j$  (left singular vectors) is a  $m \times m$  unitary matrix and  $V_j$  (right singular vectors) is a  $n \times n$  unitary matrix. The diagonal elements of  $\Sigma_j$  are the singular values of  $A_j$  in either descending order or non-sorting order.

**gesvdjBatched** performs **gesvdj** on each matrix. It requires that all matrices are of the same size  $m, n$  no greater than 32 and are packed in contiguous way,

$$A = (A_0 \ A_1 \ \dots)$$

Each matrix is column-major with leading dimension **lda**, so the formula for random access is  $A_k(i,j) = A[i + lda*j + lda*n*k]$ .

The parameter **s** also contains singular values of each matrix in contiguous way,

$$S = (S_0 \ S_1 \ \dots)$$

The formula for random access of **s** is  $S_k(j) = S[j + \min(m,n)*k]$ .

Except for tolerance and maximum sweeps, **gesvdjBatched** can either sort the singular values in descending order (default) or chose as-is (without sorting) by the function **cusolverDnXgesvdjSetSortEig**. If the user packs several tiny matrices into diagonal blocks of one matrix, non-sorting option can separate singular values of those tiny matrices.

**gesvdjBatched** cannot report residual and executed sweeps by function **cusolverDnXgesvdjGetResidual** and **cusolverDnXgesvdjGetSweeps**. Any call of the above two returns **CUSOLVER\_STATUS\_NOT\_SUPPORTED**. The user needs to compute residual explicitly.

The user has to provide working space pointed by input parameter **work**. The input parameter **lwork** is the size of the working space, and it is returned by **gesvdjBatched\_bufferSize()**.

The output parameter **info** is an integer array of size **batchSize**. If the function returns **CUSOLVER\_STATUS\_INVALID\_VALUE**, the first element **info[0] = -i** (less than zero) indicates **i-th** parameter is wrong (not counting handle). Otherwise, if **info[i] = min(m, n) + 1**, **gesvdjBatched** does not converge on **i-th** matrix under given tolerance and maximum sweeps.

Appendix G.3 provides a simple example of **gesvdjBatched**.

### API of syevjBatched

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverDN library context.
<b>jobz</b>	host	input	specifies options to either compute singular value only or singular vectors as well: <b>jobz = CUSOLVER_EIG_MODE_NOVECTOR</b> : Compute singular values only; <b>jobz = CUSOLVER_EIG_MODE_VECTOR</b> : Compute singular values and singular vectors.
<b>m</b>	host	input	number of rows of matrix <b>A<sub>j</sub></b> . <b>m</b> is no greater than 32.
<b>n</b>	host	input	number of columns of matrix <b>A<sub>j</sub></b> . <b>n</b> is no greater than 32.
<b>A</b>	device	in/out	<type> array of dimension <b>lda * n * batchSize</b> with <b>lda</b> is not less than <b>max(1, n)</b> . on Exit: the contents of <b>A<sub>j</sub></b> are destroyed.
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>A<sub>j</sub></b> .

<b>S</b>	<b>device</b>	<b>output</b>	a real array of dimension $\min(m,n) * batchSize$ . It stores the singular values of $A_j$ in descending order or non-sorting order.
<b>U</b>	<b>device</b>	<b>output</b>	<type> array of dimension $ldu * m * batchSize$ . $U_j$ contains the left singular vectors of $A_j$ .
<b>ldu</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix $U_j$ . $ldu$ is not less than $\max(1,m)$ .
<b>V</b>	<b>device</b>	<b>output</b>	<type> array of dimension $ldv * n * batchSize$ . $V_j$ contains the right singular vectors of $A_j$ .
<b>ldv</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix $V_j$ . $ldv$ is not less than $\max(1,n)$ .
<b>work</b>	<b>device</b>	<b>in/out</b>	<type> array of size $lwork$ , working space.
<b>lwork</b>	<b>host</b>	<b>input</b>	size of <b>work</b> , returned by <code>gesvdjBatched_bufferSize</code> .
<b>info</b>	<b>device</b>	<b>output</b>	an integer array of dimension $batchSize$ . If <code>CUSOLVER_STATUS_INVALID_VALUE</code> is returned, $info[0] = -i$ (less than zero) indicates $i$ -th parameter is wrong (not counting handle). Otherwise, if $info[i] = 0$ , the operation is successful. if $info[i] = \min(m,n)+1$ , <code>gesvdjBatched</code> dose not converge on $i$ -th matrix under given tolerance and maximum sweeps.
<b>params</b>	<b>host</b>	<b>in/out</b>	structure filled with parameters of Jacobi algorithm.
<b>batchSize</b>	<b>host</b>	<b>input</b>	number of matrices.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $m, n < 0$ or $lda < \max(1,m)$ or $ldu < \max(1,m)$ or $ldv < \max(1,n)$ or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or $batchSize < 0$ ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.10. cusolverDn<t>gesvdaStridedBatched()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSgesvdaStridedBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
    const float *A,
    int lda,
    long long int strideA,
    const float *S,
    long long int strideS,
    const float *U,
    int ldu,
    long long int strideU,
    const float *V,
    int ldv,
    long long int strideV,
    int *lwork,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnDgesvdaStridedBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
    const double *A,
    int lda,
    long long int strideA,
    const double *S,
    long long int strideS,
    const double *U,
    int ldu,
    long long int strideU,
    const double *V,
    int ldv,
    long long int strideV,
    int *lwork,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnCgesvdaStridedBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
    const cuComplex *A,
    int lda,
    long long int strideA,
    const float *S,
    long long int strideS,
    const cuComplex *U,
    int ldu,
    long long int strideU,
    const cuComplex *V,
    int ldv,
    long long int strideV,
    int *lwork,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnZgesvdaStridedBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSgesvdaStridedBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
    const float *A,
    int lda,
    long long int strideA,
    float *S,
    long long int strideS,
    float *U,
    int ldu,
    long long int strideU,
    float *V,
    int ldv,
    long long int strideV,
    float *work,
    int lwork,
    int *info,
    double *h_R_nrmF,
    int batchSize);
```

```
cusolverStatus_t
cusolverDnDgesvdaStridedBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
    const double *A,
    int lda,
    long long int strideA,
    double *S,
    long long int strideS,
    double *U,
    int ldu,
    long long int strideU,
    double *V,
    int ldv,
    long long int strideV,
    double *work,
    int lwork,
    int *info,
    double *h_R_nrmF,
    int batchSize);
```

The C and Z data types are complex valued single and double precision, respectively.

```

cusolverStatus_t
cusolverDnCgesvdaStridedBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
    const cuComplex *A,
    int lda,
    long long int strideA,
    float *S,
    long long int strideS,
    cuComplex *U,
    int ldu,
    long long int strideU,
    cuComplex *V,
    int ldv,
    long long int strideV,
    cuComplex *work,
    int lwork,
    int *info,
    double *h_R_nrmF,
    int batchSize);

cusolverStatus_t
cusolverDnZgesvdaStridedBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    int rank,
    int m,
    int n,
    const cuDoubleComplex *A,
    int lda,
    long long int strideA,
    double *S,
    long long int strideS,
    cuDoubleComplex *U,
    int ldu,
    long long int strideU,
    cuDoubleComplex *V,
    int ldv,
    long long int strideV,
    cuDoubleComplex *work,
    int lwork,
    int *info,
    double *h_R_nrmF,
    int batchSize);

```

This function **gesvda** (**a** stands for approximate) approximates the singular value decomposition of a tall skinny  $m \times n$  matrix **A** and corresponding the left and right singular vectors. The economy form of SVD is written by

$$A = U * \Sigma * V^H$$

where  $\Sigma$  is an  $n \times n$  matrix.  $U$  is an  $m \times n$  unitary matrix, and  $V$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\Sigma$  are the singular values of  $A$ ; they are real and non-negative, and are returned in descending order.  $U$  and  $V$  are the left and right singular vectors of  $A$ .

**gesvda** computes eigenvalues of  $A^*T^*A$  to approximate singular values and singular vectors. It generates matrices  $U$  and  $V$  and transforms the matrix  $A$  to the following form

$$U^H * A * V = S + E$$

where  $S$  is diagonal and  $E$  depends on rounding errors. To certain conditions,  $U$ ,  $V$  and  $S$  approximate singular values and singular vectors up to machine zero of single precision. In general,  $V$  is unitary,  $S$  is more accurate than  $U$ . If singular value is far from zero, then left singular vector  $U$  is accurate. In other words, the accuracy of singular values and left singular vectors depend on the distance between singular value and zero.

The input parameter **rank** decides the number of singular values and singular vectors are computed in parameter  $S$ ,  $U$  and  $V$ .

The output parameter **h\_RnormF** computes Frobenius norm of residual.

$$A - U * S * V^H$$

if the parameter **rank** is equal  $n$ . Otherwise, **h\_RnormF** reports

$$\|U * S * V^H\| - \|S\|$$

in Frobenius norm sense. That is, how far  $U$  is from unitary.

**gesvdaStridedBatched** performs **gesvda** on each matrix. It requires that all matrices are of the same size  $m, n$  and are packed in contiguous way,

$$A = (A_0 \ A_1 \ \dots)$$

Each matrix is column-major with leading dimension **lda**, so the formula for random access is  $A_k(i,j) = A[i + lda*j + strideA*k]$ . Similarly, the formula for random access of  $S$  is  $S_k(j) = S[j + StrideS*k]$ , the formula for random access of  $U$  is  $U_k(i,j) = U[i + ldu*j + strideU*k]$  and the formula for random access of  $V$  is  $V_k(i,j) = V[i + ldv*j + strideV*k]$ .

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is the size of the working space, and it is returned by **gesvdaStridedBatched\_bufferSize()**.

The output parameter **info** is an integer array of size **batchSize**. If the function returns **CUSOLVER\_STATUS\_INVALID\_VALUE**, the first element **info[0] = -i** (less than zero) indicates **i-th** parameter is wrong (not counting handle). Otherwise, if **info[i] = min(m, n) + 1**, **gesvdaStridedBatched** does not converge on **i-th** matrix under given tolerance.

Appendix G.4 provides a simple example of **gesvda**.

Remark 1: the routine returns  $V$ , not  $V^H$ . This is different from **gesvd**.

Remark 2: if the user is confident on the accuracy of singular values and singular vectors, for example, certain conditions hold (required singular value is far from zero), then the performance can be improved by passing null pointer to `h_RnormF`, i.e. no computation of residual norm.

### API of gesvda

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>jobz</code>	host	input	specifies options to either compute singular value only or singular vectors as well: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute singular values only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute singular values and singular vectors.
<code>rank</code>	host	input	number of singular values (from largest to smallest).
<code>m</code>	host	input	number of rows of matrix $A_j$ .
<code>n</code>	host	input	number of columns of matrix $A_j$ .
<code>A</code>	device	input	<type> array of dimension <code>strideA * batchSize</code> with <code>lda</code> is not less than <code>max(1,m)</code> . $A_j$ is of dimension $m * n$ .
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix $A_j$ .
<code>strideA</code>	host	input	value of type long long int that gives the address offset between <code>A[i]</code> and <code>A[i+1]</code> . <code>strideA</code> is not less than <code>lda*n</code> .
<code>S</code>	device	output	a real array of dimension <code>strideS*batchSize</code> . It stores the singular values of $A_j$ in descending order. $S_j$ is of dimension <code>rank * 1</code> .
<code>strideS</code>	host	input	value of type long long int that gives the address offset between <code>S[i]</code> and <code>S[i+1]</code> . <code>strideS</code> is not less than <code>rank</code> .
<code>U</code>	device	output	<type> array of dimension <code>strideU * batchSize</code> . $U_j$ contains the left singular vectors of $A_j$ . $U_j$ is of dimension $m * rank$ .
<code>ldu</code>	host	input	leading dimension of two-dimensional array used to store matrix $U_j$ . <code>ldu</code> is not less than <code>max(1,m)</code> .
<code>strideU</code>	host	input	value of type long long int that gives the address offset between <code>U[i]</code> and <code>U[i+1]</code> . <code>strideU</code> is not less than <code>ldu*rank</code> .
<code>V</code>	device	output	<type> array of dimension <code>strideV * batchSize</code> . $V_j$ contains the right singular vectors of $A_j$ . $V_j$ is of dimension $n * rank$ .

<code>ldv</code>	<code>host</code>	<code>input</code>	leading dimension of two-dimensional array used to store matrix $v_j$ . <code>ldv</code> is not less than $\max(1, n)$ .
<code>strideV</code>	<code>host</code>	<code>input</code>	value of type long long int that gives the address offset between $v[i]$ and $v[i+1]$ . <code>strideV</code> is not less than $ldv * rank$ .
<code>work</code>	<code>device</code>	<code>in/out</code>	<type> array of size <code>lwork</code> , working space.
<code>lwork</code>	<code>host</code>	<code>input</code>	size of <code>work</code> , returned by <code>gesvdaStridedBatched_bufferSize</code> .
<code>info</code>	<code>device</code>	<code>output</code>	an integer array of dimension <code>batchSize</code> . If <code>CUSOLVER_STATUS_INVALID_VALUE</code> is returned, <code>info[0] = -i</code> (less than zero) indicates <i>i</i> -th parameter is wrong (not counting handle). Otherwise, if <code>info[i] = 0</code> , the operation is successful. if <code>info[i] = min(m, n) + 1</code> , <code>gesvdaStridedBatched</code> dose not converge on <i>i</i> -th matrix.
<code>h_RnormF</code>	<code>host</code>	<code>output</code>	<double> array of size <code>batchSize</code> . <code>h_RnormF[i]</code> is norm of residual of <i>i</i> -th matrix.
<code>batchSize</code>	<code>host</code>	<code>input</code>	number of matrices. <code>batchSize</code> is not less than 1.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $m, n < 0$ or $lda < \max(1, m)$ or $ldu < \max(1, m)$ or $ldv < \max(1, n)$ or $strideA < lda * n$ or $strideS < rank$ or $strideU < ldu * rank$ or $strideV < ldv * rank$ or $batchSize < 1$ or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.11. cusolverDn<t>syevd()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSsyevd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnDsyevd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnCheevd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnZheevd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const double *W,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSsyevd(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    float *W,
    float *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnDsyevd(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    double *W,
    double *work,
    int lwork,
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCcheevd(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    float *W,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZcheevd(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    double *W,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function computes eigenvalues and eigenvectors of a symmetric (Hermitian)  $n \times n$  matrix  $\mathbf{A}$ . The standard symmetric eigenvalue problem is

$$\mathbf{A}^* \mathbf{V} = \mathbf{V}^* \mathbf{A}$$

where  $\mathbf{\Lambda}$  is a real  $n \times n$  diagonal matrix.  $\mathbf{V}$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\mathbf{\Lambda}$  are the eigenvalues of  $\mathbf{A}$  in ascending order.

The user has to provide working space which is pointed by input parameter `work`. The input parameter `lwork` is size of the working space, and it is returned by `syevd_bufferSize()`.

If output parameter `devInfo = -i` (less than zero), the `i-th` parameter is wrong (not counting handle). If `devInfo = i` (greater than zero), `i` off-diagonal elements of an intermediate tridiagonal form did not converge to zero.

if `jobz = CUSOLVER_EIG_MODE_VECTOR`,  $\mathbf{A}$  contains the orthonormal eigenvectors of the matrix  $\mathbf{A}$ . The eigenvectors are computed by a divide and conquer algorithm.

Appendix F.1 provides a simple example of `syevd`.

### API of `syevd`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>jobz</code>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute eigenvalues only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute eigenvalues and eigenvectors.
<code>uplo</code>	host	input	specifies which part of $\mathbf{A}$ is stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of $\mathbf{A}$ is stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of $\mathbf{A}$ is stored.
<code>n</code>	host	input	number of rows (or columns) of matrix $\mathbf{A}$ .
<code>A</code>	device	in/out	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <code>n</code> -by- <code>n</code> upper triangular part of $\mathbf{A}$ contains the upper triangular part of the matrix $\mathbf{A}$ . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <code>n</code> -by- <code>n</code> lower triangular part of $\mathbf{A}$ contains the lower triangular part of the matrix $\mathbf{A}$ . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>devInfo = 0</code> , $\mathbf{A}$ contains the orthonormal eigenvectors of the matrix $\mathbf{A}$ . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of $\mathbf{A}$ are destroyed.
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix $\mathbf{A}$ .
<code>W</code>	device	output	a real array of dimension <code>n</code> . The eigenvalue values of $\mathbf{A}$ , in ascending order ie, sorted so that <code>w(i) &lt;= w(i+1)</code> .
<code>work</code>	device	in/out	working space, <type> array of size <code>lwork</code> .

<b>Lwork</b>	<b>host</b>	<b>input</b>	size of work, returned by <code>syevd_bufferSize</code> .
<b>devInfo</b>	<b>device</b>	<b>output</b>	if <code>devInfo = 0</code> , the operation is successful. if <code>devInfo = -i</code> , the <i>i</i> -th parameter is wrong (not counting handle). if <code>devInfo = i (&gt; 0)</code> , <code>devInfo</code> indicates <i>i</i> off-diagonal elements of an intermediate tridiagonal form did not converge to zero;

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( <code>n &lt; 0</code> , or <code>lda &lt; max(1, n)</code> , or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.

### 2.4.3.12. cusolverDnSyevd()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSyevd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverParams_t params,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cudaDataType dataTypeA,
    const void *A,
    int64_t lda,
    cudaDataType dataTypeW,
    const void *W,
    cudaDataType computeType,
    size_t *workspaceInBytes);
```

The routine bellow

```
cusolverStatus_t
cusolverDnSyevd(
  cusolverDnHandle_t handle,
  cusolverParams_t params,
  cusolverEigMode_t jobz,
  cublasFillMode_t uplo,
  int n,
  cudaDataType dataTypeA,
  const void *A,
  int64_t lda,
  cudaDataType dataTypeW,
  const void *W,
  cudaDataType computeType,
  void *pBuffer,
  size_t workspaceInBytes,
  int *info);
```

computes eigenvalues and eigenvectors of a symmetric (Hermitian)  $n \times n$  matrix  $\mathbf{A}$  using the generic API interface. The standard symmetric eigenvalue problem is

$$\mathbf{A} \mathbf{V} = \mathbf{V} \mathbf{\Lambda}$$

where  $\mathbf{\Lambda}$  is a real  $n \times n$  diagonal matrix.  $\mathbf{V}$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\mathbf{\Lambda}$  are the eigenvalues of  $\mathbf{A}$  in ascending order.

The user has to provide working space which is pointed by input parameter `pBuffer`. The input parameter `workspaceInBytes` is size in bytes of the working space, and it is returned by `cusolverDnSyevd_bufferSize()`.

If output parameter `info = -i` (less than zero), the `i`-th parameter is wrong (not counting handle). If `info = i` (greater than zero), `i` off-diagonal elements of an intermediate tridiagonal form did not converge to zero.

if `jobz = CUSOLVER_EIG_MODE_VECTOR`,  $\mathbf{A}$  contains the orthonormal eigenvectors of the matrix  $\mathbf{A}$ . The eigenvectors are computed by a divide and conquer algorithm.

Currently, `cusolverDnSyevd` supports only the default algorithm.

#### Table of algorithms supported by `cusolverDnSyevd`

CUSOLVER_ALG_0 or NULL	Default algorithm.
------------------------	--------------------

List of input arguments for `cusolverDnSyevd_bufferSize` and `cusolverDnSyevd`:

#### API of `cusolverDnSyevd`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>params</code>	host	input	structure with information collected by <code>cusolverDnSetAdvOptions</code> .
<code>jobz</code>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute eigenvalues only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute eigenvalues and eigenvectors.

<code>uplo</code>	host	input	specifies which part of <b>A</b> is stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of <b>A</b> is stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of <b>A</b> is stored.
<code>n</code>	host	input	number of rows (or columns) of matrix <b>A</b> .
<code>dataTypeA</code>	host	in	data type of array <b>A</b> .
<b>A</b>	device	in/out	array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <code>n</code> -by- <code>n</code> upper triangular part of <b>A</b> contains the upper triangular part of the matrix <b>A</b> . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <code>n</code> -by- <code>n</code> lower triangular part of <b>A</b> contains the lower triangular part of the matrix <b>A</b> . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>info = 0</code> , <b>A</b> contains the orthonormal eigenvectors of the matrix <b>A</b> . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of <b>A</b> are destroyed.
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<code>dataTypeW</code>	host	in	data type of array <b>w</b> .
<b>w</b>	device	output	a real array of dimension <code>n</code> . The eigenvalue values of <b>A</b> , in ascending order <code>ie</code> , sorted so that <code>w(i) &lt;= w(i+1)</code> .
<code>computeType</code>	host	in	data type of computation.
<code>pBuffer</code>	device	in/out	Working space. Array of type <code>void</code> of size <code>workspaceInBytes</code> bytes.
<code>workspaceInBytes</code>	host	input	Size in bytes of <code>pBuffer</code> , returned by <code>cusolverDnSyevd_bufferSize</code> .
<code>info</code>	device	output	if <code>info = 0</code> , the operation is successful. if <code>info = -i</code> , the <code>i</code> -th parameter is wrong (not counting handle). if <code>info = i (&gt; 0)</code> , <code>info</code> indicates <code>i</code> off-diagonal elements of an intermediate tridiagonal form did not converge to zero;

The generic API has three different types, `dataTypeA` is data type of the matrix **A**, `dataTypeW` is data type of the matrix **w** and `computeType` is compute type of the operation. `cusolverDnSyevd` only supports the following four combinations.

#### Valid combination of data type and compute type

<code>DataTypeA</code>	<code>DataTypeW</code>	<code>ComputeType</code>	Meaning
<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	SSYEVD
<code>CUDA_R_64F</code>	<code>CUDA_R_64F</code>	<code>CUDA_R_64F</code>	DSYEVD
<code>CUDA_C_32F</code>	<code>CUDA_C_32F</code>	<code>CUDA_C_32F</code>	CHEEVD
<code>CUDA_C_64F</code>	<code>CUDA_C_64F</code>	<code>CUDA_C_64F</code>	ZHEEVD

**Status Returned**

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $n < 0$ , or $lda < \max(1, n)$ , or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.13. cusolverDn<t>syevdx()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSsyevdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnDsyevdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    double vl,
    double vu,
    int il,
    int iu,
    int *h_meig,
    const double *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnCcheevdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnZcheevdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    double vl,
    double vu,
    int il,
    int iu,
    int *h_meig,
    const double *W,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```

cusolverStatus_t
cusolverDnSsyevdx(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    float *W,
    float *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnDsyevdx(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    double vl,
    double vu,
    int il,
    int iu,
    int *h_meig,
    double *W,
    double *work,
    int lwork,
    int *devInfo);

```

The C and Z data types are complex valued single and double precision, respectively.

```

cusolverStatus_t
cusolverDnCheevdx(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    float *W,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZheevdx(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    double vl,
    double vu,
    int il,
    int iu,
    int *h_meig,
    double *W,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);

```

This function computes all or selection of the eigenvalues and optionally eigenvectors of a symmetric (Hermitian)  $n \times n$  matrix  $\mathbf{A}$ . The standard symmetric eigenvalue problem is

$$\mathbf{A} \cdot \mathbf{V} = \mathbf{V} \cdot \mathbf{\Lambda}$$

where  $\mathbf{\Lambda}$  is a real  $n \times h\_meig$  diagonal matrix.  $\mathbf{V}$  is an  $n \times h\_meig$  unitary matrix.  $h\_meig$  is the number of eigenvalues/eigenvectors computed by the routine,  $h\_meig$  is equal to  $n$  when the whole spectrum (e.g., `range = CUSOLVER_EIG_RANGE_ALL`) is requested. The diagonal elements of  $\mathbf{\Lambda}$  are the eigenvalues of  $\mathbf{A}$  in ascending order.

The user has to provide working space which is pointed by input parameter `work`. The input parameter `lwork` is size of the working space, and it is returned by `syevdx_bufferSize()`.

If output parameter `devInfo = -i` (less than zero), the  $i$ -th parameter is wrong (not counting handle). If `devInfo = i` (greater than zero),  $i$  off-diagonal elements of an intermediate tridiagonal form did not converge to zero.

if `jobz = CUSOLVER_EIG_MODE_VECTOR`, **A** contains the orthonormal eigenvectors of the matrix **A**. The eigenvectors are computed by a divide and conquer algorithm.

Appendix F.1 provides a simple example of `syevdx`.

#### API of `syevdx`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>jobz</code>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute eigenvalues only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute eigenvalues and eigenvectors.
<code>range</code>	host	input	specifies options to which selection of eigenvalues and optionally eigenvectors that need to be computed: <code>range = CUSOLVER_EIG_RANGE_ALL</code> : all eigenvalues/eigenvectors will be found, will becomes the classical <code>syevd/heeve</code> routine; <code>range = CUSOLVER_EIG_RANGE_V</code> : all eigenvalues/eigenvectors in the half-open interval $(vl, vu]$ will be found; <code>range = CUSOLVER_EIG_RANGE_I</code> : the <i>il</i> -th through <i>iu</i> -th eigenvalues/eigenvectors will be found;
<code>uplo</code>	host	input	specifies which part of <b>A</b> is stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of <b>A</b> is stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of <b>A</b> is stored.
<code>n</code>	host	input	number of rows (or columns) of matrix <b>A</b> .
<b>A</b>	device	in/out	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <i>n</i> -by- <i>n</i> upper triangular part of <b>A</b> contains the upper triangular part of the matrix <b>A</b> . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <i>n</i> -by- <i>n</i> lower triangular part of <b>A</b> contains the lower triangular part of the matrix <b>A</b> . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>devInfo = 0</code> , <b>A</b> contains the orthonormal eigenvectors of the matrix <b>A</b> . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of <b>A</b> are destroyed.
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> . <code>lda</code> is not less than <code>max(1, n)</code> .
<code>vl, vu</code>	host	input	real values float or double for (C, S) or (Z, D) precision respectively. If <code>range = CUSOLVER_EIG_RANGE_V</code> , the lower and upper bounds of the

			interval to be searched for eigenvalues. $vl > vu$ . Not referenced if <code>range = CUSOLVER_EIG_RANGE_ALL</code> or <code>range = CUSOLVER_EIG_RANGE_I</code> . Note that, if eigenvalues are very close to each other, it is well known that two different eigenvalues routines might find slightly different number of eigenvalues inside the same interval. This is due to the fact that different eigenvalue algorithms, or even same algorithm but different run might find eigenvalues within some rounding error close to the machine precision. Thus, if the user want to be sure not to miss any eigenvalue within the interval bound, we suggest that, the user subtract/add epsilon (machine precision) to the interval bound such as ( $vl=vl-eps$ , $vu=vu+eps$ ). this suggestion is valid for any selective routine from cuSolver or LAPACK.
<code>il, iu</code>	<code>host</code>	<code>input</code>	integer. If <code>range = CUSOLVER_EIG_RANGE_I</code> , the indices (in ascending order) of the smallest and largest eigenvalues to be returned. $1 \leq il \leq iu \leq n$ , if $n > 0$ ; $il = 1$ and $iu = 0$ if $n = 0$ . Not referenced if <code>range = CUSOLVER_EIG_RANGE_ALL</code> or <code>range = CUSOLVER_EIG_RANGE_V</code> .
<code>h_meig</code>	<code>host</code>	<code>output</code>	integer. The total number of eigenvalues found. $0 \leq h\_meig \leq n$ . If <code>range = CUSOLVER_EIG_RANGE_ALL</code> , $h\_meig = n$ , and if <code>range = CUSOLVER_EIG_RANGE_I</code> , $h\_meig = iu-il+1$ .
<code>w</code>	<code>device</code>	<code>output</code>	a real array of dimension $n$ . The eigenvalue values of $A$ , in ascending order $ie$ , sorted so that $w(i) \leq w(i+1)$ .
<code>work</code>	<code>device</code>	<code>in/out</code>	working space, <type> array of size <code>lwork</code> .
<code>lwork</code>	<code>host</code>	<code>input</code>	size of <code>work</code> , returned by <code>syevdx_bufferSize</code> .
<code>devInfo</code>	<code>device</code>	<code>output</code>	if <code>devInfo = 0</code> , the operation is successful. if <code>devInfo = -i</code> , the $i$ -th parameter is wrong (not counting handle). if <code>devInfo = i (&gt; 0)</code> , <code>devInfo</code> indicates $i$ off-diagonal elements of an intermediate tridiagonal form did not converge to zero;

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.

<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n &lt; 0</code> , or <code>lda &lt; max(1, n)</code> , or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>range</code> is not <code>CUSOLVER_EIG_RANGE_ALL</code> or <code>CUSOLVER_EIG_RANGE_V</code> or <code>CUSOLVER_EIG_RANGE_I</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.14. `cusolverDnSyevedx()`

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSyevedx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverParams_t params,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    cudaDataType dataTypeA,
    const void *A,
    int64_t lda,
    void *vl,
    void *vu,
    int64_t il,
    int64_t iu,
    int64_t *h_meig,
    cudaDataType dataTypeW,
    const void *W,
    cudaDataType computeType,
    size_t *workspaceInBytes);
```

## The routine bellow

```
cusolverStatus_t
cusolverDnSyevdx (
  cusolverDnHandle_t handle,
  cusolverParams_t params,
  cusolverEigMode_t jobz,
  cusolverEigRange_t range,
  cublasFillMode_t uplo,
  int n,
  cudaDataType_t dataTypeA,
  const void *A,
  int64_t lda,
  void *vl,
  void *vu,
  int64_t il,
  int64_t iu,
  int64_t *h_meig,
  cudaDataType_t dataTypeW,
  const void *W,
  cudaDataType_t computeType,
  void *pBuffer,
  size_t workspaceInBytes,
  int *info);
```

computes all or selection of the eigenvalues and optionally eigenvectors of a symmetric (Hermitian)  $n \times n$  matrix  $\mathbf{A}$  using the generic API interface. The standard symmetric eigenvalue problem is

$$\mathbf{A}^* \mathbf{V} = \mathbf{V}^* \mathbf{\Lambda}$$

where  $\mathbf{\Lambda}$  is a real  $n \times h\_meig$  diagonal matrix.  $\mathbf{V}$  is an  $n \times h\_meig$  unitary matrix.  $h\_meig$  is the number of eigenvalues/eigenvectors computed by the routine,  $h\_meig$  is equal to  $n$  when the whole spectrum (e.g., `range = CUSOLVER_EIG_RANGE_ALL`) is requested. The diagonal elements of  $\mathbf{\Lambda}$  are the eigenvalues of  $\mathbf{A}$  in ascending order.

The user has to provide working space which is pointed by input parameter `pBuffer`. The input parameter `workspaceInBytes` is size in bytes of the working space, and it is returned by `cusolverDnSyevdx_bufferSize()`.

If output parameter `info = -i` (less than zero), the  $i$ -th parameter is wrong (not counting handle). If `info = i` (greater than zero),  $i$  off-diagonal elements of an intermediate tridiagonal form did not converge to zero.

if `jobz = CUSOLVER_EIG_MODE_VECTOR`,  $\mathbf{A}$  contains the orthonormal eigenvectors of the matrix  $\mathbf{A}$ . The eigenvectors are computed by a divide and conquer algorithm.

Currently, `cusolverDnSyevdx` supports only the default algorithm.

### Table of algorithms supported by `cusolverDnSyevdx`

<code>CUSOLVER_ALG_0</code> or <code>NULL</code>	Default algorithm.
--------------------------------------------------	--------------------

List of input arguments for `cusolverDnSyevdx_bufferSize` and `cusolverDnSyevdx`:

### API of `cusolverDnSyevdx`

parameter	Memory	In/out	Meaning
-----------	--------	--------	---------

<b>handle</b>	<b>host</b>	<b>input</b>	handle to the cuSolverDN library context.
<b>params</b>	<b>host</b>	<b>input</b>	structure with information collected by <code>cusolverDnSetAdvOptions</code> .
<b>jobz</b>	<b>host</b>	<b>input</b>	specifies options to either compute eigenvalue only or compute eigen-pair: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute eigenvalues only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute eigenvalues and eigenvectors.
<b>range</b>	<b>host</b>	<b>input</b>	specifies options to which selection of eigenvalues and optionally eigenvectors that need to be computed: <code>range = CUSOLVER_EIG_RANGE_ALL</code> : all eigenvalues/eigenvectors will be found, will becomes the classical <code>syevd/heeved</code> routine; <code>range = CUSOLVER_EIG_RANGE_V</code> : all eigenvalues/eigenvectors in the half-open interval $(vl, vu]$ will be found; <code>range = CUSOLVER_EIG_RANGE_I</code> : the <i>il</i> -th through <i>iu</i> -th eigenvalues/eigenvectors will be found;
<b>uplo</b>	<b>host</b>	<b>input</b>	specifies which part of <b>A</b> is stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of <b>A</b> is stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of <b>A</b> is stored.
<b>n</b>	<b>host</b>	<b>input</b>	number of rows (or columns) of matrix <b>A</b> .
<b>dataTypeA</b>	<b>host</b>	<b>in</b>	data type of array <b>A</b> .
<b>A</b>	<b>device</b>	<b>in/out</b>	array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <i>n</i> -by- <i>n</i> upper triangular part of <b>A</b> contains the upper triangular part of the matrix <b>A</b> . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <i>n</i> -by- <i>n</i> lower triangular part of <b>A</b> contains the lower triangular part of the matrix <b>A</b> . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>info = 0</code> , <b>A</b> contains the orthonormal eigenvectors of the matrix <b>A</b> . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of <b>A</b> are destroyed.
<b>lda</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>A</b> . <code>lda</code> is not less than <code>max(1, n)</code> .
<b>vl, vu</b>	<b>host</b>	<b>input</b>	If <code>range = CUSOLVER_EIG_RANGE_V</code> , the lower and upper bounds of the interval to be searched for eigenvalues. <code>vl &gt; vu</code> . Not referenced if <code>range = CUSOLVER_EIG_RANGE_ALL</code> or <code>range = CUSOLVER_EIG_RANGE_I</code> . Note that, if eigenvalues are very close to each

			other, it is well known that two different eigenvalues routines might find slightly different number of eigenvalues inside the same interval. This is due to the fact that different eigenvalue algorithms, or even same algorithm but different run might find eigenvalues within some rounding error close to the machine precision. Thus, if the user want to be sure not to miss any eigenvalue within the interval bound, we suggest that, the user subtract/add epsilon (machine precision) to the interval bound such as $(vl=vl-eps, vu=vu+eps)$ . this suggestion is valid for any selective routine from cuSolver or LAPACK.
<code>il, iu</code>	<code>host</code>	<code>input</code>	integer. If <code>range = CUSOLVER_EIG_RANGE_I</code> , the indices (in ascending order) of the smallest and largest eigenvalues to be returned. $1 \leq il \leq iu \leq n$ , if $n > 0$ ; $il = 1$ and $iu = 0$ if $n = 0$ . Not referenced if <code>range = CUSOLVER_EIG_RANGE_ALL</code> or <code>range = CUSOLVER_EIG_RANGE_V</code> .
<code>h_meig</code>	<code>host</code>	<code>output</code>	integer. The total number of eigenvalues found. $0 \leq h\_meig \leq n$ . If <code>range = CUSOLVER_EIG_RANGE_ALL</code> , $h\_meig = n$ , and if <code>range = CUSOLVER_EIG_RANGE_I</code> , $h\_meig = iu-il+1$ .
<code>dataTypeW</code>	<code>host</code>	<code>in</code>	data type of array <code>w</code> .
<code>w</code>	<code>device</code>	<code>output</code>	a real array of dimension $n$ . The eigenvalue values of <code>A</code> , in ascending order ie, sorted so that $w(i) \leq w(i+1)$ .
<code>computeType</code>	<code>host</code>	<code>in</code>	data type of computation.
<code>pBuffer</code>	<code>device</code>	<code>in/out</code>	Working space. Array of type <code>void</code> of size <code>workspaceInBytes</code> bytes.
<code>workspaceInBytes</code>	<code>host</code>	<code>input</code>	Size in bytes of <code>pBuffer</code> , returned by <code>cusolverDnSyevdx_bufferSize</code> .
<code>info</code>	<code>device</code>	<code>output</code>	if <code>info = 0</code> , the operation is successful. if <code>info = -i</code> , the $i$ -th parameter is wrong (not counting handle). if <code>info = i (&gt; 0)</code> , <code>info</code> indicates $i$ off-diagonal elements of an intermediate tridiagonal form did not converge to zero;

The generic API has three different types, `dataTypeA` is data type of the matrix `A`, `dataTypeW` is data type of the matrix `W` and `computeType` is compute type of the operation. `cusolverDnSyevdx` only supports the following four combinations.

#### Valid combination of data type and compute type

<code>dataTypeA</code>	<code>dataTypeW</code>	<code>computeType</code>	Meaning
<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	<code>CUDA_R_32F</code>	<code>SSYEVDX</code>

CUDA_R_64F	CUDA_R_64F	CUDA_R_64F	DSYEVDX
CUDA_C_32F	CUDA_C_32F	CUDA_C_32F	CHEEVDX
CUDA_C_64F	CUDA_C_64F	CUDA_C_64F	ZHEEVDX

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $n < 0$ , or $lda < \max(1, n)$ , or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>range</code> is not <code>CUSOLVER_EIG_RANGE_ALL</code> or <code>CUSOLVER_EIG_RANGE_V</code> or <code>CUSOLVER_EIG_RANGE_I</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

### 2.4.3.15. cusolverDn<t>sygvd()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSsygvd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *B,
    int ldb,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnDsygvd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *B,
    int ldb,
    const double *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnChegvd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const cuComplex *B,
    int ldb,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnZhegvd_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *B,
    int ldb,
    const double *W,
    int *lwork);
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t  
cusolverDnSsygvd(  
    cusolverDnHandle_t handle,  
    cusolverEigType_t itype,  
    cusolverEigMode_t jobz,  
    cublasFillMode_t uplo,  
    int n,  
    float *A,  
    int lda,  
    float *B,  
    int ldb,  
    float *W,  
    float *work,  
    int lwork,  
    int *devInfo);
```

```
cusolverStatus_t  
cusolverDnDsygvd(  
    cusolverDnHandle_t handle,  
    cusolverEigType_t itype,  
    cusolverEigMode_t jobz,  
    cublasFillMode_t uplo,  
    int n,  
    double *A,  
    int lda,  
    double *B,  
    int ldb,  
    double *W,  
    double *work,  
    int lwork,  
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnChegvd(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    cuComplex *B,
    int ldb,
    float *W,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZhegvd(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    cuDoubleComplex *B,
    int ldb,
    double *W,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function computes eigenvalues and eigenvectors of a symmetric (Hermitian)  $n \times n$  matrix-pair  $(\mathbf{A}, \mathbf{B})$ . The generalized symmetric-definite eigenvalue problem is

$$\text{eig}(\mathbf{A}, \mathbf{B}) = \begin{cases} \mathbf{A}^* \mathbf{V} = \mathbf{B}^* \mathbf{V} \boldsymbol{\Lambda} & \text{if } \text{itype} = \text{CUSOLVER\_EIG\_TYPE\_1} \\ \mathbf{A}^* \mathbf{B}^* \mathbf{V} = \mathbf{V} \boldsymbol{\Lambda} & \text{if } \text{itype} = \text{CUSOLVER\_EIG\_TYPE\_2} \\ \mathbf{B}^* \mathbf{A}^* \mathbf{V} = \mathbf{V} \boldsymbol{\Lambda} & \text{if } \text{itype} = \text{CUSOLVER\_EIG\_TYPE\_3} \end{cases}$$

where the matrix  $\mathbf{B}$  is positive definite.  $\boldsymbol{\Lambda}$  is a real  $n \times n$  diagonal matrix. The diagonal elements of  $\boldsymbol{\Lambda}$  are the eigenvalues of  $(\mathbf{A}, \mathbf{B})$  in ascending order.  $\mathbf{V}$  is an  $n \times n$  orthogonal matrix. The eigenvectors are normalized as follows:

$$\begin{cases} \mathbf{V}^H * \mathbf{B} * \mathbf{V} = \mathbf{I} & \text{if } \text{itype} = \text{CUSOLVER\_EIG\_TYPE\_1}, \text{CUSOLVER\_EIG\_TYPE\_2} \\ \mathbf{V}^H * \text{inv}(\mathbf{B}) * \mathbf{V} = \mathbf{I} & \text{if } \text{itype} = \text{CUSOLVER\_EIG\_TYPE\_3} \end{cases}$$

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is size of the working space, and it is returned by **sygvd\_bufferSize()**.

If output parameter **devInfo** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle). If **devInfo** = **i** ( $i > 0$  and  $i \leq n$ ) and **jobz** = CUSOLVER\_EIG\_MODE\_NOVECTOR, **i** off-diagonal elements of an intermediate tridiagonal form did not converge to zero. If **devInfo** = **N + i** ( $i > 0$ ), then the

leading minor of order  $i$  of  $\mathbf{B}$  is not positive definite. The factorization of  $\mathbf{B}$  could not be completed and no eigenvalues or eigenvectors were computed.

if `jobz = CUSOLVER_EIG_MODE_VECTOR`,  $\mathbf{A}$  contains the orthogonal eigenvectors of the matrix  $\mathbf{A}$ . The eigenvectors are computed by divide and conquer algorithm.

Appendix F.2 provides a simple example of `sygvd`.

### API of `sygvd`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>itype</code>	host	input	Specifies the problem type to be solved: <code>itype=CUSOLVER_EIG_TYPE_1</code> : $A*x = (\text{lambda})*B*x$ . <code>itype=CUSOLVER_EIG_TYPE_2</code> : $A*B*x = (\text{lambda})*x$ . <code>itype=CUSOLVER_EIG_TYPE_3</code> : $B*A*x = (\text{lambda})*x$ .
<code>jobz</code>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute eigenvalues only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute eigenvalues and eigenvectors.
<code>uplo</code>	host	input	specifies which part of $\mathbf{A}$ and $\mathbf{B}$ are stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of $\mathbf{A}$ and $\mathbf{B}$ are stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of $\mathbf{A}$ and $\mathbf{B}$ are stored.
<code>n</code>	host	input	number of rows (or columns) of matrix $\mathbf{A}$ and $\mathbf{B}$ .
<code>A</code>	device	in/out	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading $n$ -by- $n$ upper triangular part of $\mathbf{A}$ contains the upper triangular part of the matrix $\mathbf{A}$ . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading $n$ -by- $n$ lower triangular part of $\mathbf{A}$ contains the lower triangular part of the matrix $\mathbf{A}$ . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>devInfo = 0</code> , $\mathbf{A}$ contains the orthonormal eigenvectors of the matrix $\mathbf{A}$ . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of $\mathbf{A}$ are destroyed.
<code>lda</code>	host	input	leading dimension of two-dimensional array used to store matrix $\mathbf{A}$ . <code>lda</code> is not less than <code>max(1, n)</code> .
<code>B</code>	device	in/out	<type> array of dimension <code>ldb * n</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading $n$ -by- $n$ upper triangular part of $\mathbf{B}$ contains the upper triangular part of the matrix $\mathbf{B}$ . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading

			n-by-n lower triangular part of <b>B</b> contains the lower triangular part of the matrix <b>B</b> . On exit, if <code>devInfo</code> is less than <code>n</code> , <b>B</b> is overwritten by triangular factor <b>U</b> or <b>L</b> from the Cholesky factorization of <b>B</b> .
<code>ldb</code>	<code>host</code>	<code>input</code>	leading dimension of two-dimensional array used to store matrix <b>B</b> . <code>ldb</code> is not less than <code>max(1, n)</code> .
<code>W</code>	<code>device</code>	<code>output</code>	a real array of dimension <code>n</code> . The eigenvalue values of <b>A</b> , sorted so that <code>W(i) &gt;= W(i+1)</code> .
<code>work</code>	<code>device</code>	<code>in/out</code>	working space, <code>&lt;type&gt;</code> array of size <code>lwork</code> .
<code>lwork</code>	<code>host</code>	<code>input</code>	size of <code>work</code> , returned by <code>sygvd_bufferSize</code> .
<code>devInfo</code>	<code>device</code>	<code>output</code>	if <code>devInfo = 0</code> , the operation is successful. if <code>devInfo = -i</code> , the <code>i</code> -th parameter is wrong (not counting handle). if <code>devInfo = i (&gt; 0)</code> , <code>devInfo</code> indicates either <code>potrf</code> or <code>syevd</code> is wrong.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n &lt; 0</code> , or <code>lda &lt; max(1, n)</code> , or <code>ldb &lt; max(1, n)</code> , or <code>itype</code> is not 1, 2 or 3, or <code>jobz</code> is not 'N' or 'V', or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.16. cusolverDn<t>sygvdx()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverDnSsygvdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *B,
    int ldb,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnDsygvdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *B,
    int ldb,
    double vl,
    double vu,
    int il,
    int iu,
    int *h_meig,
    const double *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnChegvdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const cuComplex *B,
    int ldb,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    const float *W,
    int *lwork);
```

```
cusolverStatus_t
cusolverDnZhegvdx_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *B,
```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSsygvdx(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    float *B,
    int ldb,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    float *W,
    float *work,
    int lwork,
    int *devInfo);
```

```
cusolverStatus_t
cusolverDnDsygvdx(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    double *B,
    int ldb,
    double vl,
    double vu,
    int il,
    int iu,
    int *h_meig,
    double *W,
    double *work,
    int lwork,
    int *devInfo);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnChegvdx(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    cuComplex *B,
    int ldb,
    float vl,
    float vu,
    int il,
    int iu,
    int *h_meig,
    float *W,
    cuComplex *work,
    int lwork,
    int *devInfo);

cusolverStatus_t
cusolverDnZhegvdx(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cusolverEigRange_t range,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    cuDoubleComplex *B,
    int ldb,
    double vl,
    double vu,
    int il,
    int iu,
    int *h_meig,
    double *W,
    cuDoubleComplex *work,
    int lwork,
    int *devInfo);
```

This function computes all or selection of the eigenvalues and optionally eigenvectors of a symmetric (Hermitian)  $n \times n$  matrix-pair  $(\mathbf{A}, \mathbf{B})$ . The generalized symmetric-definite eigenvalue problem is

$$\text{eig}(\mathbf{A}, \mathbf{B}) = \begin{cases} \mathbf{A}^* \mathbf{V} = \mathbf{B}^* \mathbf{V} \boldsymbol{\Lambda} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_1} \\ \mathbf{A}^* \mathbf{B}^* \mathbf{V} = \mathbf{V} \boldsymbol{\Lambda} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_2} \\ \mathbf{B}^* \mathbf{A}^* \mathbf{V} = \mathbf{V} \boldsymbol{\Lambda} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_3} \end{cases}$$

where the matrix  $\mathbf{B}$  is positive definite.  $\boldsymbol{\Lambda}$  is a real  $n \times h\_meig$  diagonal matrix. The diagonal elements of  $\boldsymbol{\Lambda}$  are the eigenvalues of  $(\mathbf{A}, \mathbf{B})$  in ascending order.  $\mathbf{V}$  is an  $n \times h\_meig$  orthogonal matrix.  $h\_meig$  is the number of eigenvalues/eigenvectors

computed by the routine, **h\_meig** is equal to **n** when the whole spectrum (e.g., **range = CUSOLVER\_EIG\_RANGE\_ALL**) is requested. The eigenvectors are normalized as follows:

$$\begin{cases} V^H * B * V = I & \text{if } \text{itype} = \text{CUSOLVER\_EIG\_TYPE\_1}, \text{CUSOLVER\_EIG\_TYPE\_2} \\ V^H * \text{inv}(B) * V = I & \text{if } \text{itype} = \text{CUSOLVER\_EIG\_TYPE\_3} \end{cases}$$

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is size of the working space, and it is returned by **sygvdx\_bufferSize()**.

If output parameter **devInfo = -i** (less than zero), the **i-th** parameter is wrong (not counting handle). If **devInfo = i** ( $i > 0$  and  $i \leq n$ ) and **jobz = CUSOLVER\_EIG\_MODE\_NOVECTOR**, **i** off-diagonal elements of an intermediate tridiagonal form did not converge to zero. If **devInfo = n + i** ( $i > 0$ ), then the leading minor of order **i** of **B** is not positive definite. The factorization of **B** could not be completed and no eigenvalues or eigenvectors were computed.

if **jobz = CUSOLVER\_EIG\_MODE\_VECTOR**, **A** contains the orthogonal eigenvectors of the matrix **A**. The eigenvectors are computed by divide and conquer algorithm.

Appendix F.2 provides a simple example of **sygvdx**.

#### API of sygvdx

parameter	Memory	In/out	Meaning
<b>handle</b>	<b>host</b>	<b>input</b>	handle to the cuSolverDN library context.
<b>itype</b>	<b>host</b>	<b>input</b>	Specifies the problem type to be solved: <b>itype=CUSOLVER_EIG_TYPE_1</b> : $A*x = (\text{lambda})*B*x$ . <b>itype=CUSOLVER_EIG_TYPE_2</b> : $A*B*x = (\text{lambda})*x$ . <b>itype=CUSOLVER_EIG_TYPE_3</b> : $B*A*x = (\text{lambda})*x$ .
<b>jobz</b>	<b>host</b>	<b>input</b>	specifies options to either compute eigenvalue only or compute eigen-pair: <b>jobz = CUSOLVER_EIG_MODE_NOVECTOR</b> : Compute eigenvalues only; <b>jobz = CUSOLVER_EIG_MODE_VECTOR</b> : Compute eigenvalues and eigenvectors.
<b>range</b>	<b>host</b>	<b>input</b>	specifies options to which selection of eigenvalues and optionally eigenvectors that need to be computed: <b>range = CUSOLVER_EIG_RANGE_ALL</b> : all eigenvalues/eigenvectors will be found, will become the classical syevd/heevd routine; <b>range = CUSOLVER_EIG_RANGE_V</b> : all eigenvalues/eigenvectors in the half-open interval $(vl, vu]$ will be found; <b>range = CUSOLVER_EIG_RANGE_I</b> : the <b>il</b> -th through <b>iu</b> -th eigenvalues/eigenvectors will be found;
<b>uplo</b>	<b>host</b>	<b>input</b>	specifies which part of <b>A</b> and <b>B</b> are stored. <b>uplo = CUBLAS_FILL_MODE_LOWER</b> :

			Lower triangle of <b>A</b> and <b>B</b> are stored. <b>uplo</b> = <b>CUBLAS_FILL_MODE_UPPER</b> : Upper triangle of <b>A</b> and <b>B</b> are stored.
<b>n</b>	<b>host</b>	<b>input</b>	number of rows (or columns) of matrix <b>A</b> and <b>B</b> .
<b>A</b>	<b>device</b>	<b>in/out</b>	<type> array of dimension <b>lda</b> * <b>n</b> with <b>lda</b> is not less than <b>max(1, n)</b> . If <b>uplo</b> = <b>CUBLAS_FILL_MODE_UPPER</b> , the leading <b>n</b> -by- <b>n</b> upper triangular part of <b>A</b> contains the upper triangular part of the matrix <b>A</b> . If <b>uplo</b> = <b>CUBLAS_FILL_MODE_LOWER</b> , the leading <b>n</b> -by- <b>n</b> lower triangular part of <b>A</b> contains the lower triangular part of the matrix <b>A</b> . On exit, if <b>jobz</b> = <b>CUSOLVER_EIG_MODE_VECTOR</b> , and <b>devInfo</b> = 0, <b>A</b> contains the orthonormal eigenvectors of the matrix <b>A</b> . If <b>jobz</b> = <b>CUSOLVER_EIG_MODE_NOVECTOR</b> , the contents of <b>A</b> are destroyed.
<b>lda</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>A</b> . <b>lda</b> is not less than <b>max(1, n)</b> .
<b>B</b>	<b>device</b>	<b>in/out</b>	<type> array of dimension <b>ldb</b> * <b>n</b> . If <b>uplo</b> = <b>CUBLAS_FILL_MODE_UPPER</b> , the leading <b>n</b> -by- <b>n</b> upper triangular part of <b>B</b> contains the upper triangular part of the matrix <b>B</b> . If <b>uplo</b> = <b>CUBLAS_FILL_MODE_LOWER</b> , the leading <b>n</b> -by- <b>n</b> lower triangular part of <b>B</b> contains the lower triangular part of the matrix <b>B</b> . On exit, if <b>devInfo</b> is less than <b>n</b> , <b>B</b> is overwritten by triangular factor <b>U</b> or <b>L</b> from the Cholesky factorization of <b>B</b> .
<b>ldb</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>B</b> . <b>ldb</b> is not less than <b>max(1, n)</b> .
<b>vl, vu</b>	<b>host</b>	<b>input</b>	real values float or double for (C, S) or (Z, D) precision respectively. If <b>range</b> = <b>CUSOLVER_EIG_RANGE_V</b> , the lower and upper bounds of the interval to be searched for eigenvalues. <b>vl</b> > <b>vu</b> . Not referenced if <b>range</b> = <b>CUSOLVER_EIG_RANGE_ALL</b> or <b>range</b> = <b>CUSOLVER_EIG_RANGE_I</b> . Note that, if eigenvalues are very close to each other, it is well known that two different eigenvalues routines might find slightly different number of eigenvalues inside the same interval. This is due to the fact that different eigenvalue algorithms, or even same algorithm but different run might find eigenvalues within some rounding error close to the machine precision. Thus, if the user want to be sure not to miss any eigenvalue within the interval bound, we suggest that, the user

			subtract/add epsilon (machine precision) to the interval bound such as (vl=vl-eps, vu=vu+eps]. this suggestion is valid for any selective routine from cuSolver or LAPACK.
il,iu	host	input	integer. If <code>range = CUSOLVER_EIG_RANGE_I</code> , the indices (in ascending order) of the smallest and largest eigenvalues to be returned. $1 \leq il \leq iu \leq n$ , if $n > 0$ ; $il = 1$ and $iu = 0$ if $n = 0$ . Not referenced if <code>range = CUSOLVER_EIG_RANGE_ALL</code> or <code>range = CUSOLVER_EIG_RANGE_V</code> .
h_meig	host	output	integer. The total number of eigenvalues found. $0 \leq h\_meig \leq n$ . If <code>range = CUSOLVER_EIG_RANGE_ALL</code> , $h\_meig = n$ , and if <code>range = CUSOLVER_EIG_RANGE_I</code> , $h\_meig = iu-il+1$ .
W	device	output	a real array of dimension $n$ . The eigenvalue values of $A$ , sorted so that $W(i) \geq W(i+1)$ .
work	device	in/out	working space, <type> array of size <code>lwork</code> .
lwork	host	input	size of <code>work</code> , returned by <code>sygvdx_bufferSize</code> .
devInfo	device	output	if <code>devInfo = 0</code> , the operation is successful. if <code>devInfo = -i</code> , the $i$ -th parameter is wrong (not counting handle). if <code>devInfo = i (&gt; 0)</code> , <code>devInfo</code> indicates either <code>potrf</code> or <code>syevd</code> is wrong.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $n < 0$ , or $lda < \max(1, n)$ , or $ldb < \max(1, n)$ , or <code>itype</code> is not <code>CUSOLVER_EIG_TYPE_1</code> or <code>CUSOLVER_EIG_TYPE_2</code> or <code>CUSOLVER_EIG_TYPE_3</code> or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTORL</code> , or <code>range</code> is not <code>CUSOLVER_EIG_RANGE_ALL</code> or <code>CUSOLVER_EIG_RANGE_V</code> or <code>CUSOLVER_EIG_RANGE_I</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.17. cusolverDn<t>syevj()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```

cusolverStatus_t
cusolverDnSsyevj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *W,
    int *lwork,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnDsyevj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *W,
    int *lwork,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnCsyevj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const float *W,
    int *lwork,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnZsyevj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const double *W,
    int *lwork,
    syevjInfo_t params);

```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSsyevj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    float *W,
    float *work,
    int lwork,
    int *info,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnDsyevj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    double *W,
    double *work,
    int lwork,
    int *info,
    syevjInfo_t params);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCchevj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    float *W,
    cuComplex *work,
    int lwork,
    int *info,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnZchevj(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    double *W,
    cuDoubleComplex *work,
    int lwork,
    int *info,
    syevjInfo_t params);
```

This function computes eigenvalues and eigenvectors of a symmetric (Hermitian)  $n \times n$  matrix **A**. The standard symmetric eigenvalue problem is

$$A * Q = Q * \Lambda$$

where  $\Lambda$  is a real  $n \times n$  diagonal matrix.  $Q$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\Lambda$  are the eigenvalues of  $A$  in ascending order.

**syevj** has the same functionality as **syevd**. The difference is that **syevd** uses QR algorithm and **syevj** uses Jacobi method. The parallelism of Jacobi method gives GPU better performance on small and medium size matrices. Moreover the user can configure **syevj** to perform approximation up to certain accuracy.

How does it work?

**syevj** iteratively generates a sequence of unitary matrices to transform matrix  $A$  to the following form

$$V^H * A * V = W + E$$

where  $W$  is diagonal and  $E$  is symmetric without diagonal.

During the iterations, the Frobenius norm of  $E$  decreases monotonically. As  $E$  goes down to zero,  $W$  is the set of eigenvalues. In practice, Jacobi method stops if

$$\|E\|_F \leq \text{eps} * \|A\|_F$$

where **eps** is given tolerance.

**syevj** has two parameters to control the accuracy. First parameter is tolerance (**eps**). The default value is machine accuracy but The user can use function **cusolverDnXsyevjSetTolerance** to set a priori tolerance. The second parameter is maximum number of sweeps which controls number of iterations of Jacobi method. The default value is 100 but the user can use function **cusolverDnXsyevjSetMaxSweeps** to set a proper bound. The experimentis show 15 sweeps are good enough to converge to machine accuracy. **syevj** stops either tolerance is met or maximum number of sweeps is met.

Jacobi method has quadratic convergence, so the accuracy is not proportional to number of sweeps. To guarantee certain accuracy, the user should configure tolerance only.

After **syevj**, the user can query residual by function **cusolverDnXsyevjGetResidual** and number of executed sweeps by function **cusolverDnXsyevjGetSweeps**. However the user needs to be aware that residual is the Frobenius norm of  $E$ , not accuracy of individual eigenvalue, i.e.

$$\text{residual} = \|E\|_F = \|\Lambda - W\|_F$$

The same as **syevd**, the user has to provide working space pointed by input parameter **work**. The input parameter **lwork** is the size of the working space, and it is returned by **syevj\_bufferSize()**.

If output parameter **info** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle). If **info** = **n+1**, **syevj** does not converge under given tolerance and maximum sweeps.

If the user sets an improper tolerance, **syevj** may not converge. For example, tolerance should not be smaller than machine accuracy.

if **jobz** = CUSOLVER\_EIG\_MODE\_VECTOR, **A** contains the orthonormal eigenvectors **V**.

Appendix F.3 provides a simple example of **syevj**.

### API of syevj

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverDN library context.
<b>jobz</b>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <b>jobz</b> = CUSOLVER_EIG_MODE_NOVECTOR : Compute eigenvalues only; <b>jobz</b> = CUSOLVER_EIG_MODE_VECTOR : Compute eigenvalues and eigenvectors.
<b>uplo</b>	host	input	specifies which part of <b>A</b> is stored. <b>uplo</b> = CUBLAS_FILL_MODE_LOWER: Lower triangle of <b>A</b> is stored. <b>uplo</b> = CUBLAS_FILL_MODE_UPPER: Upper triangle of <b>A</b> is stored.
<b>n</b>	host	input	number of rows (or columns) of matrix <b>A</b> .
<b>A</b>	device	in/out	<type> array of dimension <b>lda</b> * <b>n</b> with <b>lda</b> is not less than $\max(1, n)$ . If <b>uplo</b> = CUBLAS_FILL_MODE_UPPER, the leading n-by-n upper triangular part of <b>A</b> contains the upper triangular part of the matrix <b>A</b> . If <b>uplo</b> = CUBLAS_FILL_MODE_LOWER, the leading n-by-n lower triangular part of <b>A</b> contains the lower triangular part of the matrix <b>A</b> . On exit, if <b>jobz</b> = CUSOLVER_EIG_MODE_VECTOR, and <b>info</b> = 0, <b>A</b> contains the orthonormal eigenvectors of the matrix <b>A</b> . If <b>jobz</b> = CUSOLVER_EIG_MODE_NOVECTOR, the contents of <b>A</b> are destroyed.
<b>lda</b>	host	input	leading dimension of two-dimensional array used to store matrix <b>A</b> .
<b>W</b>	device	output	a real array of dimension <b>n</b> . The eigenvalue values of <b>A</b> , in ascending order ie, sorted so that $W(i) \leq W(i+1)$ .
<b>work</b>	device	in/out	working space, <type> array of size <b>lwork</b> .
<b>lwork</b>	host	input	size of <b>work</b> , returned by <b>syevj_bufferSize</b> .
<b>info</b>	device	output	if <b>info</b> = 0, the operation is successful. if <b>info</b> = -i, the i-th parameter is wrong (not counting handle). if <b>info</b> = <b>n</b> +1, <b>syevj</b> dose not converge under given tolerance and maximum sweeps.
<b>params</b>	host	in/out	structure filled with parameters of Jacobi algorithm and results of <b>syevj</b> .

**Status Returned**

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $n < 0$ , or $lda < \max(1, n)$ , or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.18. cusolverDn<t>sygvj()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```

cusolverStatus_t
cusolverDnSsygvj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *B,
    int ldb,
    const float *W,
    int *lwork,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnDsygvj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *B,
    int ldb,
    const double *W,
    int *lwork,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnCsygvj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const cuComplex *B,
    int ldb,
    const float *W,
    int *lwork,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnZsygvj_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *B,
    int ldb,
    const double *W,
    int *lwork,
    syevjInfo_t params);

```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSsygvj(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    float *B,
    int ldb,
    float *W,
    float *work,
    int lwork,
    int *info,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnDsygvj(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    double *B,
    int ldb,
    double *W,
    double *work,
    int lwork,
    int *info,
    syevjInfo_t params);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnChegvj(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    cuComplex *B,
    int ldb,
    float *W,
    cuComplex *work,
    int lwork,
    int *info,
    syevjInfo_t params);

cusolverStatus_t
cusolverDnZhegvj(
    cusolverDnHandle_t handle,
    cusolverEigType_t itype,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    cuDoubleComplex *B,
    int ldb,
    double *W,
    cuDoubleComplex *work,
    int lwork,
    int *info,
    syevjInfo_t params);
```

This function computes eigenvalues and eigenvectors of a symmetric (Hermitian)  $\mathbf{n} \times \mathbf{n}$  matrix-pair  $(\mathbf{A}, \mathbf{B})$ . The generalized symmetric-definite eigenvalue problem is

$$\text{eig}(\mathbf{A}, \mathbf{B}) = \begin{cases} \mathbf{A}^* \mathbf{V} = \mathbf{B}^* \mathbf{V}^* \boldsymbol{\Lambda} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_1} \\ \mathbf{A}^* \mathbf{B}^* \mathbf{V} = \mathbf{V}^* \boldsymbol{\Lambda} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_2} \\ \mathbf{B}^* \mathbf{A}^* \mathbf{V} = \mathbf{V}^* \boldsymbol{\Lambda} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_3} \end{cases}$$

where the matrix  $\mathbf{B}$  is positive definite.  $\boldsymbol{\Lambda}$  is a real  $\mathbf{n} \times \mathbf{n}$  diagonal matrix. The diagonal elements of  $\boldsymbol{\Lambda}$  are the eigenvalues of  $(\mathbf{A}, \mathbf{B})$  in ascending order.  $\mathbf{V}$  is an  $\mathbf{n} \times \mathbf{n}$  orthogonal matrix. The eigenvectors are normalized as follows:

$$\begin{cases} \mathbf{V}^H \mathbf{B}^* \mathbf{V} = \mathbf{I} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_1}, \text{CUSOLVER\_EIG\_TYPE\_2} \\ \mathbf{V}^H \text{inv}(\mathbf{B})^* \mathbf{V} = \mathbf{I} & \text{if itype} = \text{CUSOLVER\_EIG\_TYPE\_3} \end{cases}$$

This function has the same functionality as `sygvd` except that `syevd` in `sygvd` is replaced by `syevj` in `sygvj`. Therefore, `sygvj` inherits properties of `syevj`, the user can use `cusolverDnXsyevjSetTolerance` and `cusolverDnXsyevjSetMaxSweeps` to configure tolerance and maximum sweeps.

However the meaning of residual is different from **syevj**. **sygvj** first computes Cholesky factorization of matrix **B**,

$$B = L * L^H$$

transform the problem to standard eigenvalue problem, then calls **syevj**.

For example, the standard eigenvalue problem of type I is

$$M * Q = Q * \Lambda$$

where matrix **M** is symmetric

$$M = L^{-1} * A * L^{-H}$$

The residual is the result of **syevj** on matrix **M**, not **A**.

The user has to provide working space which is pointed by input parameter **work**. The input parameter **lwork** is the size of the working space, and it is returned by **sygvj\_bufferSize()**.

If output parameter **info** = **-i** (less than zero), the **i-th** parameter is wrong (not counting handle). If **info** = **i** ( $i > 0$  and  $i \leq n$ ), **B** is not positive definite, the factorization of **B** could not be completed and no eigenvalues or eigenvectors were computed. If **info** = **n+1**, **syevj** does not converge under given tolerance and maximum sweeps. In this case, the eigenvalues and eigenvectors are still computed because non-convergence comes from improper tolerance of maximum sweeps.

if **jobz** = CUSOLVER\_EIG\_MODE\_VECTOR, **A** contains the orthogonal eigenvectors **v**.

Appendix F.4 provides a simple example of **sygvj**.

#### API of sygvj

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverDN library context.
<b>itype</b>	host	input	Specifies the problem type to be solved: <b>itype</b> =CUSOLVER_EIG_TYPE_1: $A * x = (\text{lambda}) * B * x$ . <b>itype</b> =CUSOLVER_EIG_TYPE_2: $A * B * x = (\text{lambda}) * x$ . <b>itype</b> =CUSOLVER_EIG_TYPE_3: $B * A * x = (\text{lambda}) * x$ .
<b>jobz</b>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <b>jobz</b> = CUSOLVER_EIG_MODE_NOVECTOR : Compute eigenvalues only; <b>jobz</b> = CUSOLVER_EIG_MODE_VECTOR : Compute eigenvalues and eigenvectors.
<b>uplo</b>	host	input	specifies which part of <b>A</b> and <b>B</b> are stored. <b>uplo</b> = CUBLAS_FILL_MODE_LOWER: Lower triangle of <b>A</b> and <b>B</b> are stored. <b>uplo</b> = CUBLAS_FILL_MODE_UPPER: Upper triangle of <b>A</b> and <b>B</b> are stored.

<b>n</b>	<b>host</b>	<b>input</b>	number of rows (or columns) of matrix <b>A</b> and <b>B</b> .
<b>A</b>	<b>device</b>	<b>in/out</b>	<type> array of dimension <code>lda * n</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <code>n</code> -by- <code>n</code> upper triangular part of <b>A</b> contains the upper triangular part of the matrix <b>A</b> . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <code>n</code> -by- <code>n</code> lower triangular part of <b>A</b> contains the lower triangular part of the matrix <b>A</b> . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>info = 0</code> , <b>A</b> contains the orthonormal eigenvectors of the matrix <b>A</b> . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of <b>A</b> are destroyed.
<b>lda</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>A</b> . <code>lda</code> is not less than <code>max(1, n)</code> .
<b>B</b>	<b>device</b>	<b>in/out</b>	<type> array of dimension <code>ldb * n</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <code>n</code> -by- <code>n</code> upper triangular part of <b>B</b> contains the upper triangular part of the matrix <b>B</b> . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <code>n</code> -by- <code>n</code> lower triangular part of <b>B</b> contains the lower triangular part of the matrix <b>B</b> . On exit, if <code>info</code> is less than <code>n</code> , <b>B</b> is overwritten by triangular factor <b>U</b> or <b>L</b> from the Cholesky factorization of <b>B</b> .
<b>ldb</b>	<b>host</b>	<b>input</b>	leading dimension of two-dimensional array used to store matrix <b>B</b> . <code>ldb</code> is not less than <code>max(1, n)</code> .
<b>W</b>	<b>device</b>	<b>output</b>	a real array of dimension <code>n</code> . The eigenvalue values of <b>A</b> , sorted so that <code>W(i) &gt;= W(i+1)</code> .
<b>work</b>	<b>device</b>	<b>in/out</b>	working space, <type> array of size <code>lwork</code> .
<b>lwork</b>	<b>host</b>	<b>input</b>	size of <b>work</b> , returned by <code>sygvj_bufferSize</code> .
<b>info</b>	<b>device</b>	<b>output</b>	if <code>info = 0</code> , the operation is successful. if <code>info = -i</code> , the <code>i</code> -th parameter is wrong (not counting handle). if <code>info = i (&gt; 0)</code> , <code>info</code> indicates either <b>B</b> is not positive definite or <code>syevj</code> (called by <code>sygvj</code> ) does not converge.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.

<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $n < 0$ , or $lda < \max(1, n)$ , or $ldb < \max(1, n)$ , or <code>itype</code> is not 1, 2 or 3, or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.4.3.19. cusolverDn<t>syevjBatched()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```

cusolverStatus_t
cusolverDnSsyevjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const float *A,
    int lda,
    const float *W,
    int *lwork,
    syevjInfo_t params,
    int batchSize
);

cusolverStatus_t
cusolverDnDsyevjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const double *A,
    int lda,
    const double *W,
    int *lwork,
    syevjInfo_t params,
    int batchSize
);

cusolverStatus_t
cusolverDnCsyevjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuComplex *A,
    int lda,
    const float *W,
    int *lwork,
    syevjInfo_t params,
    int batchSize
);

cusolverStatus_t
cusolverDnZsyevjBatched_bufferSize(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    const cuDoubleComplex *A,
    int lda,
    const double *W,
    int *lwork,
    syevjInfo_t params,
    int batchSize
);

```

The S and D data types are real valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnSsyevjBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    float *A,
    int lda,
    float *W,
    float *work,
    int lwork,
    int *info,
    syevjInfo_t params,
    int batchSize
);
```

```
cusolverStatus_t
cusolverDnDsyevjBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    double *A,
    int lda,
    double *W,
    double *work,
    int lwork,
    int *info,
    syevjInfo_t params,
    int batchSize
);
```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverDnCheevjBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuComplex *A,
    int lda,
    float *W,
    cuComplex *work,
    int lwork,
    int *info,
    syevjInfo_t params,
    int batchSize
);

cusolverStatus_t
cusolverDnZheevjBatched(
    cusolverDnHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int n,
    cuDoubleComplex *A,
    int lda,
    double *W,
    cuDoubleComplex *work,
    int lwork,
    int *info,
    syevjInfo_t params,
    int batchSize
);
```

This function computes eigenvalues and eigenvectors of a sequence of symmetric (Hermitian)  $n \times n$  matrices

$$A_j * Q_j = Q_j * \Lambda_j$$

where  $\Lambda_j$  is a real  $n \times n$  diagonal matrix.  $Q_j$  is an  $n \times n$  unitary matrix. The diagonal elements of  $\Lambda_j$  are the eigenvalues of  $A_j$  in either ascending order or non-sorting order.

**syevjBatched** performs **syevj** on each matrix. It requires that all matrices are of the same size  $n$  and are packed in contiguous way,

$$A = (A_0 \ A_1 \ \dots)$$

Each matrix is column-major with leading dimension **lda**, so the formula for random access is  $A_k(i,j) = A[i + lda*j + lda*n*k]$ .

The parameter **w** also contains eigenvalues of each matrix in contiguous way,

$$W = (W_0 \ W_1 \ \dots)$$

The formula for random access of **w** is  $W_k(j) = W[j + n*k]$ .

Except for tolerance and maximum sweeps, `syevjBatched` can either sort the eigenvalues in ascending order (default) or chose as-is (without sorting) by the function `cusolverDnXsyevjSetSortEig`. If the user packs several tiny matrices into diagonal blocks of one matrix, non-sorting option can separate spectrum of those tiny matrices.

`syevjBatched` cannot report residual and executed sweeps by function `cusolverDnXsyevjGetResidual` and `cusolverDnXsyevjGetSweeps`. Any call of the above two returns `CUSOLVER_STATUS_NOT_SUPPORTED`. The user needs to compute residual explicitly.

The user has to provide working space pointed by input parameter `work`. The input parameter `lwork` is the size of the working space, and it is returned by `syevjBatched_bufferSize()`.

The output parameter `info` is an integer array of size `batchSize`. If the function returns `CUSOLVER_STATUS_INVALID_VALUE`, the first element `info[0] = -i` (less than zero) indicates `i`-th parameter is wrong (not counting handle). Otherwise, if `info[i] = n+1`, `syevjBatched` does not converge on `i`-th matrix under given tolerance and maximum sweeps.

if `jobz = CUSOLVER_EIG_MODE_VECTOR`,  $A_j$  contains the orthonormal eigenvectors  $V_j$ .

Appendix F.5 provides a simple example of `syevjBatched`.

#### API of `syevjBatched`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverDN library context.
<code>jobz</code>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute eigenvalues only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute eigenvalues and eigenvectors.
<code>uplo</code>	host	input	specifies which part of $A_j$ is stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of $A_j$ is stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of $A_j$ is stored.
<code>n</code>	host	input	number of rows (or columns) of matrix each $A_j$ .
<code>A</code>	device	in/out	<type> array of dimension <code>lda * n * batchSize</code> with <code>lda</code> is not less than <code>max(1, n)</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <code>n</code> -by- <code>n</code> upper triangular part of $A_j$ contains the upper triangular part of the matrix $A_j$ . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading <code>n</code> -by- <code>n</code> lower triangular part of $A_j$ contains the lower triangular part of the matrix $A_j$ . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>info[j] = 0</code> , $A_j$ contains the orthonormal

			eigenvectors of the matrix $A_j$ . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of $A_j$ are destroyed.
<code>lda</code>	<code>host</code>	<code>input</code>	leading dimension of two-dimensional array used to store matrix $A_j$ .
<code>W</code>	<code>device</code>	<code>output</code>	a real array of dimension $n \times \text{batchSize}$ . It stores the eigenvalues of $A_j$ in ascending order or non-sorting order.
<code>work</code>	<code>device</code>	<code>in/out</code>	<type> array of size <code>lwork</code> , workspace.
<code>lwork</code>	<code>host</code>	<code>input</code>	size of <code>work</code> , returned by <code>syevjBatched_bufferSize</code> .
<code>info</code>	<code>device</code>	<code>output</code>	an integer array of dimension <code>batchSize</code> . If <code>CUSOLVER_STATUS_INVALID_VALUE</code> is returned, <code>info[0] = -i</code> (less than zero) indicates <i>i</i> -th parameter is wrong (not counting handle). Otherwise, if <code>info[i] = 0</code> , the operation is successful. if <code>info[i] = n+1</code> , <code>syevjBatched</code> dose not converge on <i>i</i> -th matrix under given tolerance and maximum sweeps.
<code>params</code>	<code>host</code>	<code>in/out</code>	structure filled with parameters of Jacobi algorithm.
<code>batchSize</code>	<code>host</code>	<code>input</code>	number of matrices.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $n < 0$ , or $lda < \max(1, n)$ , or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> or <code>CUBLAS_FILL_MODE_UPPER</code> ), or <code>batchSize &lt; 0</code> .
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

## 2.5. cuSolverSP: sparse LAPACK Function Reference

This chapter describes the API of cuSolverSP, which provides a subset of LAPACK funtions for sparse matrices in CSR or CSC format.

### 2.5.1. Helper Function Reference

### 2.5.1.1. cusolverSpCreate()

```
cusolverStatus_t
cusolverSpCreate(cusolverSpHandle_t *handle)
```

This function initializes the cuSolverSP library and creates a handle on the cuSolver context. It must be called before any other cuSolverSP API function is invoked. It allocates hardware resources necessary for accessing the GPU.

#### Output

<b>handle</b>	the pointer to the handle to the cuSolverSP context.
---------------	------------------------------------------------------

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the initialization succeeded.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the CUDA Runtime initialization failed.
<b>CUSOLVER_STATUS_ALLOC_FAILED</b>	the resources could not be allocated.
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.

### 2.5.1.2. cusolverSpDestroy()

```
cusolverStatus_t
cusolverSpDestroy(cusolverSpHandle_t handle)
```

This function releases CPU-side resources used by the cuSolverSP library.

#### Input

<b>handle</b>	the handle to the cuSolverSP context.
---------------	---------------------------------------

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the shutdown succeeded.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.

### 2.5.1.3. cusolverSpSetStream()

```
cusolverStatus_t
cusolverSpSetStream(cusolverSpHandle_t handle, cudaStream_t streamId)
```

This function sets the stream to be used by the cuSolverSP library to execute its routines.

#### Input

<b>handle</b>	the handle to the cuSolverSP context.
<b>streamId</b>	the stream to be used by the library.

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the stream was set successfully.
--------------------------------	----------------------------------

CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
---------------------------------	----------------------------------

### 2.5.1.4. cusolverSpXcsrissym()

```
cusolverStatus_t
cusolverSpXcsrissymHost(cusolverSpHandle_t handle,
                        int m,
                        int nnzA,
                        const cusparseMatDescr_t descrA,
                        const int *csrRowPtrA,
                        const int *csrEndPtrA,
                        const int *csrColIndA,
                        int *issym);
```

This function checks if **A** has symmetric pattern or not. The output parameter **issym** reports 1 if **A** is symmetric; otherwise, it reports 0.

The matrix **A** is an  $m \times m$  sparse matrix that is defined in CSR storage format by the four arrays **csrValA**, **csrRowPtrA**, **csrEndPtrA** and **csrColIndA**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**.

The **csrslsvlu** and **csrslsvqr** do not accept non-general matrix. the user has to extend the matrix into its missing upper/lower part, otherwise the result is not expected. The user can use **csrissym** to check if the matrix has symmetric pattern or not.

Remark 1: only CPU path is provided.

Remark 2: the user has to check returned status to get valid information. The function converts **A** to CSC format and compare CSR and CSC format. If the CSC failed because of insufficient resources, **issym** is undefined, and this state can only be detected by the return status code.

#### Input

parameter	MemorySpace	description
<b>handle</b>	host	handle to the cuSolverSP library context.
<b>m</b>	host	number of rows and columns of matrix <b>A</b> .
<b>nnzA</b>	host	number of nonzeros of matrix <b>A</b> . It is the size of <b>csrValA</b> and <b>csrColIndA</b> .
<b>descrA</b>	host	the descriptor of matrix <b>A</b> . The supported matrix type is <b>CUSPARSE_MATRIX_TYPE_GENERAL</b> . Also, the supported index bases are <b>CUSPARSE_INDEX_BASE_ZERO</b> and <b>CUSPARSE_INDEX_BASE_ONE</b> .
<b>csrRowPtrA</b>	host	integer array of <b>m</b> elements that contains the start of every row.
<b>csrEndPtrA</b>	host	integer array of <b>m</b> elements that contains the end of the last row plus one.
<b>csrColIndA</b>	host	integer array of <b>nnzA</b> column indices of the nonzero elements of matrix <b>A</b> .

## Output

parameter	MemorySpace	description
issym	host	1 if $A$ is symmetric; 0 otherwise.

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_ALLOC_FAILED	the resources could not be allocated.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $m, nnzA \leq 0$ ), base index is not 0 or 1.
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.
CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED	the matrix type is not supported.

## 2.5.2. High Level Function Reference

This section describes high level API of cuSolverSP, including linear solver, least-square solver and eigenvalue solver. The high-level API is designed for ease-of-use, so it allocates any required memory under the hood automatically. If the host or GPU system memory is not enough, an error is returned.

### 2.5.2.1. cusolverSp<t>csrslsvlu()

```

cusolverStatus_t
cusolverSpScsrslsvlu[Host](cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const float *b,
    float tol,
    int reorder,
    float *x,
    int *singularity);

cusolverStatus_t
cusolverSpDcsrslsvlu[Host](cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const double *b,
    double tol,
    int reorder,
    double *x,
    int *singularity);

cusolverStatus_t
cusolverSpCcsrslsvlu[Host](cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const cuComplex *b,
    float tol,
    int reorder,
    cuComplex *x,
    int *singularity);

cusolverStatus_t
cusolverSpZcsrslsvlu[Host](cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const cuDoubleComplex *b,
    double tol,
    int reorder,
    cuDoubleComplex *x,
    int *singularity);

```

This function solves the linear system

$$A*x = b$$

**A** is an  $n \times n$  sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**. **b** is the right-hand-side vector of size **n**, and **x** is the solution vector of size **n**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. If matrix **A** is symmetric/Hermitian and only lower/upper part is used or meaningful, the user has to extend the matrix into its missing upper/lower part, otherwise the result would be wrong.

The linear system is solved by sparse LU with partial pivoting,

$$P*A = L*U$$

**cusolver** library provides three reordering schemes, **symrcm**, **symamd**, and **csrmetisnd** to reduce zero fill-in which dramatically affects the performance of LU factorization. The input parameter **reorder** can enable **symrcm** (**symamd** or **csrmetisnd**) if **reorder** is 1 (2, or 3), otherwise, no reordering is performed.

If **reorder** is nonzero, **csrslvlu** does

$$P*A*Q^T = L*U$$

where  $Q = \text{symrcm}(A + A^T)$ .

If **A** is singular under given tolerance (**max(tol, 0)**), then some diagonal elements of **U** is zero, i.e.

$$|U(j,j)| < \text{tol for some } j$$

The output parameter **singularity** is the smallest index of such **j**. If **A** is non-singular, **singularity** is -1. The index is base-0, independent of base index of **A**. For example, if 2nd column of **A** is the same as first column, then **A** is singular and **singularity** = 1 which means  $U(1,1) \approx 0$ .

Remark 1: **csrslvlu** performs traditional LU with partial pivoting, the pivot of k-th column is determined dynamically based on the k-th column of intermediate matrix. **csrslvlu** follows Gilbert and Peierls's algorithm [4] which uses depth-first-search and topological ordering to solve triangular system (Davis also describes this algorithm in detail in his book [1]). since cuda 10.1, **csrslvlu** will incrementally reallocate the memory to store **L** and **U**. This feature can avoid over-estimate size from QR factorization. In some cases, zero fill-in of QR can be order of magnitude higher than LU.

Remark 2: only CPU (Host) path is provided.

### Input

parameter	cusolverSp MemSpace	*Host MemSpace	description
handle	host	host	handle to the cuSolverSP library context.
n	host	host	number of rows and columns of matrix <b>A</b> .

<code>nnzA</code>	<code>host</code>	<code>host</code>	number of nonzeros of matrix <b>A</b> .
<code>descrA</code>	<code>host</code>	<code>host</code>	the descriptor of matrix <b>A</b> . The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> . Also, the supported index bases are <code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code> .
<code>csrValA</code>	<code>device</code>	<code>host</code>	<type> array of <code>nnzA (= csrRowPtrA(n) - csrRowPtrA(0) )</code> nonzero elements of matrix <b>A</b> .
<code>csrRowPtrA</code>	<code>device</code>	<code>host</code>	integer array of <code>n + 1</code> elements that contains the start of every row and the end of the last row plus one.
<code>csrColIndA</code>	<code>device</code>	<code>host</code>	integer array of <code>nnzA (= csrRowPtrA(n) - csrRowPtrA(0) )</code> column indices of the nonzero elements of matrix <b>A</b> .
<code>b</code>	<code>device</code>	<code>host</code>	right hand side vector of size <code>n</code> .
<code>tol</code>	<code>host</code>	<code>host</code>	tolerance to decide if singular or not.
<code>reorder</code>	<code>host</code>	<code>host</code>	no ordering if <code>reorder=0</code> . Otherwise, <code>symrcm</code> , <code>symamd</code> , or <code>csrmetisnd</code> is used to reduce zero fill-in.

## Output

<code>parameter</code>	<code>cusolverSp MemSpace</code>	<code>*Host MemSpace</code>	<code>description</code>
<code>x</code>	<code>device</code>	<code>host</code>	solution vector of size <code>n</code> , $x = \text{inv}(A)*b$ .
<code>singularity</code>	<code>host</code>	<code>host</code>	-1 if <b>A</b> is invertible. Otherwise, first index <code>j</code> such that $\sigma(j, j) \approx 0$

## Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n, nnzA &lt;= 0</code> , base index is not 0 or 1, <code>reorder</code> is not 0, 1, 2, 3)
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.
<code>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code>	the matrix type is not supported.

### 2.5.2.2. cusolverSp<t>csrslsvqr()

```

cusolverStatus_t
cusolverSpScsrslsvqr[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const float *b,
    float tol,
    int reorder,
    float *x,
    int *singularity);

cusolverStatus_t
cusolverSpDcsrslsvqr[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const double *b,
    double tol,
    int reorder,
    double *x,
    int *singularity);

cusolverStatus_t
cusolverSpCcsrslsvqr[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const cuComplex *b,
    float tol,
    int reorder,
    cuComplex *x,
    int *singularity);

cusolverStatus_t
cusolverSpZcsrslsvqr[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const cuDoubleComplex *b,
    double tol,
    int reorder,
    cuDoubleComplex *x,
    int *singularity);

```

This function solves the linear system

$$A * x = b$$

**A** is an  $m \times m$  sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**. **b** is the right-hand-side vector of size **m**, and **x** is the solution vector of size **m**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. If matrix **A** is symmetric/Hermitian and only lower/upper part is used or meaningful, the user has to extend the matrix into its missing upper/lower part, otherwise the result would be wrong.

The linear system is solved by sparse QR factorization,

$$A = Q * R$$

If **A** is singular under given tolerance (**max(tol, 0)**), then some diagonal elements of **R** is zero, i.e.

$$|R(j,j)| < \text{tol for some } j$$

The output parameter **singularity** is the smallest index of such **j**. If **A** is non-singular, **singularity** is -1. The **singularity** is base-0, independent of base index of **A**. For example, if 2nd column of **A** is the same as first column, then **A** is singular and **singularity** = 1 which means  $R(1,1) \approx 0$ .

**cusolver** library provides three reordering schemes, **symrcm**, **symamd**, and **csrmetisnd** to reduce zero fill-in which dramatically affects the performance of QR factorization. The input parameter **reorder** can enable **symrcm** (**symamd** or **csrmetisnd**) if **reorder** is 1 (2, or 3), otherwise, no reordering is performed.

### Input

parameter	cusolverSp MemSpace	*Host MemSpace	description
<b>handle</b>	host	host	handle to the cuSolverSP library context.
<b>m</b>	host	host	number of rows and columns of matrix <b>A</b> .
<b>nnz</b>	host	host	number of nonzeros of matrix <b>A</b> .
<b>descrA</b>	host	host	the descriptor of matrix <b>A</b> . The supported matrix type is <b>CUSPARSE_MATRIX_TYPE_GENERAL</b> . Also, the supported index bases are <b>CUSPARSE_INDEX_BASE_ZERO</b> and <b>CUSPARSE_INDEX_BASE_ONE</b> .
<b>csrValA</b>	device	host	<type> array of <b>nnz</b> (= <b>csrRowPtrA</b> ( <b>m</b> ) - <b>csrRowPtrA</b> (0)) nonzero elements of matrix <b>A</b> .
<b>csrRowPtrA</b>	device	host	integer array of <b>m + 1</b> elements that contains the start of every row and the end of the last row plus one.

<code>csrColIndA</code>	<code>device</code>	<code>host</code>	integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> column indices of the nonzero elements of matrix <code>A</code> .
<code>b</code>	<code>device</code>	<code>host</code>	right hand side vector of size <code>m</code> .
<code>tol</code>	<code>host</code>	<code>host</code>	tolerance to decide if singular or not.
<code>reorder</code>	<code>host</code>	<code>host</code>	no ordering if <code>reorder=0</code> . Otherwise, <code>symrcm</code> , <code>symamd</code> , or <code>csrmetisnd</code> is used to reduce zero fill-in.

## Output

parameter	cusolverSp MemSpace	*Host MemSpace	description
<code>x</code>	<code>device</code>	<code>host</code>	solution vector of size <code>m</code> , $x = \text{inv}(A)*b$ .
<code>singularity</code>	<code>host</code>	<code>host</code>	-1 if <code>A</code> is invertible. Otherwise, first index <code>j</code> such that $R(j, j) \approx 0$

## Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>m</code> , <code>nnz</code> $\leq 0$ , base index is not 0 or 1, <code>reorder</code> is not 0,1,2,3)
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.
<code>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code>	the matrix type is not supported.

### 2.5.2.3. cusolverSp<t>csrsvchol()

```
cusolverStatus_t
cusolverSpScsrsvchol[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const float *csrVal,
    const int *csrRowPtr,
    const int *csrColInd,
    const float *b,
    float tol,
    int reorder,
    float *x,
    int *singularity);
```

```
cusolverStatus_t
cusolverSpDcsrsvchol[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const double *csrVal,
    const int *csrRowPtr,
    const int *csrColInd,
    const double *b,
    double tol,
    int reorder,
    double *x,
    int *singularity);
```

```
cusolverStatus_t
cusolverSpCcsrsvchol[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrVal,
    const int *csrRowPtr,
    const int *csrColInd,
    const cuComplex *b,
    float tol,
    int reorder,
    cuComplex *x,
    int *singularity);
```

```
cusolverStatus_t
cusolverSpZcsrsvchol[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrVal,
    const int *csrRowPtr,
    const int *csrColInd,
    const cuDoubleComplex *b,
    double tol,
    int reorder,
    cuDoubleComplex *x,
    int *singularity);
```

This function solves the linear system

$$A*x = b$$

**A** is an  $m \times m$  symmetric positive definite sparse matrix that is defined in CSR storage format by the three arrays `csrValA`, `csrRowPtrA`, and `csrColIndA`. **b** is the right-hand-side vector of size **m**, and **x** is the solution vector of size **m**.

The supported matrix type is `CUSPARSE_MATRIX_TYPE_GENERAL` and upper triangular part of **A** is ignored (if parameter `reorder` is zero). In other words, suppose input matrix **A** is decomposed as  $A = L + D + U$ , where **L** is lower triangular, **D** is diagonal and **U** is upper triangular. The function would ignore **U** and regard **A** as a symmetric matrix with the formula  $A = L + D + L^H$ . If parameter `reorder` is nonzero, the user has to extend **A** to a full matrix, otherwise the solution would be wrong.

The linear system is solved by sparse Cholesky factorization,

$$A = G * G^H$$

where **G** is the Cholesky factor, a lower triangular matrix.

The output parameter `singularity` has two meanings:

- ▶ If **A** is not positive definite, there exists some integer **k** such that **A**(0:k, 0:k) is not positive definite. `singularity` is the minimum of such **k**.
- ▶ If **A** is positive definite but near singular under tolerance (`max(tol, 0)`), i.e. there exists some integer **k** such that  $G(k,k) \leq \text{tol}$ . `singularity` is the minimum of such **k**.

`singularity` is base-0. If **A** is positive definite and not near singular under tolerance, `singularity` is -1. If the user wants to know if **A** is positive definite or not, `tol=0` is enough.

`cusolver` library provides three reordering schemes, `symrcm`, `symamd`, and `csrmetisnd` to reduce zero fill-in which dramatically affects the performance of Cholesky factorization. The input parameter `reorder` can enable `symrcm` (`symamd` or `csrmetisnd`) if `reorder` is 1 (2, or 3), otherwise, no reordering is performed.

Remark 1: the function works for in-place (**x** and **b** point to the same memory block) and out-of-place.

Remark 2: the function only works on 32-bit index, if matrix **G** has large zero fill-in such that number of nonzeros is bigger than  $2^{31}$ , then `CUSOLVER_STATUS_ALLOC_FAILED` is returned.

### Input

parameter	cusolverSp MemSpace	*Host MemSpace	description
<code>handle</code>	host	host	handle to the cuSolverSP library context.
<code>m</code>	host	host	number of rows and columns of matrix <b>A</b> .
<code>nnz</code>	host	host	number of nonzeros of matrix <b>A</b> .

<code>descrA</code>	<code>host</code>	<code>host</code>	the descriptor of matrix <b>A</b> . The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> . Also, the supported index bases are <code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code> .
<code>csrValA</code>	<code>device</code>	<code>host</code>	<type> array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0) )</code> nonzero elements of matrix <b>A</b> .
<code>csrRowPtrA</code>	<code>device</code>	<code>host</code>	integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.
<code>csrColIndA</code>	<code>device</code>	<code>host</code>	integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0) )</code> column indices of the nonzero elements of matrix <b>A</b> .
<code>b</code>	<code>device</code>	<code>host</code>	right hand side vector of size <code>m</code> .
<code>tol</code>	<code>host</code>	<code>host</code>	tolerance to decide singularity.
<code>reorder</code>	<code>host</code>	<code>host</code>	no ordering if <code>reorder=0</code> . Otherwise, <code>symrcm</code> , <code>symamd</code> , or <code>csrmetisnd</code> is used to reduce zero fill-in.

## Output

parameter	<code>cusolverSp MemSpace</code>	<code>*Host MemSpace</code>	description
<code>x</code>	<code>device</code>	<code>host</code>	solution vector of size <code>m</code> , <code>x = inv(A)*b</code> .
<code>singularity</code>	<code>host</code>	<code>host</code>	-1 if <b>A</b> is symmetric positive definite.

## Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>m, nnz &lt;= 0</code> , base index is not 0 or 1, <code>reorder</code> is not 0,1,2,3)
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.
<code>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code>	the matrix type is not supported.

### 2.5.2.4. cusolverSp<t>csrIlsqvqr()

The S and D data types are real valued single and double precision, respectively.

```

cusolverStatus_t
cusolverSpScsrIlsqvqr[Host](cusolverSpHandle_t handle,
                             int m,
                             int n,
                             int nnz,
                             const cusparseMatDescr_t descrA,
                             const float *csrValA,
                             const int *csrRowPtrA,
                             const int *csrColIndA,
                             const float *b,
                             float tol,
                             int *rankA,
                             float *x,
                             int *p,
                             float *min_norm);

cusolverStatus_t
cusolverSpDcsrIlsqvqr[Host](cusolverSpHandle_t handle,
                             int m,
                             int n,
                             int nnz,
                             const cusparseMatDescr_t descrA,
                             const double *csrValA,
                             const int *csrRowPtrA,
                             const int *csrColIndA,
                             const double *b,
                             double tol,
                             int *rankA,
                             double *x,
                             int *p,
                             double *min_norm);

```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverSpCcsr1sqvqr[Host](cusolverSpHandle_t handle,
    int m,
    int n,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const cuComplex *b,
    float tol,
    int *rankA,
    cuComplex *x,
    int *p,
    float *min_norm);

cusolverStatus_t
cusolverSpZcsr1sqvqr[Host](cusolverSpHandle_t handle,
    int m,
    int n,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const cuDoubleComplex *b,
    double tol,
    int *rankA,
    cuDoubleComplex *x,
    int *p,
    double *min_norm);
```

This function solves the following least-square problem

$$x = \operatorname{argmin} \|A * z - b\|$$

**A** is an  $m \times n$  sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**. **b** is the right-hand-side vector of size **m**, and **x** is the least-square solution vector of size **n**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. If **A** is square, symmetric/Hermitian and only lower/upper part is used or meaningful, the user has to extend the matrix into its missing upper/lower part, otherwise the result is wrong.

This function only works if **m** is greater or equal to **n**, in other words, **A** is a tall matrix.

The least-square problem is solved by sparse QR factorization with column pivoting,

$$A * P^T = Q * R$$

If **A** is of full rank (i.e. all columns of **A** are linear independent), then matrix **P** is an identity. Suppose rank of **A** is **k**, less than **n**, the permutation matrix **P** reorders columns of **A** in the following sense:

$$A * P^T = (A_1 \ A_2) = (Q_1 \ Q_2) \begin{pmatrix} R_{11} & R_{12} \\ & R_{22} \end{pmatrix}$$

where  $R_{11}$  and  $\mathbf{A}$  have the same rank, but  $R_{22}$  is almost zero, i.e. every column of  $A_2$  is linear combination of  $A_1$ .

The input parameter **tol** decides numerical rank. The absolute value of every entry in  $R_{22}$  is less than or equal to **tolerance=max(tol, 0)**.

The output parameter **rankA** denotes numerical rank of  $\mathbf{A}$ .

Suppose  $y = P^*x$  and  $c = Q^H*b$ , the least square problem can be reformed by

$$\min \|A^*x - b\| = \min \|R^*y - c\|$$

or in matrix form

$$\begin{pmatrix} R_{11} & R_{12} \\ & R_{22} \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}$$

The output parameter **min\_norm** is  $\|c_2\|$ , which is minimum value of least-square problem.

If  $\mathbf{A}$  is not of full rank, above equation does not have a unique solution. The least-square problem is equivalent to

$$\begin{aligned} & \min \|y\| \\ & \text{subject to } R_{11}^*y_1 + R_{12}^*y_2 = c_1 \end{aligned}$$

Or equivalently another least-square problem

$$\min \left\| \begin{pmatrix} R_{11} \setminus R_{12} \\ I \end{pmatrix}^* y_2 - \begin{pmatrix} R_{11} \setminus c_1 \\ O \end{pmatrix} \right\|$$

The output parameter  $\mathbf{x}$  is  $P^T*y$ , the solution of least-square problem.

The output parameter  $\mathbf{p}$  is a vector of size  $\mathbf{n}$ . It corresponds to a permutation matrix  $\mathbf{P}$ .  $\mathbf{p}(i)=j$  means  $(\mathbf{P}*\mathbf{x})(i) = \mathbf{x}(j)$ . If  $\mathbf{A}$  is of full rank,  $\mathbf{p}=\mathbf{0}:\mathbf{n}-1$ .

Remark 1:  $\mathbf{p}$  is always base 0, independent of base index of  $\mathbf{A}$ .

Remark 2: only CPU (Host) path is provided.

### Input

parameter	cusolverSp MemSpace	*Host MemSpace	description
handle	host	host	handle to the cuSolver library context.
m	host	host	number of rows of matrix $\mathbf{A}$ .
n	host	host	number of columns of matrix $\mathbf{A}$ .
nnz	host	host	number of nonzeros of matrix $\mathbf{A}$ .
descrA	host	host	the descriptor of matrix $\mathbf{A}$ . The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> . Also, the supported index bases are

			CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.
csrValA	device	host	<type> array of nnz (= csrRowPtrA(m) - csrRowPtrA(0) ) nonzero elements of matrix A.
csrRowPtrA	device	host	integer array of m + 1 elements that contains the start of every row and the end of the last row plus one.
csrColIndA	device	host	integer array of nnz (= csrRowPtrA(m) - csrRowPtrA(0) ) column indices of the nonzero elements of matrix A.
b	device	host	right hand side vector of size m.
tol	host	host	tolerance to decide rank of A.

## Output

parameter	cusolverSp MemSpace	*Host MemSpace	description
rankA	host	host	numerical rank of A.
x	device	host	solution vector of size n, $x = \text{pinv}(A) * b$ .
p	device	host	a vector of size n, which represents the permutation matrix P satisfying $A * P^T = Q * R$ .
min_norm	host	host	$\ A * x - b\ $ , $x = \text{pinv}(A) * b$ .

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_ALLOC_FAILED	the resources could not be allocated.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed (m, n, nnz <= 0), base index is not 0 or 1.
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.
CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED	the matrix type is not supported.

### 2.5.2.5. cusolverSp<t>csreigvsi()

The S and D data types are real valued single and double precision, respectively.

```

cusolverStatus_t
cusolverSpScsreigvsi[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    float mu0,
    const float *x0,
    int maxite,
    float tol,
    float *mu,
    float *x);

cusolverStatus_t
cusolverSpDcsreigvsi[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    double mu0,
    const double *x0,
    int maxite,
    double tol,
    double *mu,
    double *x);

```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverSpCcsreigvsi[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    cuComplex mu0,
    const cuComplex *x0,
    int maxite,
    float tol,
    cuComplex *mu,
    cuComplex *x);

cusolverStatus_t
cusolverSpZcsreigvsi(cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    cuDoubleComplex mu0,
    const cuDoubleComplex *x0,
    int maxite,
    double tol,
    cuDoubleComplex *mu,
    cuDoubleComplex *x);
```

This function solves the simple eigenvalue problem  $A * x = \lambda * x$  by shift-inverse method.

**A** is an  $m \times m$  sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**. The output parameter **x** is the approximated eigenvector of size **m**,

The following shift-inverse method corrects eigenpair step-by-step until convergence.

It accepts several parameters:

**mu0** is an initial guess of eigenvalue. The shift-inverse method will converge to the eigenvalue **mu** nearest **mu0** if **mu** is a singleton. Otherwise, the shift-inverse method may not converge.

**x0** is an initial eigenvector. If the user has no preference, just chose **x0** randomly. **x0** must be nonzero. It can be non-unit length.

**tol** is the tolerance to decide convergence. If **tol** is less than zero, it would be treated as zero.

**maxite** is maximum number of iterations. It is useful when shift-inverse method does not converge because the tolerance is too small or the desired eigenvalue is not a singleton.

### Shift-Inverse Method

```

Given a initial guess of eigenvalue  $\mu_0$  and initial vector  $x_0$ 
 $x^{(0)} = x_0$  of unit length
for j = 0 : maxite
    solve  $(A - \mu_0 * I)$ 
    normalize  $x^{(k+1)}$  to unit length
    compute approx. eigenvalue  $\mu = x^{(k+1)H} * A * x^{(k+1)} - \mu$ 
    if ||
endfor

```

The supported matrix type is `CUSPARSE_MATRIX_TYPE_GENERAL`. If **A** is symmetric/Hermitian and only lower/upper part is used or meaningful, the user has to extend the matrix into its missing upper/lower part, otherwise the result is wrong.

Remark 1: `[cu|h]solver[S|D]csreigvsi` only allows `mu0` as a real number. This works if **A** is symmetric. Otherwise, the non-real eigenvalue has a conjugate counterpart on the complex plan, and shift-inverse method would not converge to such eigenvalue even the eigenvalue is a singleton. The user has to extend **A** to complex number and call `[cu|h]solver[C|Z]csreigvsi` with `mu0` not on real axis.

Remark 2: the tolerance `tol` should not be smaller than  $|\mu_0| * \text{eps}$ , where `eps` is machine zero. Otherwise, shift-inverse may not converge because of small tolerance.

### Input

parameter	cusolverSp MemSpace	*Host MemSpace	description
<code>handle</code>	<code>host</code>	<code>host</code>	handle to the cuSolver library context.
<code>m</code>	<code>host</code>	<code>host</code>	number of rows and columns of matrix <b>A</b> .
<code>nnz</code>	<code>host</code>	<code>host</code>	number of nonzeros of matrix <b>A</b> .
<code>descrA</code>	<code>host</code>	<code>host</code>	the descriptor of matrix <b>A</b> . The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> . Also, the supported index bases are <code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code> .
<code>csrValA</code>	<code>device</code>	<code>host</code>	<type> array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> nonzero elements of matrix <b>A</b> .
<code>csrRowPtrA</code>	<code>device</code>	<code>host</code>	integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.
<code>csrColIndA</code>	<code>device</code>	<code>host</code>	integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> column indices of the nonzero elements of matrix <b>A</b> .
<code>mu0</code>	<code>host</code>	<code>host</code>	initial guess of eigenvalue.
<code>x0</code>	<code>device</code>	<code>host</code>	initial guess of eigenvector, a vector of size <code>m</code> .
<code>maxite</code>	<code>host</code>	<code>host</code>	maximum iterations in shift-inverse method.

tol	host	host	tolerance for convergence.
-----	------	------	----------------------------

## Output

parameter	cusolverSp MemSpace	*Host MemSpace	description
mu	device	host	approximated eigenvalue nearest mu0 under tolerance.
x	device	host	approximated eigenvector of size m.

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_ALLOC_FAILED	the resources could not be allocated.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $m, nnz \leq 0$ ), base index is not 0 or 1.
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.
CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED	the matrix type is not supported.

### 2.5.2.6. cusolverSp<t>csreigs()

```

cusolverStatus_t
solverspScsreigs[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    cuComplex left_bottom_corner,
    cuComplex right_upper_corner,
    int *num_eigs);

cusolverStatus_t
cusolverSpDcsrreigs[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    cuDoubleComplex left_bottom_corner,
    cuDoubleComplex right_upper_corner,
    int *num_eigs);

cusolverStatus_t
cusolverSpCcsrreigs[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    cuComplex left_bottom_corner,
    cuComplex right_upper_corner,
    int *num_eigs);

cusolverStatus_t
cusolverSpZcsrreigs[Host](cusolverSpHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    cuDoubleComplex left_bottom_corner,
    cuDoubleComplex right_upper_corner,
    int *num_eigs);

```

This function computes number of algebraic eigenvalues in a given box **B** by contour integral

$$\text{number of algebraic eigenvalues in box } B = \frac{1}{2 * \pi * \sqrt{-1}} \oint_C \frac{P'(z)}{P(z)} dz$$

where closed line **C** is boundary of the box **B** which is a rectangle specified by two points, one is left bottom corner (input parameter `left_bottom_corner`) and the other is right upper corner (input parameter `right_upper_corner`).  $P(z) = \det(A - z \cdot I)$  is the characteristic polynomial of **A**.

**A** is an  $m \times m$  sparse matrix that is defined in CSR storage format by the three arrays `csrValA`, `csrRowPtrA`, and `csrColIndA`.

The output parameter `num_eigs` is number of algebraic eigenvalues in the box **B**. This number may not be accurate due to several reasons:

1. the contour **C** is close to some eigenvalues or even passes through some eigenvalues.
2. the numerical integration is not accurate due to coarse grid size. The default resolution is 1200 grids along contour **C** uniformly.

Even though `csreigs` may not be accurate, it still can give the user some idea how many eigenvalues in a region where the resolution of disk theorem is bad. For example, standard 3-point stencil of finite difference of Laplacian operator is a tridiagonal matrix, and disk theorem would show "all eigenvalues are in the interval  $[0, 4 \cdot N^2]$ " where  $N$  is number of grids. In this case, `csreigs` is useful for any interval inside  $[0, 4 \cdot N^2]$ .

Remark 1: if **A** is symmetric in real or hermitian in complex, all eigenvalues are real. The user still needs to specify a box, not an interval. The height of the box can be much smaller than the width.

Remark 2: only CPU (Host) path is provided.

### Input

parameter	cusolverSp MemSpace	*Host MemSpace	description
<code>handle</code>	host	host	handle to the cuSolverSP library context.
<code>m</code>	host	host	number of rows and columns of matrix <b>A</b> .
<code>nnz</code>	host	host	number of nonzeros of matrix <b>A</b> .
<code>descrA</code>	host	host	the descriptor of matrix <b>A</b> . The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> . Also, the supported index bases are <code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code> .
<code>csrValA</code>	device	host	<type> array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> nonzero elements of matrix <b>A</b> .
<code>csrRowPtrA</code>	device	host	integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.
<code>csrColIndA</code>	device	host	integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> column indices of the nonzero elements of matrix <b>A</b> .
<code>left_bottom_corner</code>	host	host	left bottom corner of the box.
<code>right_upper_corner</code>	host	host	right upper corner of the box.

## Output

parameter	cusolverSp MemSpace	*Host MemSpace	description
num_eigs	host	host	number of algebraic eigenvalues in a box.

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_ALLOC_FAILED	the resources could not be allocated.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $m, nnz \leq 0$ ), base index is not 0 or 1.
CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.
CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED	the matrix type is not supported.

## 2.5.3. Low Level Function Reference

This section describes low level API of cuSolverSP, including symrcm and batched QR.

### 2.5.3.1. cusolverSpXcsrsmrcm()

```
cusolverStatus_t
cusolverSpXcsrsmrcmHost(cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int *p);
```

This function implements Symmetric Reverse Cuthill-McKee permutation. It returns a permutation vector  $\mathbf{p}$  such that  $\mathbf{A}(\mathbf{p}, \mathbf{p})$  would concentrate nonzeros to diagonal. This is equivalent to **symrcm** in MATLAB, however the result may not be the same because of different heuristics in the pseudoperipheral finder. The **cuSolverSP** library implements **symrcm** based on the following two papers:

E. Chuthill and J. McKee, reducing the bandwidth of sparse symmetric matrices, ACM '69 Proceedings of the 1969 24th national conference, Pages 157-172

Alan George, Joseph W. H. Liu, An Implementation of a Pseudoperipheral Node Finder, ACM Transactions on Mathematical Software (TOMS) Volume 5 Issue 3, Sept. 1979, Pages 284-295

The output parameter  $\mathbf{p}$  is an integer array of  $n$  elements. It represents a permutation array and it indexed using the base-0 convention. The permutation array  $\mathbf{p}$  corresponds to a permutation matrix  $\mathbf{P}$ , and satisfies the following relation:

$$A(p,p) = P * A * P^T$$

**A** is an  $n \times n$  sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. Internally **rcm** works on  $A + A^T$ , the user does not need to extend the matrix if the matrix is not symmetric.

Remark 1: only CPU (Host) path is provided.

### Input

parameter	*Host MemSpace	description
<b>handle</b>	host	handle to the cuSolverSP library context.
<b>n</b>	host	number of rows and columns of matrix <b>A</b> .
<b>nnzA</b>	host	number of nonzeros of matrix <b>A</b> . It is the size of <b>csrValA</b> and <b>csrColIndA</b> .
<b>descrA</b>	host	the descriptor of matrix <b>A</b> . The supported matrix type is <b>CUSPARSE_MATRIX_TYPE_GENERAL</b> . Also, the supported index bases are <b>CUSPARSE_INDEX_BASE_ZERO</b> and <b>CUSPARSE_INDEX_BASE_ONE</b> .
<b>csrRowPtrA</b>	host	integer array of $n+1$ elements that contains the start of every row and the end of the last row plus one.
<b>csrColIndA</b>	host	integer array of <b>nnzA</b> column indices of the nonzero elements of matrix <b>A</b> .

### Output

parameter	hsolver	description
<b>p</b>	host	permutation vector of size <b>n</b> .

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_ALLOC_FAILED</b>	the resources could not be allocated.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( $n, nnzA \leq 0$ ), base index is not 0 or 1.
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.
<b>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</b>	the matrix type is not supported.

### 2.5.3.2. cusolverSpXcsrSymmdq()

```
cusolverStatus_t
cusolverSpXcsrSymmdqHost(cusolverSpHandle_t handle,
                        int n,
                        int nnzA,
                        const cusparseMatDescr_t descrA,
                        const int *csrRowPtrA,
                        const int *csrColIndA,
                        int *p);
```

This function implements Symmetric Minimum Degree Algorithm based on Quotient Graph. It returns a permutation vector  $\mathbf{p}$  such that  $\mathbf{A}(\mathbf{p}, \mathbf{p})$  would have less zero fill-in during Cholesky factorization. The **cuSolverSP** library implements **symmdq** based on the following two papers:

Patrick R. Amestoy, Timothy A. Davis, Iain S. Duff, An Approximate Minimum Degree Ordering Algorithm, SIAM J. Matrix Analysis Applic. Vol 17, no 4, pp. 886-905, Dec. 1996.

Alan George, Joseph W. Liu, A Fast Implementation of the Minimum Degree Algorithm Using Quotient Graphs, ACM Transactions on Mathematical Software, Vol 6, No. 3, September 1980, page 337-358.

The output parameter  $\mathbf{p}$  is an integer array of  $n$  elements. It represents a permutation array with base-0 index. The permutation array  $\mathbf{p}$  corresponds to a permutation matrix  $\mathbf{P}$ , and satisfies the following relation:

$$A(\mathbf{p}, \mathbf{p}) = \mathbf{P} * \mathbf{A} * \mathbf{P}^T$$

$\mathbf{A}$  is an  $n \times n$  sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. Internally **mdq** works on  $A + A^T$ , the user does not need to extend the matrix if the matrix is not symmetric.

Remark 1: only CPU (Host) path is provided.

#### Input

parameter	*Host MemSpace	description
<b>handle</b>	host	handle to the cuSolverSP library context.
<b>n</b>	host	number of rows and columns of matrix <b>A</b> .
<b>nnzA</b>	host	number of nonzeros of matrix <b>A</b> . It is the size of <b>csrValA</b> and <b>csrColIndA</b> .
<b>descrA</b>	host	the descriptor of matrix <b>A</b> . The supported matrix type is <b>CUSPARSE_MATRIX_TYPE_GENERAL</b> . Also, the supported index bases are <b>CUSPARSE_INDEX_BASE_ZERO</b> and <b>CUSPARSE_INDEX_BASE_ONE</b> .

<code>csrRowPtrA</code>	<code>host</code>	integer array of <code>n+1</code> elements that contains the start of every row and the end of the last row plus one.
<code>csrColIndA</code>	<code>host</code>	integer array of <code>nnzA</code> column indices of the nonzero elements of matrix <code>A</code> .

### Output

<code>parameter</code>	<code>hsolver</code>	<code>description</code>
<code>p</code>	<code>host</code>	permutation vector of size <code>n</code> .

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n</code> , <code>nnzA</code> ≤ 0), base index is not 0 or 1.
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.
<code>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code>	the matrix type is not supported.

### 2.5.3.3. `cusolverSpXcsrSymamd()`

```
cusolverStatus_t
cusolverSpXcsrSymamdHost(cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int *p);
```

This function implements Symmetric Approximate Minimum Degree Algorithm based on Quotient Graph. It returns a permutation vector `p` such that `A(p, p)` would have less zero fill-in during Cholesky factorization. The `cuSolverSP` library implements `symamd` based on the following paper:

Patrick R. Amestoy, Timothy A. Davis, Iain S. Duff, An Approximate Minimum Degree Ordering Algorithm, SIAM J. Matrix Analysis Applic. Vol 17, no 4, pp. 886-905, Dec. 1996.

The output parameter `p` is an integer array of `n` elements. It represents a permutation array with base-0 index. The permutation array `p` corresponds to a permutation matrix `P`, and satisfies the following relation:

$$A(p,p) = P * A * P^T$$

**A** is an  $n \times n$  sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. Internally **amd** works on  $A + A^T$ , the user does not need to extend the matrix if the matrix is not symmetric.

Remark 1: only CPU (Host) path is provided.

### Input

parameter	*Host MemSpace	description
<b>handle</b>	host	handle to the cuSolverSP library context.
<b>n</b>	host	number of rows and columns of matrix <b>A</b> .
<b>nnzA</b>	host	number of nonzeros of matrix <b>A</b> . It is the size of <b>csrValA</b> and <b>csrColIndA</b> .
<b>descrA</b>	host	the descriptor of matrix <b>A</b> . The supported matrix type is <b>CUSPARSE_MATRIX_TYPE_GENERAL</b> . Also, the supported index bases are <b>CUSPARSE_INDEX_BASE_ZERO</b> and <b>CUSPARSE_INDEX_BASE_ONE</b> .
<b>csrRowPtrA</b>	host	integer array of $n+1$ elements that contains the start of every row and the end of the last row plus one.
<b>csrColIndA</b>	host	integer array of <b>nnzA</b> column indices of the nonzero elements of matrix <b>A</b> .

### Output

parameter	hsolver	description
<b>p</b>	host	permutation vector of size $n$ .

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_ALLOC_FAILED</b>	the resources could not be allocated.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( $n, nnzA \leq 0$ ), base index is not 0 or 1.
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.
<b>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</b>	the matrix type is not supported.

### 2.5.3.4. cusolverSpXcsrmetisnd()

```
cusolverStatus_t
cusolverSpXcsrmetisndHost(
    cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const int64_t *options,
    int *p);
```

This function is a wrapper of **METIS\_NodeND**. It returns a permutation vector **p** such that **A(p,p)** would have less zero fill-in during nested dissection. The **cuSolverSP** library links **libmetis\_static.a** which is 64-bit metis-5.1.0 .

The parameter **options** is the configuration of **metis**. For those who do not have experiences of **metis**, set **options = NULL** for default setting.

The output parameter **p** is an integer array of **n** elements. It represents a permutation array with base-0 index. The permutation array **p** corresponds to a permutation matrix **P**, and satisfies the following relation:

$$A(p,p) = P * A * P^T$$

**A** is an **n×n** sparse matrix that is defined in CSR storage format by the three arrays **csrValA**, **csrRowPtrA**, and **csrColIndA**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. Internally **csrmetisnd** works on  $A + A^T$ , the user does not need to extend the matrix if the matrix is not symmetric.

Remark 1: only CPU (Host) path is provided.

#### Input

parameter	*Host MemSpace	description
<b>handle</b>	host	handle to the cuSolverSP library context.
<b>n</b>	host	number of rows and columns of matrix <b>A</b> .
<b>nnzA</b>	host	number of nonzeros of matrix <b>A</b> . It is the size of <b>csrValA</b> and <b>csrColIndA</b> .
<b>descrA</b>	host	the descriptor of matrix <b>A</b> . The supported matrix type is <b>CUSPARSE_MATRIX_TYPE_GENERAL</b> . Also, the supported index bases are <b>CUSPARSE_INDEX_BASE_ZERO</b> and <b>CUSPARSE_INDEX_BASE_ONE</b> .
<b>csrRowPtrA</b>	host	integer array of <b>n+1</b> elements that contains the start of every row and the end of the last row plus one.

<code>csrColIndA</code>	<code>host</code>	integer array of <code>nnzA</code> column indices of the nonzero elements of matrix <code>A</code> .
<code>options</code>	<code>host</code>	integer array to configure <code>metis</code> .

### Output

<code>parameter</code>	<code>*Host MemSpace</code>	<code>description</code>
<code>p</code>	<code>host</code>	permutation vector of size <code>n</code> .

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>n</code> , <code>nnzA</code> $\leq$ 0), base index is not 0 or 1.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.
<code>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code>	the matrix type is not supported.

### 2.5.3.5. cusolverSpXcsrzd()

```

cusolverStatus_t
cusolverSpScsrzfdHost(
    cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int *P,
    int *numnz)

cusolverStatus_t
cusolverSpDcsrzdHost(
    cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int *P,
    int *numnz)

cusolverStatus_t
cusolverSpCcsrzdHost(
    cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int *P,
    int *numnz)

cusolverStatus_t
cusolverSpZcsrzdHost(
    cusolverSpHandle_t handle,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int *P,
    int *numnz)

```

This function implements MC21, zero-free diagonal algorithm. It returns a permutation vector  $\mathbf{p}$  such that  $\mathbf{A}(\mathbf{p}, :)$  has no zero diagonal.

$\mathbf{A}$  is an  $n \times n$  sparse matrix that is defined in CSR storage format by the three arrays `csrValA`, `csrRowPtrA`, and `csrColIndA`. The supported matrix type is `CUSPARSE_MATRIX_TYPE_GENERAL`.

The output parameter **p** is an integer array of **n** elements. It represents a permutation array with base-0 index. The permutation array **p** corresponds to a permutation matrix **P**, and satisfies the following relation:

$$A(p,:)=P*A$$

The output parameter **numnz** describes number of nonzero diagonal in permuted matrix **A(p, :)**. If **numnz** is less than **n**, matrix **A** has structural singularity.

Remark 1: only CPU (Host) path is provided.

Remark 2: this routine does not maximize diagonal value of permuted matrix. The user cannot expect this routine can make "LU without pivoting" stable.

### Input

parameter	*Host MemSpace	description
handle	host	handle to the cuSolverSP library context.
n	host	number of rows and columns of matrix <b>A</b> .
nnzA	host	number of nonzeros of matrix <b>A</b> . It is the size of <b>csrValA</b> and <b>csrColIndA</b> .
descrA	host	the descriptor of matrix <b>A</b> . The supported matrix type is <b>CUSPARSE_MATRIX_TYPE_GENERAL</b> . Also, the supported index bases are <b>CUSPARSE_INDEX_BASE_ZERO</b> and <b>CUSPARSE_INDEX_BASE_ONE</b> .
csrValA	host	<type> array of <b>nnzA</b> (= <b>csrRowPtrA</b> ( <b>m</b> ) - <b>csrRowPtrA</b> (0) ) nonzero elements of matrix <b>A</b> .
csrRowPtrA	host	integer array of <b>n+1</b> elements that contains the start of every row and the end of the last row plus one.
csrColIndA	host	integer array of <b>nnzA</b> column indices of the nonzero elements of matrix <b>A</b> .

### Output

parameter	*Host MemSpace	description
p	host	permutation vector of size <b>n</b> .
numnz	host	number of nonzeros on diagonal of permuted matrix.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_ALLOC_FAILED</b>	the resources could not be allocated.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( <b>n</b> , <b>nnzA</b> ≤ 0), base index is not 0 or 1.
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.

CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.
CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED	the matrix type is not supported.

### 2.5.3.6. cusolverSpXcsrperm()

```
cusolverStatus_t
cusolverSpXcsrperm_bufferSizeHost(cusolverSpHandle_t handle,
                                   int m,
                                   int n,
                                   int nnzA,
                                   const cusparseMatDescr_t descrA,
                                   int *csrRowPtrA,
                                   int *csrColIndA,
                                   const int *p,
                                   const int *q,
                                   size_t *bufferSizeInBytes);

cusolverStatus_t
cusolverSpXcsrpermHost(cusolverSpHandle_t handle,
                       int m,
                       int n,
                       int nnzA,
                       const cusparseMatDescr_t descrA,
                       int *csrRowPtrA,
                       int *csrColIndA,
                       const int *p,
                       const int *q,
                       int *map,
                       void *pBuffer);
```

Given a left permutation vector  $\mathbf{p}$  which corresponds to permutation matrix  $\mathbf{P}$  and a right permutation vector  $\mathbf{q}$  which corresponds to permutation matrix  $\mathbf{Q}$ , this function computes permutation of matrix  $\mathbf{A}$  by

$$B = P * A * Q^T$$

$\mathbf{A}$  is an  $m \times n$  sparse matrix that is defined in CSR storage format by the three arrays `csrValA`, `csrRowPtrA` and `csrColIndA`.

The operation is in-place, i.e. the matrix  $\mathbf{A}$  is overwritten by  $\mathbf{B}$ .

The permutation vector  $\mathbf{p}$  and  $\mathbf{q}$  are base 0.  $\mathbf{p}$  performs row permutation while  $\mathbf{q}$  performs column permutation. One can also use MATLAB command  $B = A(p,q)$  to permute matrix  $\mathbf{A}$ .

This function only computes sparsity pattern of  $\mathbf{B}$ . The user can use parameter `map` to get `csrValB` as well. The parameter `map` is an input/output. If the user sets `map=0:1:(nnzA-1)` before calling `csrperm`, `csrValB=csrValA(map)`.

The supported matrix type is `CUSPARSE_MATRIX_TYPE_GENERAL`. If  $\mathbf{A}$  is symmetric and only lower/upper part is provided, the user has to pass  $A + A^T$  into this function.

This function requires a buffer size returned by `csrperm_bufferSize()`. The address of `pBuffer` must be a multiple of 128 bytes. If it is not, `CUSOLVER_STATUS_INVALID_VALUE` is returned.

For example, if matrix **A** is

$$A = \begin{pmatrix} 1.0 & 2.0 & 3.0 \\ 4.0 & 5.0 & 6.0 \\ 7.0 & 8.0 & 9.0 \end{pmatrix}$$

and left permutation vector  $\mathbf{p} = (0, 2, 1)$ , right permutation vector  $\mathbf{q} = (2, 1, 0)$ , then  $P^*A^*Q^T$  is

$$P^*A^*Q^T = \begin{pmatrix} 3.0 & 2.0 & 1.0 \\ 9.0 & 8.0 & 7.0 \\ 6.0 & 5.0 & 4.0 \end{pmatrix}$$

Remark 1: only CPU (Host) path is provided.

Remark 2: the user can combine `csrsmrcm` and `csrperm` to get  $P^*A^*P^T$  which has less zero fill-in during QR factorization.

### Input

parameter	cusolverSp MemSpace	description
<code>handle</code>	host	handle to the cuSolver library context.
<code>m</code>	host	number of rows of matrix <b>A</b> .
<code>n</code>	host	number of columns of matrix <b>A</b> .
<code>nnzA</code>	host	number of nonzeros of matrix <b>A</b> . It is the size of <code>csrValA</code> and <code>csrColIndA</code> .
<code>descrA</code>	host	the descriptor of matrix <b>A</b> . The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> . Also, the supported index bases are <code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code> .
<code>csrRowPtrA</code>	host	integer array of <code>m+1</code> elements that contains the start of every row and end of last row plus one of matrix <b>A</b> .
<code>csrColIndA</code>	host	integer array of <code>nnzA</code> column indices of the nonzero elements of matrix <b>A</b> .
<code>p</code>	host	left permutation vector of size <code>m</code> .
<code>q</code>	host	right permutation vector of size <code>n</code> .
<code>map</code>	host	integer array of <code>nnzA</code> indices. If the user wants to get relationship between <b>A</b> and <b>B</b> , <code>map</code> must be set <code>0:1:(nnzA-1)</code> .
<code>pBuffer</code>	host	buffer allocated by the user, the size is returned by <code>csrperm_bufferSize()</code> .

### Output

parameter	hsolver	description
<code>csrRowPtrA</code>	host	integer array of $m+1$ elements that contains the start of every row and end of last row plus one of matrix <b>B</b> .
<code>csrColIndA</code>	host	integer array of <code>nnzA</code> column indices of the nonzero elements of matrix <b>B</b> .
<code>map</code>	host	integer array of <code>nnzA</code> indices that maps matrix <b>A</b> to matrix <b>B</b> .
<code>pBufferSizeInBytes</code>	host	number of bytes of the buffer.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $m, n, nnzA \leq 0$ ), base index is not 0 or 1.
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.
<code>CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code>	the matrix type is not supported.

### 2.5.3.7. `cusolverSpXcsrqrBatched()`

The create and destroy methods start and end the lifetime of a `csrqrInfo` object.

```
cusolverStatus_t
cusolverSpCreateCsrqrInfo(csrqrInfo_t *info);

cusolverStatus_t
cusolverSpDestroyCsrqrInfo(csrqrInfo_t info);
```

Analysis is the same for all data types, but each data type has a unique buffer size.

```

cusolverStatus_t
cusolverSpXcsrqrAnalysisBatched(cusolverSpHandle_t handle,
                                int m,
                                int n,
                                int nnzA,
                                const cusparseMatDescr_t descrA,
                                const int *csrRowPtrA,
                                const int *csrColIndA,
                                csrqrInfo_t info);

cusolverStatus_t
cusolverSpScsrqrBufferInfoBatched(cusolverSpHandle_t handle,
                                   int m,
                                   int n,
                                   int nnzA,
                                   const cusparseMatDescr_t descrA,
                                   const float *csrValA,
                                   const int *csrRowPtrA,
                                   const int *csrColIndA,
                                   int batchSize,
                                   csrqrInfo_t info,
                                   size_t *internalDataInBytes,
                                   size_t *workspaceInBytes);

cusolverStatus_t
cusolverSpDcsrqrBufferInfoBatched(cusolverSpHandle_t handle,
                                   int m,
                                   int n,
                                   int nnzA,
                                   const cusparseMatDescr_t descrA,
                                   const double *csrValA,
                                   const int *csrRowPtrA,
                                   const int *csrColIndA,
                                   int batchSize,
                                   csrqrInfo_t info,
                                   size_t *internalDataInBytes,
                                   size_t *workspaceInBytes);

```

Calculate buffer sizes for complex valued data types.

```

cusolverStatus_t
cusolverSpCcsrqrBufferInfoBatched(cusolverSpHandle_t handle,
                                   int m,
                                   int n,
                                   int nnzA,
                                   const cusparseMatDescr_t descrA,
                                   const cuComplex *csrValA,
                                   const int *csrRowPtrA,
                                   const int *csrColIndA,
                                   int batchSize,
                                   csrqrInfo_t info,
                                   size_t *internalDataInBytes,
                                   size_t *workspaceInBytes);

cusolverStatus_t
cusolverSpZcsrqrBufferInfoBatched(cusolverSpHandle_t handle,
                                   int m,
                                   int n,
                                   int nnzA,
                                   const cusparseMatDescr_t descrA,
                                   const cuDoubleComplex *csrValA,
                                   const int *csrRowPtrA,
                                   const int *csrColIndA,
                                   int batchSize,
                                   csrqrInfo_t info,
                                   size_t *internalDataInBytes,
                                   size_t *workspaceInBytes);

```

The S and D data types are real valued single and double precision, respectively.

```

cusolverStatus_t
cusolverSpScsrqrsvBatched(cusolverSpHandle_t handle,
                          int m,
                          int n,
                          int nnzA,
                          const cusparseMatDescr_t descrA,
                          const float *csrValA,
                          const int *csrRowPtrA,
                          const int *csrColIndA,
                          const float *b,
                          float *x,
                          int batchSize,
                          csrqrInfo_t info,
                          void *pBuffer);

cusolverStatus_t
cusolverSpDcsrqrsvBatched(cusolverSpHandle_t handle,
                          int m,
                          int n,
                          int nnz,
                          const cusparseMatDescr_t descrA,
                          const double *csrValA,
                          const int *csrRowPtrA,
                          const int *csrColIndA,
                          const double *b,
                          double *x,
                          int batchSize,
                          csrqrInfo_t info,
                          void *pBuffer);

```

The C and Z data types are complex valued single and double precision, respectively.

```
cusolverStatus_t
cusolverSpCsrqrsvBatched(cusolverSpHandle_t handle,
                        int m,
                        int n,
                        int nnzA,
                        const cusparseMatDescr_t descrA,
                        const cuComplex *csrValA,
                        const int *csrRowPtrA,
                        const int *csrColIndA,
                        const cuComplex *b,
                        cuComplex *x,
                        int batchSize,
                        csrqrInfo_t info,
                        void *pBuffer);

cusolverStatus_t
cusolverSpZcsrqrsvBatched(cusolverSpHandle_t handle,
                        int m,
                        int n,
                        int nnzA,
                        const cusparseMatDescr_t descrA,
                        const cuDoubleComplex *csrValA,
                        const int *csrRowPtrA,
                        const int *csrColIndA,
                        const cuDoubleComplex *b,
                        cuDoubleComplex *x,
                        int batchSize,
                        csrqrInfo_t info,
                        void *pBuffer);
```

The batched sparse QR factorization is used to solve either a set of least-squares problems

$$x_j = \operatorname{argmin} \|A_j z - b_j\|, j = 1, 2, \dots, \text{batchSize}$$

or a set of linear systems

$$A_j x_j = b_j, j = 1, 2, \dots, \text{batchSize}$$

where each  $A_j$  is a  $m \times n$  sparse matrix that is defined in CSR storage format by the four arrays **csrValA**, **csrRowPtrA** and **csrColIndA**.

The supported matrix type is **CUSPARSE\_MATRIX\_TYPE\_GENERAL**. If **A** is symmetric and only lower/upper part is provided, the user has to pass  $A + A^H$  into this function.

The prerequisite to use batched sparse QR has two-folds. First all matrices  $A_j$  must have the same sparsity pattern. Second, no column pivoting is used in least-square problem, so the solution is valid only if  $A_j$  is of full rank for all  $j = 1, 2, \dots, \text{batchSize}$ . All matrices have the same sparsity pattern, so only one copy of **csrRowPtrA** and **csrColIndA** is used. But the array **csrValA** stores coefficients of  $A_j$  one after another. In other words, **csrValA**[ $k \cdot \text{nnzA} : (k+1) \cdot \text{nnzA}$ ] is the value of  $A_k$ .

The batched QR uses opaque data structure **csrqrInfo** to keep intermediate data, for example, matrix **Q** and matrix **R** of QR factorization. The user needs to create **csrqrInfo** first by **cusolverSpCreateCsrqrInfo** before any function in batched QR operation.

The `csrqrInfo` would not release internal data until `cusolverSpDestroyCsrqrInfo` is called.

There are three routines in batched sparse QR, `cusolverSpXcsrqrAnalysisBatched`, `cusolverSp[S|D|C|Z]csrqrBufferInfoBatched` and `cusolverSp[S|D|C|Z]csrqrsvBatched`.

First, `cusolverSpXcsrqrAnalysisBatched` is the analysis phase, used to analyze sparsity pattern of matrix **Q** and matrix **R** of QR factorization. Also parallelism is extracted during analysis phase. Once analysis phase is done, the size of working space to perform QR is known. However `cusolverSpXcsrqrAnalysisBatched` uses CPU to analyze the structure of matrix **A**, and this may consume a lot of memory. If host memory is not sufficient to finish the analysis, `CUSOLVER_STATUS_ALLOC_FAILED` is returned. The required memory for analysis is proportional to zero fill-in in QR factorization. The user may need to perform some kind of reordering to minimize zero fill-in, for example, `colamd` or `symrcm` in MATLAB. `cuSolverSP` library provides `symrcm` (`cusolverSpXcsrsymrcm`).

Second, the user needs to choose proper `batchSize` and to prepare working space for sparse QR. There are two memory blocks used in batched sparse QR. One is internal memory block used to store matrix **Q** and matrix **R**. The other is working space used to perform numerical factorization. The size of the former is proportional to `batchSize`, and the size is specified by returned parameter `internalDataInBytes` of `cusolverSp[S|D|C|Z]csrqrBufferInfoBatched`. while the size of the latter is almost independent of `batchSize`, and the size is specified by returned parameter `workspaceInBytes` of `cusolverSp[S|D|C|Z]csrqrBufferInfoBatched`. The internal memory block is allocated implicitly during first call of `cusolverSp[S|D|C|Z]csrqrsvBatched`. The user only needs to allocate working space for `cusolverSp[S|D|C|Z]csrqrsvBatched`.

Instead of trying all batched matrices, the user can find maximum `batchSize` by querying `cusolverSp[S|D|C|Z]csrqrBufferInfoBatched`. For example, the user can increase `batchSize` till summation of `internalDataInBytes` and `workspaceInBytes` is greater than size of available device memory.

Suppose that the user needs to perform 253 linear solvers and available device memory is 2GB. if `cusolverSp[S|D|C|Z]csrqrsvBatched` can only afford `batchSize` 100, the user has to call `cusolverSp[S|D|C|Z]csrqrsvBatched` three times to finish all. The user calls `cusolverSp[S|D|C|Z]csrqrBufferInfoBatched` with `batchSize` 100. The opaque `info` would remember this `batchSize` and any subsequent call of `cusolverSp[S|D|C|Z]csrqrsvBatched` cannot exceed this value. In this example, the first two calls of `cusolverSp[S|D|C|Z]csrqrsvBatched` will use `batchSize` 100, and last call of `cusolverSp[S|D|C|Z]csrqrsvBatched` will use `batchSize` 53.

Example: suppose that  $A_0, A_1, \dots, A_9$  have the same sparsity pattern, the following code solves 10 linear systems  $A_j x_j = b_j, j = 0, 2, \dots, 9$  by batched sparse QR.

```
// Suppose that A0, A1, ..., A9 are m x m sparse matrix represented by CSR
// format,
// Each matrix Aj has nonzero nnzA, and shares the same csrRowPtrA and
// csrColIndA.
// csrValA is aggregation of A0, A1, ..., A9.
int m ; // number of rows and columns of each Aj
int nnzA ; // number of nonzeros of each Aj
int *csrRowPtrA ; // each Aj has the same csrRowPtrA
int *csrColIndA ; // each Aj has the same csrColIndA
double *csrValA ; // aggregation of A0,A1,...,A9
const int batchSize = 10; // 10 linear systems

cusolverSpHandle_t handle; // handle to cusolver library
csrqrInfo_t info = NULL;
cusparsedMatDescr_t descrA = NULL;
void *pBuffer = NULL; // working space for numerical factorization

// step 1: create a descriptor
cusparsedMatDescr(&descrA);
cusparsedMatIndexBase(descrA, CUSPARSE_INDEX_BASE_ONE); // A is base-1
cusparsedMatType(descrA, CUSPARSE_MATRIX_TYPE_GENERAL); // A is a general
matrix

// step 2: create empty info structure
cusolverSpCreateCsrqrInfo(&info);

// step 3: symbolic analysis
cusolverSpXcsrqrAnalysisBatched(
    handle, m, m, nnzA,
    descrA, csrRowPtrA, csrColIndA, info);

// step 4: allocate working space for Aj*xj=bj
cusolverSpDcsrqrBufferInfoBatched(
    handle, m, m, nnzA,
    descrA,
    csrValA, csrRowPtrA, csrColIndA,
    batchSize,
    info,
    &internalDataInBytes,
    &workspaceInBytes);

cudaMalloc(&pBuffer, workspaceInBytes);

// step 5: solve Aj*xj = bj
cusolverSpDcsrqrsvBatched(
    handle, m, m, nnzA,
    descrA, csrValA, csrRowPtrA, csrColIndA,
    b,
    x,
    batchSize,
    info,
    pBuffer);

// step 7: destroy info
cusolverSpDestroyCsrqrInfo(info);
```

Please refer to Appendix B for detailed examples.

Remark 1: only GPU (device) path is provided.

### Input

parameter	cusolverSp MemSpace	description
handle	host	handle to the cuSolverSP library context.
m	host	number of rows of each matrix $A_j$ .
n	host	number of columns of each matrix $A_j$ .
nnzA	host	number of nonzeros of each matrix $A_j$ . It is the size <code>csrColIndA</code> .
descrA	host	the descriptor of each matrix $A_j$ . The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> . Also, the supported index bases are <code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code> .
csrValA	device	<type> array of <code>nnzA*batchSize</code> nonzero elements of matrices $A_0, A_1, \dots$ . All matrices are aggregated one after another.
csrRowPtrA	device	integer array of <code>m+1</code> elements that contains the start of every row and the end of the last row plus one.
csrColIndA	device	integer array of <code>nnzA</code> column indices of the nonzero elements of each matrix $A_j$ .
b	device	<type> array of <code>m*batchSize</code> of right-hand-side vectors $b_0, b_1, \dots$ . All vectors are aggregated one after another.
batchSize	host	number of systems to be solved.
info	host	opaque structure for QR factorization.
pBuffer	device	buffer allocated by the user, the size is returned by <code>cusolverSpXcsrqrBufferInfoBatched()</code> .

## Output

parameter	cusolverSp MemSpace	description
x	device	<type> array of <code>m*batchSize</code> of solution vectors $x_0, x_1, \dots$ . All vectors are aggregated one after another.
internalDataInBytes	host	number of bytes of the internal data.
workspaceInBytes	host	number of bytes of the buffer in numerical factorization.

## Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $m, n, nnzA \leq 0$ ), base index is not 0 or 1.

CUSOLVER_STATUS_ARCH_MISMATCH	the device only supports compute capability 2.0 and above.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.
CUSOLVER_STATUS_MATRIX_TYPE_NOT_SUPPORTED	the matrix type is not supported.

## 2.6. cuSolverRF: Refactorization Reference

This chapter describes API of cuSolverRF, a library for fast refactorization.

### 2.6.1. cusolverRfAccessBundledFactors()

```
cusolverStatus_t
cusolverRfAccessBundledFactors( /* Input */
                               cusolverRfHandle_t handle,
                               /* Output (in the host memory) */
                               int* nnzM,
                               /* Output (in the device memory) */
                               int** Mp,
                               int** Mi,
                               double** Mx);
```

This routine allows direct access to the lower **L** and upper **U** triangular factors stored in the cuSolverRF library handle. The factors are compressed into a single matrix **M**= (**L**-**I**)+**U**, where the unitary diagonal of **L** is not stored. It is assumed that a prior call to the **cusolverRfRefactor()** was done in order to generate these triangular factors.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
nnzM	host	output	the number of non-zero elements of matrix <b>M</b> .
Mp	device	output	the array of offsets corresponding to the start of each row in the arrays <b>Mi</b> and <b>Mx</b> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <b>M</b> . The array size is <b>n+1</b> .
Mi	device	output	the array of column indices corresponding to the non-zero elements in the matrix <b>M</b> . It is assumed that this array is sorted by row and by column within each row. The array size is <b>nnzM</b> .
Mx	device	output	the array of values corresponding to the non-zero elements in the matrix <b>M</b> . It is assumed that this array is sorted by row and by column within each row. The array size is <b>nnzM</b> .

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.

## 2.6.2. `cusolverRfAnalyze()`

```
cusolverStatus_t
cusolverRfAnalyze(cusolverRfHandle_t handle);
```

This routine performs the appropriate analysis of parallelism available in the LU re-factorization depending upon the algorithm chosen by the user.

$$A = L * U$$

It is assumed that a prior call to the `cusolverRfSetup[Host|Device]()` was done in order to create internal data structures needed for the analysis.

This routine needs to be called only once for a single linear system

$$A_i x_i = f_i$$

parameter	MemSpace	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>in/out</code>	the handle to the cuSolverRF library.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	an allocation of memory failed.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.6.3. cusolverRfSetupDevice()

```
cusolverStatus_t
cusolverRfSetupDevice(/* Input (in the device memory) */
    int n,
    int nnzA,
    int* csrRowPtrA,
    int* csrColIndA,
    double* csrValA,
    int nnzL,
    int* csrRowPtrL,
    int* csrColIndL,
    double* csrValL,
    int nnzU,
    int* csrRowPtrU,
    int* csrColIndU,
    double* csrValU,
    int* P,
    int* Q,
    /* Output */
    cusolverRfHandle_t handle);
```

This routine assembles the internal data structures of the cuSolverRF library. It is often the first routine to be called after the call to the `cusolverRfCreate()` routine.

This routine accepts as input (on the device) the original matrix **A**, the lower (**L**) and upper (**U**) triangular factors, as well as the left (**P**) and the right (**Q**) permutations resulting from the full LU factorization of the first (**i=1**) linear system

$$A_i x_i = f_i$$

The permutations **P** and **Q** represent the final composition of all the left and right reorderings applied to the original matrix **A**, respectively. However, these permutations are often associated with partial pivoting and reordering to minimize fill-in, respectively.

This routine needs to be called only once for a single linear system

$$A_i x_i = f_i$$

parameter	MemSpace	In/out	Meaning
<b>n</b>	host	input	the number of rows (and columns) of matrix <b>A</b> .
<b>nnzA</b>	host	input	the number of non-zero elements of matrix <b>A</b> .
<b>csrRowPtrA</b>	device	input	the array of offsets corresponding to the start of each row in the arrays <b>csrColIndA</b> and <b>csrValA</b> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix. The array size is <b>n+1</b> .
<b>csrColIndA</b>	device	input	the array of column indices corresponding to the non-zero elements in the matrix. It

			is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzA</code> .
<code>csrValA</code>	<code>device</code>	<code>input</code>	the array of values corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzA</code> .
<code>nnzL</code>	<code>host</code>	<code>input</code>	the number of non-zero elements of matrix <code>L</code> .
<code>csrRowPtrL</code>	<code>device</code>	<code>input</code>	the array of offsets corresponding to the start of each row in the arrays <code>csrColIndL</code> and <code>csrValL</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <code>L</code> . The array size is <code>n+1</code> .
<code>csrColIndL</code>	<code>device</code>	<code>input</code>	the array of column indices corresponding to the non-zero elements in the matrix <code>L</code> . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzL</code> .
<code>csrValL</code>	<code>device</code>	<code>input</code>	the array of values corresponding to the non-zero elements in the matrix <code>L</code> . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzL</code> .
<code>nnzU</code>	<code>host</code>	<code>input</code>	the number of non-zero elements of matrix <code>U</code> .
<code>csrRowPtrU</code>	<code>device</code>	<code>input</code>	the array of offsets corresponding to the start of each row in the arrays <code>csrColIndU</code> and <code>csrValU</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <code>U</code> . The array size is <code>n+1</code> .
<code>csrColIndU</code>	<code>device</code>	<code>input</code>	the array of column indices corresponding to the non-zero elements in the matrix <code>U</code> . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzU</code> .
<code>csrValU</code>	<code>device</code>	<code>input</code>	the array of values corresponding to the non-zero elements in the matrix <code>U</code> . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzU</code> .
<code>P</code>	<code>device</code>	<code>input</code>	the left permutation (often associated with pivoting). The array size is <code>n</code> .
<code>Q</code>	<code>device</code>	<code>input</code>	the right permutation (often associated with reordering). The array size is <code>n</code> .
<code>handle</code>	<code>host</code>	<code>output</code>	the handle to the GLU library.

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	an unsupported value or parameter was passed.
CUSOLVER_STATUS_ALLOC_FAILED	an allocation of memory failed.
CUSOLVER_STATUS_EXECUTION_FAILED	a kernel failed to launch on the GPU.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

## 2.6.4. cusolverRfSetupHost()

```

cusolverStatus_t
cusolverRfSetupHost(/* Input (in the host memory) */
    int n,
    int nnzA,
    int* h_csrRowPtrA,
    int* h_csrColIndA,
    double* h_csrValA,
    int nnzL,
    int* h_csrRowPtrL,
    int* h_csrColIndL,
    double* h_csrValL,
    int nnzU,
    int* h_csrRowPtrU,
    int* h_csrColIndU,
    double* h_csrValU,
    int* h_P,
    int* h_Q,
    /* Output */
    cusolverRfHandle_t handle);

```

This routine assembles the internal data structures of the cuSolverRF library. It is often the first routine to be called after the call to the `cusolverRfCreate()` routine.

This routine accepts as input (on the host) the original matrix **A**, the lower (**L**) and upper (**U**) triangular factors, as well as the left (**P**) and the right (**Q**) permutations resulting from the full LU factorization of the first (**i=1**) linear system

$$A_i x_i = f_i$$

The permutations **P** and **Q** represent the final composition of all the left and right reorderings applied to the original matrix **A**, respectively. However, these permutations are often associated with partial pivoting and reordering to minimize fill-in, respectively.

This routine needs to be called only once for a single linear system

$$A_i x_i = f_i$$

parameter	MemSpace	In/out	Meaning
<b>n</b>	host	input	the number of rows (and columns) of matrix <b>A</b> .

<code>nnzA</code>	<code>host</code>	<code>input</code>	the number of non-zero elements of matrix <code>A</code> .
<code>h_csrRowPtrA</code>	<code>host</code>	<code>input</code>	the array of offsets corresponding to the start of each row in the arrays <code>h_csrColIndA</code> and <code>h_csrValA</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix. The array size is <code>n + 1</code> .
<code>h_csrColIndA</code>	<code>host</code>	<code>input</code>	the array of column indices corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzA</code> .
<code>h_csrValA</code>	<code>host</code>	<code>input</code>	the array of values corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzA</code> .
<code>nnzL</code>	<code>host</code>	<code>input</code>	the number of non-zero elements of matrix <code>L</code> .
<code>h_csrRowPtrL</code>	<code>host</code>	<code>input</code>	the array of offsets corresponding to the start of each row in the arrays <code>h_csrColIndL</code> and <code>h_csrValL</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <code>L</code> . The array size is <code>n + 1</code> .
<code>h_csrColIndL</code>	<code>host</code>	<code>input</code>	the array of column indices corresponding to the non-zero elements in the matrix <code>L</code> . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzL</code> .
<code>h_csrValL</code>	<code>host</code>	<code>input</code>	the array of values corresponding to the non-zero elements in the matrix <code>L</code> . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzL</code> .
<code>nnzU</code>	<code>host</code>	<code>input</code>	the number of non-zero elements of matrix <code>U</code> .
<code>h_csrRowPtrU</code>	<code>host</code>	<code>input</code>	the array of offsets corresponding to the start of each row in the arrays <code>h_csrColIndU</code> and <code>h_csrValU</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <code>U</code> . The array size is <code>n + 1</code> .
<code>h_csrColIndU</code>	<code>host</code>	<code>input</code>	the array of column indices corresponding to the non-zero elements in the matrix <code>U</code> . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzU</code> .

<code>h_csrValU</code>	host	input	the array of values corresponding to the non-zero elements in the matrix $u$ . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzU</code> .
<code>h_P</code>	host	input	the left permutation (often associated with pivoting). The array size is <code>n</code> .
<code>h_Q</code>	host	input	the right permutation (often associated with reordering). The array size is <code>n</code> .
<code>handle</code>	host	output	the handle to the cuSolverRF library.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	an unsupported value or parameter was passed.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	an allocation of memory failed.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

## 2.6.5. `cusolverRfCreate()`

```
cusolverStatus_t cusolverRfCreate(cusolverRfHandle_t *handle);
```

This routine initializes the cuSolverRF library. It allocates required resources and must be called prior to any other cuSolverRF library routine.

parameter	MemSpace	In/out	Meaning
<code>handle</code>	host	output	the pointer to the cuSolverRF library handle.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	an allocation of memory failed.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

## 2.6.6. cusolverRfExtractBundledFactorsHost()

```
cusolverStatus_t
cusolverRfExtractBundledFactorsHost(/* Input */
    cusolverRfHandle_t handle,
    /* Output (in the host memory) */
    int* h_nnzM,
    int** h_Mp,
    int** h_Mi,
    double** h_Mx);
```

This routine extracts lower (**L**) and upper (**U**) triangular factors from the cuSolverRF library handle into the host memory. The factors are compressed into a single matrix  $\mathbf{M} = (\mathbf{L} - \mathbf{I}) + \mathbf{U}$ , where the unitary diagonal of (**L**) is not stored. It is assumed that a prior call to the `cusolverRfRefactor()` was done in order to generate these triangular factors.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
h_nnzM	host	output	the number of non-zero elements of matrix <b>M</b> .
h_Mp	host	output	the array of offsets corresponding to the start of each row in the arrays <code>h_Mi</code> and <code>h_Mx</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <b>M</b> . The array size is <code>n+1</code> .
h_Mi	host	output	the array of column indices corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <code>h_nnzM</code> .
h_Mx	host	output	the array of values corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <code>h_nnzM</code> .

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_ALLOC_FAILED	an allocation of memory failed.
CUSOLVER_STATUS_EXECUTION_FAILED	a kernel failed to launch on the GPU.

## 2.6.7. cusolverRfExtractSplitFactorsHost()

```

cusolverStatus_t
cusolverRfExtractSplitFactorsHost(/* Input */
                                  cusolverRfHandle_t handle,
                                  /* Output (in the host memory) */
                                  int* h_nnzL,
                                  int** h_Lp,
                                  int** h_Li,
                                  double** h_Lx,
                                  int* h_nnzU,
                                  int** h_Up,
                                  int** h_Ui,
                                  double** h_Ux);

```

This routine extracts lower (**L**) and upper (**U**) triangular factors from the cuSolverRF library handle into the host memory. It is assumed that a prior call to the **cusolverRfRefactor()** was done in order to generate these triangular factors.

parameter	MemSpace	In/out	Meaning
<b>handle</b>	host	input	the handle to the cuSolverRF library.
<b>h_nnzL</b>	host	output	the number of non-zero elements of matrix <b>L</b> .
<b>h_Lp</b>	host	output	the array of offsets corresponding to the start of each row in the arrays <b>h_Li</b> and <b>h_Lx</b> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <b>L</b> . The array size is <b>n+1</b> .
<b>h_Li</b>	host	output	the array of column indices corresponding to the non-zero elements in the matrix <b>L</b> . It is assumed that this array is sorted by row and by column within each row. The array size is <b>h_nnzL</b> .
<b>h_Lx</b>	host	output	the array of values corresponding to the non-zero elements in the matrix <b>L</b> . It is assumed that this array is sorted by row and by column within each row. The array size is <b>h_nnzL</b> .
<b>h_nnzU</b>	host	output	the number of non-zero elements of matrix <b>U</b> .
<b>h_Up</b>	host	output	the array of offsets corresponding to the start of each row in the arrays <b>h_Ui</b> and <b>h_Ux</b> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <b>U</b> . The array size is <b>n+1</b> .
<b>h_Ui</b>	host	output	the array of column indices corresponding to the non-zero elements in the matrix <b>U</b> . It is assumed that this array is sorted by

			row and by column within each row. The array size is <code>h_nnzU</code> .
<code>h_Ux</code>	<code>host</code>	<code>output</code>	the array of values corresponding to the non-zero elements in the matrix $U$ . It is assumed that this array is sorted by row and by column within each row. The array size is <code>h_nnzU</code> .

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	an allocation of memory failed.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.

## 2.6.8. `cusolverRfDestroy()`

```
cusolverStatus_t cusolverRfDestroy(cusolverRfHandle_t handle);
```

This routine shuts down the `cuSolverRF` library. It releases acquired resources and must be called after all the `cuSolverRF` library routines.

parameter	MemSpace	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>input</code>	the <code>cuSolverRF</code> library handle.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.

## 2.6.9. `cusolverRfGetMatrixFormat()`

```
cusolverStatus_t
cusolverRfGetMatrixFormat(cusolverRfHandle_t handle,
                          cusolverRfMatrixFormat_t *format,
                          cusolverRfUnitDiagonal_t *diag);
```

This routine gets the matrix format used in the `cusolverRfSetupDevice()`, `cusolverRfSetupHost()`, `cusolverRfResetValues()`, `cusolverRfExtractBundledFactorsHost()` and `cusolverRfExtractSplitFactorsHost()` routines.

parameter	MemSpace	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>input</code>	the handle to the <code>cuSolverRF</code> library.
<code>format</code>	<code>host</code>	<code>output</code>	the enumerated matrix format type.
<code>diag</code>	<code>host</code>	<code>output</code>	the enumerated unit diagonal type.

**Status Returned**

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

**2.6.10. cusolverRfGetNumericProperties()**

```
cusolverStatus_t
cusolverRfGetNumericProperties(cusolverRfHandle_t handle,
                              double *zero,
                              double *boost);
```

This routine gets the numeric values used for checking for "zero" pivot and for boosting it in the **cusolverRfRefactor()** and **cusolverRfSolve()** routines. The numeric boosting will be used only if **boost > 0.0**.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
zero	host	output	the value below which zero pivot is flagged.
boost	host	output	the value which is substituted for zero pivot (if the later is flagged).

**Status Returned**

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

**2.6.11. cusolverRfGetNumericBoostReport()**

```
cusolverStatus_t
cusolverRfGetNumericBoostReport(cusolverRfHandle_t handle,
                                cusolverRfNumericBoostReport_t *report);
```

This routine gets the report whether numeric boosting was used in the **cusolverRfRefactor()** and **cusolverRfSolve()** routines.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
report	host	output	the enumerated boosting report type.

**Status Returned**

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

## 2.6.12. cusolverRfGetResetValuesFastMode()

```
cusolverStatus_t
cusolverRfGetResetValuesFastMode(cusolverRfHandle_t handle,
                                 rfResetValuesFastMode_t *fastMode);
```

This routine gets the mode used in the **cusolverRfResetValues** routine.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
fastMode	host	output	the enumerated mode type.

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

## 2.6.13. cusolverRfGet\_Algs()

```
cusolverStatus_t
cusolverRfGet_Algs(cusolverRfHandle_t handle,
                   cusolverRfFactorization_t* fact_alg,
                   cusolverRfTriangularSolve_t* solve_alg);
```

This routine gets the algorithm used for the refactorization in **cusolverRfRefactor()** and the triangular solve in **cusolverRfSolve()**.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
alg	host	output	the enumerated algorithm type.

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

## 2.6.14. cusolverRfRefactor()

```
cusolverStatus_t cusolverRfRefactor(cusolverRfHandle_t handle);
```

This routine performs the LU re-factorization

$$A = L * U$$

exploring the available parallelism on the GPU. It is assumed that a prior call to the **glu\_analyze()** was done in order to find the available parallelism.

This routine may be called multiple times, once for each of the linear systems

$$A_i x_i = f_i$$

There are some constraints to the combination of algorithms used for refactorization and solving routines, `cusolverRfRefactor()` and `cusolverRfSolve()`. The wrong combination generates the error code `CUSOLVER_STATUS_INVALID_VALUE`. The table below summarizes the supported combinations of algorithms:

#### Compatible algorithms for solving and refactorization routines.

Factorization	Solving
<code>CUSOLVERRF_FACTORIZATION_ALG0</code>	<code>TRIANGULAR_SOLVE_ALG1</code>
<code>CUSOLVERRF_FACTORIZATION_ALG1</code>	<code>TRIANGULAR_SOLVE_ALG2</code> , <code>TRIANGULAR_SOLVE_ALG3</code>
<code>CUSOLVERRF_FACTORIZATION_ALG2</code>	<code>TRIANGULAR_SOLVE_ALG2</code> , <code>TRIANGULAR_SOLVE_ALG3</code>

parameter	Memory	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>in/out</code>	the handle to the cuSolverRF library.

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.
<code>CUSOLVER_STATUS_ZERO_PIVOT</code>	a zero pivot was encountered during the computation.

## 2.6.15. `cusolverRfResetValues()`

```
cusolverStatus_t
cusolverRfResetValues(/* Input (in the device memory) */
    int n,
    int nnzA,
    int* csrRowPtrA,
    int* csrColIndA,
    double* csrValA,
    int* P,
    int* Q,
    /* Output */
    cusolverRfHandle_t handle);
```

This routine updates internal data structures with the values of the new coefficient matrix. It is assumed that the arrays `csrRowPtrA`, `csrColIndA`, `P` and `Q` have not changed since the last call to the `cusolverRfSetup[Host|Device]` routine. This assumption reflects the fact that the sparsity pattern of coefficient matrices as well as reordering to minimize fill-in and pivoting remain the same in the set of linear systems

$$A_i x_i = f_i$$

This routine may be called multiple times, once for each of the linear systems

$$A_i x_i = f_i$$

parameter	MemSpace	In/out	Meaning
<code>n</code>	host	input	the number of rows (and columns) of matrix <b>A</b> .
<code>nnzA</code>	host	input	the number of non-zero elements of matrix <b>A</b> .
<code>csrRowPtrA</code>	device	input	the array of offsets corresponding to the start of each row in the arrays <code>csrColIndA</code> and <code>csrValA</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix. The array size is <code>n+1</code> .
<code>csrColIndA</code>	device	input	the array of column indices corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzA</code> .
<code>csrValA</code>	device	input	the array of values corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzA</code> .
<code>P</code>	device	input	the left permutation (often associated with pivoting). The array size is <code>n</code> .
<code>Q</code>	device	input	the right permutation (often associated with reordering). The array size is <code>n</code> .
<code>handle</code>	host	output	the handle to the <code>cuSolverRF</code> library.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	an unsupported value or parameter was passed.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.

## 2.6.16. `cusolverRfSetMatrixFormat()`

```
cusolverStatus_t
cusolverRfSetMatrixFormat(cusolverRfHandle_t handle,
                          gluMatrixFormat_t format,
                          gluUnitDiagonal_t diag);
```

This routine sets the matrix format used in the `cusolverRfSetupDevice()`, `cusolverRfSetupHost()`, `cusolverRfResetValues()`, `cusolverRfExtractBundledFactorsHost()` and

**cusolverRfExtractSplitFactorsHost()** routines. It may be called once prior to **cusolverRfSetupDevice()** and **cusolverRfSetupHost()** routines.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
format	host	input	the enumerated matrix format type.
diag	host	input	the enumerated unit diagonal type.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	an enumerated mode parameter is wrong.

## 2.6.17. cusolverRfSetNumericProperties()

```
cusolverStatus_t
cusolverRfSetNumericProperties(cusolverRfHandle_t handle,
                              double zero,
                              double boost);
```

This routine sets the numeric values used for checking for "zero" pivot and for boosting it in the **cusolverRfRefactor()** and **cusolverRfSolve()** routines. It may be called multiple times prior to **cusolverRfRefactor()** and **cusolverRfSolve()** routines. The numeric boosting will be used only if **boost > 0.0**.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
zero	host	input	the value below which zero pivot is flagged.
boost	host	input	the value which is substituted for zero pivot (if the later is flagged).

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

## 2.6.18. cusolverRfSetResetValuesFastMode()

```
cusolverStatus_t
cusolverRfSetResetValuesFastMode(cusolverRfHandle_t handle,
                                  gluResetValuesFastMode_t fastMode);
```

This routine sets the mode used in the **cusolverRfResetValues** routine. The fast mode requires extra memory and is recommended only if very fast calls

to `cusolverRfResetValues()` are needed. It may be called once prior to `cusolverRfAnalyze()` routine.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
fastMode	host	input	the enumerated mode type.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	an enumerated mode parameter is wrong.

## 2.6.19. cusolverRfSetAlgs()

```
cusolverStatus_t
cusolverRfSetAlgs(cusolverRfHandle_t handle,
                  gluFactorization_t fact_alg,
                  gluTriangularSolve_t alg);
```

This routine sets the algorithm used for the refactorization in `cusolverRfRefactor()` and the triangular solve in `cusolverRfSolve()`. It may be called once prior to `cusolverRfAnalyze()` routine.

parameter	MemSpace	In/out	Meaning
handle	host	input	the handle to the cuSolverRF library.
alg	host	input	the enumerated algorithm type.

#### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.

## 2.6.20. cusolverRfSolve()

```
cusolverStatus_t
cusolverRfSolve(/* Input (in the device memory) */
               cusolverRfHandle_t handle,
               int *P,
               int *Q,
               int nrhs,
               double *Temp,
               int ldt,
               /* Input/Output (in the device memory) */
               double *XF,
               /* Input */
               int ldxf);
```

This routine performs the forward and backward solve with the lower  $L \in \mathbb{R}^{n \times n}$  and upper  $U \in \mathbb{R}^{n \times n}$  triangular factors resulting from the LU re-factorization

$$A = L * U$$

which is assumed to have been computed by a prior call to the `cusolverRfRefactor()` routine.

The routine can solve linear systems with multiple right-hand-sides (rhs),

$$AX = (LU)X = L(UX) = LY = F \text{ where } UX = Y$$

even though currently only a single rhs is supported.

This routine may be called multiple times, once for each of the linear systems

$$A_i x_i = f_i$$

parameter	MemSpace	In/out	Meaning
handle	host	output	the handle to the cuSolverRF library.
P	device	input	the left permutation (often associated with pivoting). The array size in n.
Q	device	input	the right permutation (often associated with reordering). The array size in n.
nrhs	host	input	the number right-hand-sides to be solved.
Temp	host	input	the dense matrix that contains temporary workspace (of size <code>ldt*nrhs</code> ).
ldt	host	input	the leading dimension of dense matrix Temp ( <code>ldt &gt;= n</code> ).
XF	host	in/out	the dense matrix that contains the right-hand-sides <code>F</code> and solutions <code>x</code> (of size <code>ldxf*nrhs</code> ).
ldxf	host	input	the leading dimension of dense matrix <code>XF</code> ( <code>ldxf &gt;= n</code> ).

### Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	an unsupported value or parameter was passed.
CUSOLVER_STATUS_EXECUTION_FAILED	a kernel failed to launch on the GPU.
CUSOLVER_STATUS_INTERNAL_ERROR	an internal operation failed.

## 2.6.21. cusolverRfBatchSetupHost()

```

cusolverStatus_t
cusolverRfBatchSetupHost(/* Input (in the host memory) */
    int batchSize,
    int n,
    int nnzA,
    int* h_csrRowPtrA,
    int* h_csrColIndA,
    double *h_csrValA_array[],
    int nnzL,
    int* h_csrRowPtrL,
    int* h_csrColIndL,
    double *h_csrValL,
    int nnzU,
    int* h_csrRowPtrU,
    int* h_csrColIndU,
    double *h_csrValU,
    int* h_P,
    int* h_Q,
    /* Output */
    cusolverRfHandle_t handle);

```

This routine assembles the internal data structures of the cuSolverRF library for batched operation. It is called after the call to the **cusolverRfCreate()** routine, and before any other batched routines.

The batched operation assumes that the user has the following linear systems

$$A_j x_j = b_j, j = 1, 2, \dots, \text{batchSize}$$

where each matrix in the set  $\{A_j\}$  has the same sparsity pattern, and quite similar such that factorization can be done by the same permutation **P** and **Q**. In other words,  $A_j, j > 1$  is a small perturbation of  $A_1$ .

This routine accepts as input (on the host) the original matrix **A** (sparsity pattern and batched values), the lower (**L**) and upper (**U**) triangular factors, as well as the left (**P**) and the right (**Q**) permutations resulting from the full LU factorization of the first (**i=1**) linear system

$$A_i x_i = f_i$$

The permutations **P** and **Q** represent the final composition of all the left and right reorderings applied to the original matrix **A**, respectively. However, these permutations are often associated with partial pivoting and reordering to minimize fill-in, respectively.

Remark 1: the matrices **A**, **L** and **U** must be CSR format and base-0.

Remark 2: to get best performance, **batchSize** should be multiple of 32 and greater or equal to 32. The algorithm is memory-bound, once bandwidth limit is reached, there is no room to improve performance by large **batchSize**. In practice, **batchSize** of 32 - 128 is often enough to obtain good performance, but in some cases larger **batchSize** might be beneficial.

This routine needs to be called only once for a single linear system

$$A_i x_i = f_i$$

parameter	MemSpace	In/out	Meaning
batchSize	host	input	the number of matrices in the batched mode.
n	host	input	the number of rows (and columns) of matrix <b>A</b> .
nnzA	host	input	the number of non-zero elements of matrix <b>A</b> .
h_csrRowPtrA	host	input	the array of offsets corresponding to the start of each row in the arrays <b>h_csrColIndA</b> and <b>h_csrValA</b> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix. The array size is <b>n + 1</b> .
h_csrColIndA	host	input	the array of column indices corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row and by column within each row. The array size is <b>nnzA</b> .
h_csrValA_array	host	input	array of pointers of size <b>batchSize</b> , each pointer points to the array of values corresponding to the non-zero elements in the matrix.
nnzL	host	input	the number of non-zero elements of matrix <b>L</b> .
h_csrRowPtrL	host	input	the array of offsets corresponding to the start of each row in the arrays <b>h_csrColIndL</b> and <b>h_csrValL</b> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix <b>L</b> . The array size is <b>n+1</b> .
h_csrColIndL	host	input	the array of column indices corresponding to the non-zero elements in the matrix <b>L</b> . It is assumed that this array is sorted by row and by column within each row. The array size is <b>nnzL</b> .
h_csrValL	host	input	the array of values corresponding to the non-zero elements in the matrix <b>L</b> . It is assumed that this array is sorted by row and by column within each row. The array size is <b>nnzL</b> .
nnzU	host	input	the number of non-zero elements of matrix <b>U</b> .
h_csrRowPtrU	host	input	the array of offsets corresponding to the start of each row in the arrays <b>h_csrColIndU</b> and <b>h_csrValU</b> . This

			array has also an extra entry at the end that stores the number of non-zero elements in the matrix $\sigma$ . The array size is $n+1$ .
<code>h_csrColIndU</code>	<code>host</code>	<code>input</code>	the array of column indices corresponding to the non-zero elements in the matrix $\sigma$ . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzU</code> .
<code>h_csrValU</code>	<code>host</code>	<code>input</code>	the array of values corresponding to the non-zero elements in the matrix $\sigma$ . It is assumed that this array is sorted by row and by column within each row. The array size is <code>nnzU</code> .
<code>h_P</code>	<code>host</code>	<code>input</code>	the left permutation (often associated with pivoting). The array size is <code>n</code> .
<code>h_Q</code>	<code>host</code>	<code>input</code>	the right permutation (often associated with reordering). The array size is <code>n</code> .
<code>handle</code>	<code>host</code>	<code>output</code>	the handle to the cuSolverRF library.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	an unsupported value or parameter was passed.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	an allocation of memory failed.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

## 2.6.22. `cusolverRfBatchAnalyze()`

```
cusolverStatus_t cusolverRfBatchAnalyze(cusolverRfHandle_t handle);
```

This routine performs the appropriate analysis of parallelism available in the batched LU re-factorization.

It is assumed that a prior call to the `cusolverRfBatchSetup[Host]()` was done in order to create internal data structures needed for the analysis.

This routine needs to be called only once for a single linear system

$$A_j x_j = b_j, j = 1, 2, \dots, \text{batchSize}$$

parameter	Memory	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>in/out</code>	the handle to the cuSolverRF library.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	an allocation of memory failed.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 2.6.23. `cusolverRfBatchResetValues()`

```
cusolverStatus_t
cusolverRfBatchResetValues(/* Input (in the device memory) */
    int batchSize,
    int n,
    int nnzA,
    int* csrRowPtrA,
    int* csrColIndA,
    double* csrValA_array[],
    int *P,
    int *Q,
    /* Output */
    cusolverRfHandle_t handle);
```

This routine updates internal data structures with the values of the new coefficient matrix. It is assumed that the arrays `csrRowPtrA`, `csrColIndA`, `P` and `Q` have not changed since the last call to the `cusolverRfbatch_setup_host` routine.

This assumption reflects the fact that the sparsity pattern of coefficient matrices as well as reordering to minimize fill-in and pivoting remain the same in the set of linear systems

$$A_j x_j = b_j, j = 1, 2, \dots, \text{batchSize}$$

The input parameter `csrValA_array` is an array of pointers on device memory. `csrValA_array(j)` points to matrix  $A_j$  which is also on device memory.

parameter	MemSpace	In/out	Meaning
<code>batchSize</code>	host	input	the number of matrices in batched mode.
<code>n</code>	host	input	the number of rows (and columns) of matrix <b>A</b> .
<code>nnzA</code>	host	input	the number of non-zero elements of matrix <b>A</b> .
<code>csrRowPtrA</code>	device	input	the array of offsets corresponding to the start of each row in the arrays <code>csrColIndA</code> and <code>csrValA</code> . This array has also an extra entry at the end that stores the number of non-zero elements in the matrix. The array size is <code>n+1</code> .
<code>csrColIndA</code>	device	input	the array of column indices corresponding to the non-zero elements in the matrix. It is assumed that this array is sorted by row

			and by column within each row. The array size is <code>nnzA</code> .
<code>csrValA_array</code>	<code>device</code>	<code>input</code>	array of pointers of size <code>batchSize</code> , each pointer points to the array of values corresponding to the non-zero elements in the matrix.
<code>P</code>	<code>device</code>	<code>input</code>	the left permutation (often associated with pivoting). The array size in <code>n</code> .
<code>Q</code>	<code>device</code>	<code>input</code>	the right permutation (often associated with reordering). The array size in <code>n</code> .
<code>handle</code>	<code>host</code>	<code>output</code>	the handle to the <code>cuSolverRF</code> library.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	an unsupported value or parameter was passed.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.

## 2.6.24. `cusolverRfBatchRefactor()`

```
cusolverStatus_t cusolverRfBatchRefactor(cusolverRfHandle_t handle);
```

This routine performs the LU re-factorization

$$M_j = P * A_j * Q^T = L_j * U_j$$

exploring the available parallelism on the GPU. It is assumed that a prior call to the `cusolverRfBatchAnalyze()` was done in order to find the available parallelism.

Remark: `cusolverRfBatchRefactor()` would not report any failure of LU refactorization. The user has to call `cusolverRfBatchZeroPivot()` to know which matrix failed the LU refactorization.

parameter	Memory	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>in/out</code>	the handle to the <code>cuSolverRF</code> library.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.

## 2.6.25. cusolverRfBatchSolve()

```

cusolverStatus_t
cusolverRfBatchSolve(/* Input (in the device memory) */
                    cusolverRfHandle_t handle,
                    int *P,
                    int *Q,
                    int nrhs,
                    double *Temp,
                    int ldt,
                    /* Input/Output (in the device memory) */
                    double *XF_array[],
                    /* Input */
                    int ldx);

```

To solve  $A_j * x_j = b_j$ , first we reform the equation by  $M_j * Q * x_j = P * b_j$

where  $M_j = P * A_j * Q^T$ . Then do refactorization  $M_j = L_j * U_j$  by

**cusolverRfBatch\_Refactor()**. Further **cusolverRfBatch\_Solve()** takes over the remaining steps, including:

$$z_j = P * b_j$$

$$M_j * y_j = z_j$$

$$x_j = Q^T * y_j$$

The input parameter **XF\_array** is an array of pointers on device memory. **XF\_array(j)** points to matrix  $x_j$  which is also on device memory.

Remark 1: only a single rhs is supported.

Remark 2: no singularity is reported during backward solve. If some matrix  $A_j$  failed the refactorization and  $U_j$  has some zero diagonal, backward solve would compute NAN. The user has to call **cusolverRfBatch\_Zero\_Pivot** to check if refactorization is successful or not.

parameter	Memory	In/out	Meaning
handle	host	output	the handle to the cuSolverRF library.
P	device	input	the left permutation (often associated with pivoting). The array size in n.
Q	device	input	the right permutation (often associated with reordering). The array size in n.
nrhs	host	input	the number right-hand-sides to be solved.
Temp	host	input	the dense matrix that contains temporary workspace (of size ldt*nrhs).
ldt	host	input	the leading dimension of dense matrix Temp (ldt >= n).
XF_array	device	in/out	array of pointers of size batchSize, each pointer points to the dense matrix

			that contains the right-hand-sides $\mathbf{F}$ and solutions $\mathbf{x}$ (of size $\text{ldxf} \times \text{nrhs}$ ).
<code>ldxf</code>	<code>host</code>	<code>input</code>	the leading dimension of dense matrix $\mathbf{XF}$ ( $\text{ldxf} \geq n$ ).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	an unsupported value or parameter was passed.
<code>CUSOLVER_STATUS_EXECUTION_FAILED</code>	a kernel failed to launch on the GPU.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

## 2.6.26. cusolverRfBatchZeroPivot()

```
cusolverStatus_t
cusolverRfBatchZeroPivot(/* Input */
                        cusolverRfHandle_t handle
                        /* Output (in the host memory) */
                        int *position);
```

Although  $A_j$  is close to each other, it does not mean  $M_j = P^* A_j^* Q^T = L_j^* U_j$  exists for every  $j$ . The user can query which matrix failed LU refactorization by checking corresponding value in `position` array. The input parameter `position` is an integer array of size `batchSize`.

The  $j$ -th component denotes the refactorization result of matrix  $A_j$ . If `position(j)` is `-1`, the LU refactorization of matrix  $A_j$  is successful. If `position(j)` is `k >= 0`, matrix  $A_j$  is not LU factorizable and its matrix  $U_j(j,j)$  is zero.

The return value of `cusolverRfBatch_Zero_Pivot` is `CUSOLVER_STATUS_ZERO_PIVOT` if there exists one  $A_j$  which failed LU refactorization. The user can redo LU factorization to get new permutation  $\mathbf{P}$  and  $\mathbf{Q}$  if error code `CUSOLVER_STATUS_ZERO_PIVOT` is returned.

parameter	MemSpace	In/out	Meaning
<code>handle</code>	<code>host</code>	<code>input</code>	the handle to the cuSolverRF library.
<code>position</code>	<code>host</code>	<code>output</code>	integer array of size <code>batchSize</code> . The value of <code>position(j)</code> reports singularity of matrix $A_j$ , <code>-1</code> if no structural/numerical zero, <code>k &gt;= 0</code> if $A_j(k,k)$ is either structural zero or numerical zero.

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.

<code>CUSOLVER_STATUS_ZERO_PIVOT</code>	a zero pivot was encountered during the computation.
-----------------------------------------	------------------------------------------------------

# Chapter 3.

## USING THE CUSOLVERMG API

### 3.1. General description

This chapter describes how to use the cuSolverMG library API. It is not a reference for the cuSolverMG API data types and functions; that is provided in subsequent chapters.

#### 3.1.1. Thread Safety

The library is thread-safe only if one cuSolverMG context per thread.

#### 3.1.2. Determinism

Currently all cuSolverMG API routines from a given toolkit version, generate the same bit-wise results when the following conditions are respected :

- ▶ all GPUs participating to the computation have the same compute-capabilities and the same number of SMs.
- ▶ the tiles size is kept the same between run.
- ▶ number of logical GPUs is kept the same. The order of GPUs are not important because all have the same compute-capabilities.

#### 3.1.3. tile strategy

The tiling strategy of cuSolverMG is compatible with ScaLAPACK. The current release only supports 1-D column block cyclic, column-major PACKED format.

Figure 1.a shows a partition of the matrix  $A$  of dimension  $\mathbf{M}_A$  by  $\mathbf{N}_A$ . Each column tile has  $\mathbf{T}_A$  columns. There are seven columns tiles, labeled as  $\bar{0}, \bar{1}, \bar{2}, \bar{3}, \bar{4}, \bar{5}, \bar{6}$ , distributed into three GPUs in a **cyclic** way, i.e. each GPU takes one column tile in turn. For example, GPU 0 has column tile  $\bar{0}, \bar{3}, \bar{6}$  (yellow tiles) and GPU 1 takes column tiles next to GPU 0 (blue tiles). Not all GPUs have the same number of tiles, in this example, GPU 0 has three tiles, others have only two tiles.

Figure 1.b shows two possible formats to store those column tiles locally in each GPU. Left side is called PACKED format and right side is UNPACKED format. PACKED format aggregates three column tiles in a contiguous memory block while UNPACKED format distributes these three column tiles into different memory blocks. The only difference between them is that PACKED format can have a big GEMM call instead of three GEMM calls in UNPACKED format. So theoretically speaking, PACKED format can deliver better performance than UNPACKED format. `cusolverMG` only supports PACKED format in the API. In order to achieve maximal performance, the user just needs to choose proper tile size  $T_A$  to partition the matrix, not too small, for example 256 or above is enough.

There is another parameter, called  $LLD_A$ , to control the leading dimension of the local matrix in each GPU.  $LLD_A$  must be greater or equal to  $M_A$ . The purpose of  $LLD_A$  is for better performance of GEMM. For small problem, GEMM is faster if  $LLD_A$  is power of 2. However for big problem,  $LLD_A$  does not show significant improvement. `cuSolverMG` only supports  $LLD_A=M_A$ .

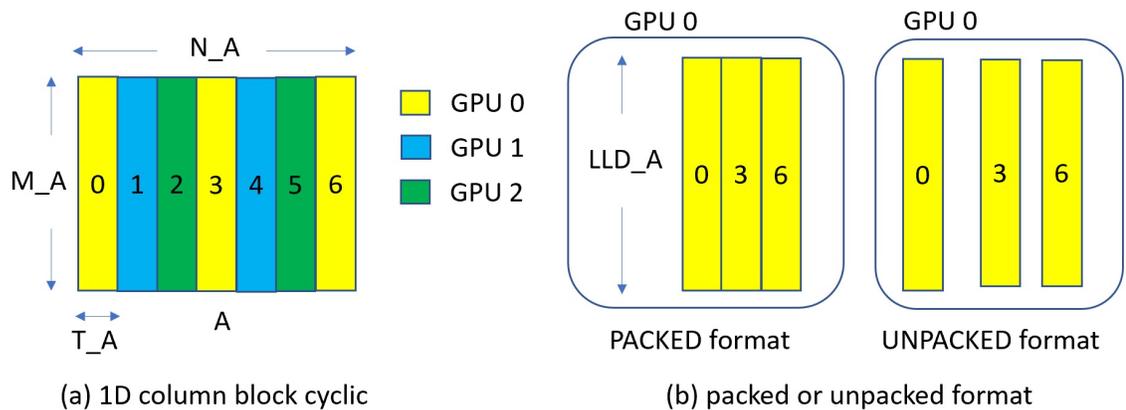


Figure 1 Example of `cusolverMG` tiling for 3 Gpus

The processing grid in `cuSolverMG` is a list of GPU IDs, similar to the process ID in `ScaLAPACK`. `cuSolverMG` only supports 1D column block cyclic, so only 1D grid is supported as well. Suppose `deviceId` is a list of GPU IDs, both `deviceId=1, 1, 1` and `deviceId=2, 1, 0` are valid. The former describes three logical devices are selected to run `cuSolverMG` routines, and all have the same physical ID, 0. The latter still uses three logical devices, but each has different physical ID. The current design only accepts 32 logical devices, that is, the length of `deviceId` is less or equal to 32. Figure 1 uses `deviceId=0, 1, 2`.

In practice, the matrix A is distributed into GPUs listed in `deviceId`. If the user chooses `deviceId=1, 1, 1`, all columns tile are located in GPU 1, this will limit the size of the problem because of memory capacity of one GPU. Besides, multiGPU routine adds extra overhead on data communication through off-chip bus, which has big performance impact if NVLINK is not supported or used. It would be faster to run on single GPU instead of running multiGPU version with devices of the same GPU ID.

### 3.1.4. Global matrix versus local matrix

To operate a submatrix of the matrix  $A$  is simple in dense linear algebra, just shift the pointer to the starting point of the submatrix relative to  $A$ . For example, `gesvd(10,10, A)` is SVD of  $A(0:9,0:9)$ . `gesvd(10,10, A + 5 + 2*lda )` is SVD of 10-by-10 submatrix starting at  $A(5,2)$ .

However it is not simple to operate on a submatrix of a distributed matrix because different starting point of the submatrix changes the distribution of the layout of that submatrix. **ScaLAPACK** introduces two parameters,  $IA$  and  $JA$ , to locate the submatrix. Figure 2 shows (global) matrix  $A$  of dimension  $M_A$  by  $N_A$ . The  $sub(A)$  is a  $M$  by  $N$  submatrix of  $A$ , starting at  $IA$  and  $JA$ . Please be aware that  $IA$  and  $JA$  are base-1.

Given a distributed matrix  $A$ , the user can compute eigenvalues of the submatrix  $sub(A)$  by either calling `syevd(A, IA, JA)` or gathering  $sub(A)$  to another distributed matrix  $B$  and calling `syevd(B, IB=1, JB=1)`.

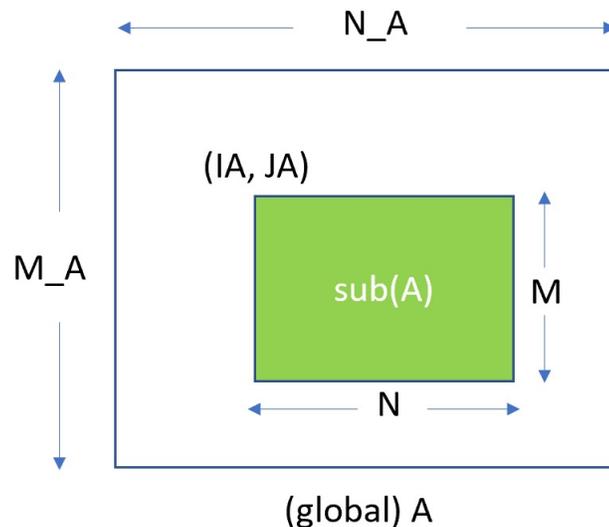


Figure 2 global matrix and local matrix

### 3.1.5. usage of `_bufferSize`

There is no `cudaMalloc` inside `cuSolverMG` library, the user must allocate the device workspace explicitly. The routine `xyz_bufferSize` is to query the size of workspace of the routine `xyz`, for example `xyz = syevd`. To make the API simple, `xyz_bufferSize` follows almost the same signature of `xyz` even it only depends on some parameters, for example, device pointer is not used to decide the size of workspace. In most cases, `xyz_bufferSize` is called in the beginning before actual device data (pointing by a device pointer) is prepared or before the device pointer is allocated. In such case, the user can pass null pointer to `xyz_bufferSize` without breaking the functionality.

`xyz_bufferSize` returns `bufferSize` for each device. The size is number of elements, not number of bytes.

### 3.1.6. synchronization

All routines are in synchronous (blocking call) manner. The data is ready after the routine. However the user has to prepare the distributed data before calling the routine. For example, if the user has multiple streams to setup the matrix, stream synchronization or device synchronization is necessary to guarantee distributed matrix is ready.

### 3.1.7. context switch

The user does not need to restore the device by `cudaSetDevice()` after each `cuSolverMG` call. All routines set the device back to what the caller has.

### 3.1.8. NVLINK

The peer-to-peer communication via NVLINK can dramatically reduce the overhead of data exchange among GPUs. `cuSolverMG` does not enable NVLINK implicitly, instead, it gives this option back to the user, not to interfere other libraries. The example code H.1 shows how to enable peer-to-peer communication.

## 3.2. cuSolverMG Types Reference

### 3.2.1. cuSolverMG Types

The `float`, `double`, `cuComplex`, and `cuDoubleComplex` data types are supported. The first two are standard C data types, while the last two are exported from `cuComplex.h`. In addition, `cuSolverMG` uses some familiar types from `cuBlas`.

### 3.2.2. cusolverMgHandle\_t

This is a pointer type to an opaque `cuSolverMG` context, in which the user must initialize by calling `cusolverMgCreate()` prior to calling any other library function. An un-initialized handle object will lead to unexpected behavior, including crashes of `cuSolverMG`. The handle created and returned by `cusolverMgCreate()` must be passed to every `cuSolverMG` function.

### 3.2.3. cusolverMgGridMapping\_t

The type indicates layout of grids.

Value	Meaning
<code>CUDALIBMG_GRID_MAPPING_ROW_MAJOR</code>	row-major ordering.
<code>CUDALIBMG_GRID_MAPPING_COL_MAJOR</code>	column-major ordering.

### 3.2.4. cudaLibMgGrid\_t

opaque structure of the distributed grid.

### 3.2.5. cudaLibMgMatrixDesc\_t

opaque structure of the distributed matrix descriptor.

## 3.3. Helper Function Reference

### 3.3.1. cusolverMgCreate()

```
cusolverStatus_t
cusolverMgCreate(cusolverMgHandle_t *handle)
```

This function initializes the cuSolverMG library and creates a handle on the cuSolverMG context. It must be called before any other cuSolverMG API function is invoked. It allocates hardware resources necessary for accessing the GPU.

#### Output

<code>handle</code>	the pointer to the handle to the cuSolverMG context.
---------------------	------------------------------------------------------

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the initialization succeeded.
<code>CUSOLVER_STATUS_ALLOC_FAILED</code>	the resources could not be allocated.

### 3.3.2. cusolverMgDestroy()

```
cusolverStatus_t
cusolverMgDestroy(cusolverMgHandle_t handle)
```

This function releases CPU-side resources used by the cuSolverMG library.

#### Input

<code>handle</code>	the handle to the cuSolverMG context.
---------------------	---------------------------------------

#### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the shutdown succeeded.
--------------------------------------	-------------------------

### 3.3.3. cusolverMgDeviceSelect()

```
cusolverStatus_t
cusolverMgDeviceSelect(
    cusolverMgHandle_t handle,
    int nbDevices,
    int deviceId[] )
```

This function registers a subset of devices (GPUs) to **cuSolverMG** handle. Such subset of devices is used in subsequent API calls. The array **deviceId** contains a list of logical device ID. The term **logical** means repeated device ID are permitted. For example, suppose the user has only one GPU in the system, say device 0, if he sets **deviceId=0,0,0**, then **cuSolverMG** treats them as three independent GPUs, one stream each, so concurrent kernel launches still hold. The current design only supports up to 32 logical devices.

#### Input

<b>handle</b>	the pointer to the handle to the cuSolverMG context.
<b>nbDevices</b>	the number of logical devices
<b>deviceId</b>	an integer array of size <b>nbDevices</b>

#### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the initialization succeeded.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	<b>nbDevices</b> must be greater than zero, and less or equal to 32.
<b>CUSOLVER_STATUS_ALLOC_FAILED</b>	the resources could not be allocated.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	internal error occurs when setting internal streams and events.

### 3.3.4. cusolverMgCreateDeviceGrid()

```
cusolverStatus_t
cusolverMgCreateDeviceGrid(
    cusolverMgGrid_t* grid,
    int32_t numRowsDevices,
    int32_t numColDevices,
    const int32_t deviceId[],
    cusolverMgGridMapping_t mapping)
```

This function setups grid of devices.

Only 1-D column block cyclic is supported, so **numRowDevices** must be equal to 1.

**WARNING:** **cusolverMgCreateDeviceGrid()** must be consistent with **cusolverMgDeviceSelect()**, i.e. **numColDevices** must be equal to **nbDevices** in **cusolverMgDeviceSelect()**.

parameter	Memory	In/out	Meaning
-----------	--------	--------	---------

<b>grid</b>	<b>host</b>	<b>output</b>	the pointer to the opaque structure.
<b>numRowDevices</b>	<b>host</b>	<b>input</b>	number of devices in the row.
<b>numColDevices</b>	<b>host</b>	<b>input</b>	number of devices in the column.
<b>deviceId</b>	<b>host</b>	<b>input</b>	integer array of size <b>numColDevices</b> , containing device IDs.
<b>mapping</b>	<b>host</b>	<b>input</b>	row-major or column-major ordering.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	<b>numColDevices</b> is not greater than 0. <b>numRowDevices</b> is not 1.

## 3.3.5. cusolverMgDestroyGrid()

```
cusolverStatus_t
cusolverMgDestroyGrid(
    cusolverMgGrid_t grid)
```

This function releases resources of a grid.

<b>parameter</b>	<b>Memory</b>	<b>In/out</b>	<b>Meaning</b>
<b>grid</b>	<b>host</b>	<b>input/output</b>	the pointer to the opaque structure.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
--------------------------------	---------------------------------------

## 3.3.6. cusolverMgCreateMatDescr()

```
cusolverStatus_t
cusolverMgCreateMatrixDesc(
    cusolverMgMatrixDesc_t * desc,
    int64_t numRows,
    int64_t numCols,
    int64_t rowBlockSize,
    int64_t colBlockSize,
    cudaDataType_t dataType,
    const cusolverMgGrid_t grid)
```

This function setups the matrix descriptor **desc**.

Only 1-D column block cyclic is supported, so **numRows** must be equal to **rowBlockSize**.

<b>parameter</b>	<b>Memory</b>	<b>In/out</b>	<b>Meaning</b>
<b>desc</b>	<b>host</b>	<b>output</b>	the matrix descriptor.
<b>numRows</b>	<b>host</b>	<b>input</b>	the number of rows of global A.
<b>numCols</b>	<b>host</b>	<b>input</b>	the number of columns of global A.

<b>rowBlockSize</b>	host	input	the number of rows per tile.
<b>colBlockSize</b>	host	input	the number of columns per tile.
<b>dataType</b>	host	input	data type of the matrix.
<b>grid</b>	host	input	the pointer to structure of grid.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	<b>numRows</b> , <b>numCols</b> , or <b>rowBlockSize</b> or <b>colBlockSize</b> is less than 0. <b>numRows</b> is not equal to <b>rowBlockSize</b> .

## 3.3.7. cusolverMgDestroyMatrixDesc()

```
cusolverStatus_t
cusolverMgDestroyMatrixDesc(
    cusolverMgMatrixDesc_t desc)
```

This function releases the matrix descriptor **desc**.

parameter	Memory	In/out	Meaning
<b>desc</b>	host	input/output	the matrix descriptor.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
--------------------------------	---------------------------------------

## 3.4. Dense Linear Solver Reference

This chapter describes linear solver API of cuSolverMG.

### 3.4.1. cusolverMgPOTrf()

The helper function below can calculate the sizes needed for pre-allocated buffer for **cusolverMgPOTrf**.

```
cusolverStatus_t
cusolverMgPOTrf_bufferSize(
    cusolverMgHandle_t handle,
    cublasFillMode_t uplo,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    cudaDataType computeType,
    int64_t *lwork)
```

The routine bellow

```
cusolverStatus_t
cusolverMgPotrf(
    cusolverMgHandle_t handle,
    cublasFillMode_t uplo,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    cudaDataType computeType,
    void *array_d_work[],
    int64_t lwork,
    int *info)
```

computes the Cholesky factorization of a Hermitian positive-definite matrix using the generic API interface.

**A** is a  $n \times n$  Hermitian matrix, only lower or upper part is meaningful. The input parameter **uplo** indicates which part of the matrix is used. The function would leave other part untouched.

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_LOWER**, only lower triangular part of **A** is processed, and replaced by lower triangular Cholesky factor **L**.

$$A = L * L^H$$

If input parameter **uplo** is **CUBLAS\_FILL\_MODE\_UPPER**, only upper triangular part of **A** is processed, and replaced by upper triangular Cholesky factor **U**.

$$A = U^H * U$$

The user has to provide device working space in **array\_d\_work**. **array\_d\_work** is a host pointer array of dimension **G**, where **G** is number of devices. **array\_d\_work[j]** is a device pointer pointing to a device memory in j-th device. The data type of **array\_d\_work[j]** is **computeType**. The size of **array\_d\_work[j]** is **lwork** which is number of elements per device, returned by **cusolverMgPotrf\_bufferSize()**.

If Cholesky factorization failed, i.e. some leading minor of **A** is not positive definite, or equivalently some diagonal elements of **L** or **U** is not a real number. The output parameter **info** would indicate smallest leading minor of **A** which is not positive definite.

If output parameter **info = -i** (less than zero), the **i-th** parameter is wrong (not counting handle).

The generic API has two different types, **dataTypeA** is data type of the matrix **A**, **computeType** is compute type of the operation and data type of the workspace (**array\_d\_work**) **descrA** contains **dataTypeA**, so there is no explicit parameter of **dataTypeA**. **cusolverMgPotrf** only supports the following four combinations.

Appendix I provides an example of **cusolverMgPotrf**.

**valid combination of data type and compute type**

DataTypeA	ComputeType	Meaning
CUDA_R_32F	CUDA_R_32F	SPOTRF
CUDA_R_64F	CUDA_R_64F	DPOTRF
CUDA_C_32F	CUDA_C_32F	CPOTRF
CUDA_C_64F	CUDA_C_64F	ZPOTRF

## API of potrf

parameter	Memory	In/out	Meaning
handle	host	input	handle to the cuSolverMg library context.
uplo	host	input	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced. Only <code>CUBLAS_FILL_MODE_LOWER</code> is supported.
N	host	input	number of rows and columns of matrix <b>sub(A)</b> .
array_d_A	host	in/out	a host pointer array of dimension <b>G</b> . It contains a distributed <type> array containing <b>sub(A)</b> of dimension $N * N$ . On exit, <b>sub(A)</b> contains the factors <b>L</b> or <b>U</b> .
IA	host	input	The row index in the global array <b>A</b> indicating the first row of <b>sub(A)</b> .
JA	host	input	The column index in the global array <b>A</b> indicating the first column of <b>sub(A)</b> .
descrA	host	input	matrix descriptor for the distributed matrix <b>A</b> .
computeType	host	input	Data type used for computation.
array_d_work	host	in/out	a host pointer array of dimension <b>G</b> . <code>array_d_work[j]</code> points to a device working space in <i>j</i> -th device, <type> array of size <b>lwork</b> .
lwork	host	input	size of <code>array_d_work[j]</code> , returned by <code>cusolverMgPOTrf_bufferSize</code> . <b>lwork</b> denotes number of elements, not number of bytes.
info	host	output	if <code>info = 0</code> , the Cholesky factorization is successful. if <code>info = -i</code> , the <i>i</i> -th parameter is wrong (not counting handle). if <code>info = i</code> , the leading minor of order <i>i</i> is not positive definite.

## Status Returned

CUSOLVER_STATUS_SUCCESS	the operation completed successfully.
CUSOLVER_STATUS_NOT_INITIALIZED	the library was not initialized.
CUSOLVER_STATUS_INVALID_VALUE	invalid parameters were passed ( $M, N < 0$ ).

<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 3.4.2. `cusolverMgPotrs()`

The helper function below can calculate the sizes needed for pre-allocated buffer for `cusolverMgPotrs`.

```
cusolverStatus_t
cusolverMgPotrs_bufferSize(
    cusolverMgHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    int nrhs,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    void *array_d_B[],
    int IB,
    int JB,
    cudaLibMgMatrixDesc_t descrB,
    cudaDataType computeType,
    int64_t *lwork )
```

The routine bellow

```
cusolverStatus_t
cusolverMgPotrs(
    cusolverMgHandle_t handle,
    cublasFillMode_t uplo,
    int n,
    int nrhs,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    void *array_d_B[],
    int IB,
    int JB,
    cudaLibMgMatrixDesc_t descrB,
    cudaDataType computeType,
    void *array_d_work[],
    int64_t lwork,
    int *info)
```

This function solves a system of linear equations

$$A * X = B$$

where **A** is a **n×n** Hermitian matrix, only lower or upper part is meaningful using the generic API interface. The input parameter **uplo** indicates which part of the matrix is used. The function would leave other part untouched.

If input parameter `uplo` is `CUBLAS_FILL_MODE_LOWER`, the matrix should **A** contain the lower triangular factor for Cholesky decomposition previously computed by `cusolverMgPotrf` routine.

$$A = L * L^H$$

If input parameter `uplo` is `CUBLAS_FILL_MODE_UPPER`, the matrix should **A** contain the upper triangular factor for Cholesky decomposition previously computed by `cusolverMgPotrf` routine.

$$A = U^H * U$$

The operation is in-place, i.e. matrix **B** contains the solution of the linear system on exit.

If output parameter `info` = `-i` (less than zero), the `i-th` parameter is wrong (not counting handle).

The user has to provide device working space in `array_d_work`. `array_d_work` is a host pointer array of dimension `G`, where `G` is number of devices. `array_d_work[j]` is a device pointer pointing to a device memory in `j-th` device. The data type of `array_d_work[j]` is `computeType`. The size of `array_d_work[j]` is `lwork` which is number of elements per device, returned by `cusolverMgPotrs_bufferSize()`.

If output parameter `info` = `-i` (less than zero), the `i-th` parameter is wrong (not counting handle).

The generic API has four different types, `dataTypeA` is data type of the matrix **A**, `dataTypeB` is data type of the matrix **B**, `computeType` is compute type of the operation and data type of the workspace (`array_d_work`) `descrA` contains `dataTypeA` and `descrB` contains `dataTypeB` and so there is no explicit parameter of `dataTypeA` and `dataTypeB`. `cusolverMgPotrs` only supports the following four combinations.

Appendix I provides an example of `cusolverMgPotrs`.

#### valid combination of data type and compute type

DataTypeA	DataTypeB	ComputeType	Meaning
CUDA_R_32F	CUDA_R_32F	CUDA_R_32F	SPOTRS
CUDA_R_64F	CUDA_R_64F	CUDA_R_64F	DPOTRS
CUDA_C_32F	CUDA_C_32F	CUDA_C_32F	CPOTRS
CUDA_C_64F	CUDA_C_64F	CUDA_C_64F	ZPOTRS

#### API of potrs

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the <code>cuSolverMg</code> library context.
<code>uplo</code>	host	input	indicates if matrix <b>A</b> lower or upper part is stored, the other part is not referenced. Only <code>CUBLAS_FILL_MODE_LOWER</code> is supported.
<code>N</code>	host	input	number of rows and columns of matrix <b>sub (A)</b> .

<code>NRHS</code>	host	input	number of columns of matrix <code>sub(A)</code> and <code>sub(B)</code> .
<code>array_d_A</code>	host	in/out	a host pointer array of dimension <code>G</code> . It contains a distributed <code>&lt;type&gt;</code> array containing <code>sub(A)</code> of dimension $M * N$ . On exit, <code>sub(A)</code> contains the factors <code>L</code> and <code>U</code> .
<code>IA</code>	host	input	The row index in the global array <code>A</code> indicating the first row of <code>sub(A)</code> .
<code>JA</code>	host	input	The column index in the global array <code>A</code> indicating the first column of <code>sub(A)</code> .
<code>descrA</code>	host	input	matrix descriptor for the distributed matrix <code>A</code> .
<code>array_d_B</code>	host	in/out	a host pointer array of dimension <code>G</code> . It contains a distributed <code>&lt;type&gt;</code> array containing <code>sub(B)</code> of dimension $N * NRHS$ . On exit, <code>sub(A)</code> contains the solution to the linear system.
<code>IB</code>	host	input	The row index in the global array <code>B</code> indicating the first row of <code>sub(B)</code> .
<code>JB</code>	host	input	The column index in the global array <code>B</code> indicating the first column of <code>sub(B)</code> .
<code>descrB</code>	host	input	matrix descriptor for the distributed matrix <code>B</code> .
<code>computeType</code>	host	input	Data type used for computation.
<code>array_d_work</code>	host	in/out	a host pointer array of dimension <code>G</code> . <code>array_d_work[j]</code> points to a device working space in <code>j</code> -th device, <code>&lt;type&gt;</code> array of size <code>lwork</code> .
<code>lwork</code>	host	input	size of <code>array_d_work[j]</code> , returned by <code>cusolverMgPotrs_bufferSize</code> . <code>lwork</code> denotes number of elements, not number of bytes.
<code>info</code>	host	output	if <code>info = 0</code> , the routine successful. if <code>info = -i</code> , the <code>i</code> -th parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_NOT_INITIALIZED</code>	the library was not initialized.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $M, N < 0$ ).
<code>CUSOLVER_STATUS_ARCH_MISMATCH</code>	the device only supports compute capability 2.0 and above.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 3.4.3. cusolverMgPotri()

The helper function below can calculate the sizes needed for pre-allocated buffer for `cusolverMgPotri`.

```
cusolverStatus_t
cusolverMgPotri_bufferSize(
    cusolverMgHandle_t handle,
    cublasFillMode_t uplo,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    cudaDataType computeType,
    int64_t *lwork)
```

The routine bellow

```
cusolverStatus_t
cusolverMgPotri(
    cusolverMgHandle_t handle,
    cublasFillMode_t uplo,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    cudaDataType computeType,
    void *array_d_work[],
    int64_t lwork,
    int *info)
```

This function computes the inverse of an Hermitian positive-definite matrix **A** using the Cholesky factorization

$$A = L * L^H = U^H * U$$

computed by `cusolverMgPotrf()`.

If the input parameter `uplo` is `CUBLAS_FILL_MODE_LOWER`, on input, matrix **A** contains the lower triangular factor of **A** computed by `cusolverMgPotrf`. Only lower triangular part of **A** is processed, and replaced the by lower triangular part of the inverse of **A**.

If the input parameter `uplo` is `CUBLAS_FILL_MODE_UPPER`, on input, matrix **A** contains the upper triangular factor of **A** computed by `cusolverMgPotrf`. Only upper triangular part of **A** is processed, and replaced the by upper triangular part of the inverse of **A**.

The user has to provide device working space in `array_d_work`. `array_d_work` is a host pointer array of dimension **G**, where **G** is number of devices. `array_d_work[j]` is a device pointer pointing to a device memory in j-th device. The data type of `array_d_work[j]` is `computeType`. The size of `array_d_work[j]` is `lwork` which is number of elements per device, returned by `cusolverMgPotri_bufferSize()`.

If the computation of the inverse fails, i.e. some leading minor of  $\mathbf{L}$  or  $\mathbf{U}$ , is null, the output parameter `info` would indicate the smallest leading minor of  $\mathbf{L}$  or  $\mathbf{U}$  which is not positive definite.

If the output parameter `info = -i` (less than zero), the `i-th` parameter is wrong (not counting the handle).

The generic API has two different types, `dataTypeA` is data type of the matrix  $\mathbf{A}$ , `computeType` is compute type of the operation and data type of the workspace (`array_d_work`) `descrA` contains `dataTypeA`, so there is no explicit parameter of `dataTypeA`. `cusolverMgPotri` only supports the following four combinations.

Appendix I provides an example of `cusolverMgPotri`.

#### valid combination of data type and compute type

DataTypeA	ComputeType	Meaning
CUDA_R_32F	CUDA_R_32F	SPOTRI
CUDA_R_64F	CUDA_R_64F	DPOTRI
CUDA_C_32F	CUDA_C_32F	CPOTRI
CUDA_C_64F	CUDA_C_64F	ZPOTRI

#### API of potrf

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverMg library context.
<code>uplo</code>	host	input	indicates if matrix $\mathbf{A}$ lower or upper part is stored, the other part is not referenced. Only <code>CUBLAS_FILL_MODE_LOWER</code> is supported.
<code>N</code>	host	input	number of rows and columns of matrix <code>sub(A)</code> .
<code>array_d_A</code>	host	in/out	a host pointer array of dimension <code>G</code> . It contains a distributed <code>&lt;type&gt;</code> array containing <code>sub(A)</code> of dimension <code>N * N</code> . On exit, <code>sub(A)</code> contains the upper or lower triangular part of the inverse of $\mathbf{A}$ depending on the value of <code>uplo</code> argument.
<code>IA</code>	host	input	The row index in the global array $\mathbf{A}$ indicating the first row of <code>sub(A)</code> .
<code>JA</code>	host	input	The column index in the global array $\mathbf{A}$ indicating the first column of <code>sub(A)</code> .
<code>descrA</code>	host	input	matrix descriptor for the distributed matrix $\mathbf{A}$ .
<code>computeType</code>	host	input	Data type used for computation.
<code>array_d_work</code>	host	in/out	a host pointer array of dimension <code>G</code> . <code>array_d_work[j]</code> points to a device working space in <code>j-th</code> device, <code>&lt;type&gt;</code> array of size <code>lwork</code> .

<b>lwork</b>	<b>host</b>	<b>input</b>	Size of <code>array_d_work[j]</code> , returned by <code>cusolverMgPotri_bufferSize</code> . <code>lwork</code> denotes number of elements, not number of bytes.
<b>info</b>	<b>host</b>	<b>output</b>	if <code>info = 0</code> , the Cholesky factorization is successful. if <code>info = -i</code> , the <i>i</i> -th parameter is wrong (not counting handle). if <code>info = i</code> , the leading minor of order <i>i</i> is zero.

### Status Returned

<b>CUSOLVER_STATUS_SUCCESS</b>	the operation completed successfully.
<b>CUSOLVER_STATUS_NOT_INITIALIZED</b>	the library was not initialized.
<b>CUSOLVER_STATUS_INVALID_VALUE</b>	invalid parameters were passed ( <code>M,N&lt;0</code> ).
<b>CUSOLVER_STATUS_ARCH_MISMATCH</b>	the device only supports compute capability 2.0 and above.
<b>CUSOLVER_STATUS_INTERNAL_ERROR</b>	an internal operation failed.

## 3.4.4. cusolverMgGetrf()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverMgGetrf_bufferSize(
    cusolverMgHandle_t handle,
    int M,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    int *array_d_IPIV[],
    cudaDataType_t computeType,
    int64_t *lwork);
```

```
cusolverStatus_t
cusolverMgGetrf(
    cusolverMgHandle_t handle,
    int M,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    int *array_d_IPIV[],
    cudaDataType_t computeType,
    void *array_d_work[],
    int64_t lwork,
    int *info );
```

This function computes the LU factorization of a  $\mathbf{M} \times \mathbf{N}$  matrix

$$P * A = L * U$$

where **A** is a **M**×**N** matrix, **P** is a permutation matrix, **L** is a lower triangular matrix with unit diagonal, and **U** is an upper triangular matrix.

The user has to provide device working space in **array\_d\_work**. **array\_d\_work** is a host pointer array of dimension **G**, where **G** is number of devices. **array\_d\_work[j]** is a device pointer pointing to a device memory in **j**-th device. The data type of **array\_d\_work[j]** is **computeType**. The size of **array\_d\_work[j]** is **lwork** which is number of elements per device, returned by **cusolverMgGetrf\_bufferSize()**.

If LU factorization failed, i.e. matrix **A** (**U**) is singular, The output parameter **info=i** indicates **U(i,i) = 0**.

If output parameter **info = -i** (less than zero), the **i**-th parameter is wrong (not counting handle).

If **array\_d\_IPIV** is null, no pivoting is performed. The factorization is **A=L\*U**, which is not numerically stable.

**array\_d\_IPIV** must be consistent with **array\_d\_A**, i.e. **JA** is the first column of **sub(A)**, also the first column of **sub(IPIV)**.

No matter LU factorization failed or not, the output parameter **array\_d\_IPIV** contains pivoting sequence, row **i** is interchanged with row **array\_d\_IPIV(i)**.

The generic API has three different types, **dataTypeA** is data type of the matrix **A**, **computeType** is compute type of the operation and data type of the workspace (**array\_d\_work**) **descrA** contains **dataTypeA**, so there is no explicit parameter of **dataTypeA**. **cusolverMgGetrf** only supports the following four combinations.

Appendix I provides an example of **cusolverMgGetrf**.

#### valid combination of data type and compute type

DataTypeA	ComputeType	Meaning
CUDA_R_32F	CUDA_R_32F	SGETRF
CUDA_R_64F	CUDA_R_64F	DGETRF
CUDA_C_32F	CUDA_C_32F	CGETRF
CUDA_C_64F	CUDA_C_64F	ZGETRF

Remark 1: tile size **TA** must be less or equal to 512.

#### API of getrf

parameter	Memory	In/out	Meaning
<b>handle</b>	host	input	handle to the cuSolverMg library context.
<b>M</b>	host	input	number of rows of matrix <b>sub(A)</b> .
<b>N</b>	host	input	number of columns of matrix <b>sub(A)</b> .
<b>array_d_A</b>	host	in/out	a host pointer array of dimension <b>G</b> . It contains a distributed <type> array containing <b>sub(A)</b> of dimension <b>M * N</b> . On exit, <b>sub(A)</b> contains the factors <b>L</b> and <b>U</b> .

<code>IA</code>	<code>host</code>	<code>input</code>	The row index in the global array <code>A</code> indicating the first row of <code>sub(A)</code> .
<code>JA</code>	<code>host</code>	<code>input</code>	The column index in the global array <code>A</code> indicating the first column of <code>sub(A)</code> .
<code>descrA</code>	<code>host</code>	<code>input</code>	matrix descriptor for the distributed matrix <code>A</code> .
<code>array_d_IPIV</code>	<code>host</code>	<code>output</code>	a host pointer array of dimension <code>G</code> . it contains a distributed integer array containing <code>sub(IPIV)</code> of size <code>min(M,N)</code> . <code>sub(IPIV)</code> contains pivot indices.
<code>computeType</code>	<code>host</code>	<code>input</code>	Data type used for computation.
<code>array_d_work</code>	<code>host</code>	<code>in/out</code>	a host pointer array of dimension <code>G</code> . <code>array_d_work[j]</code> points to a device working space in <code>j</code> -th device, <code>&lt;type&gt;</code> array of size <code>lwork</code> .
<code>lwork</code>	<code>host</code>	<code>input</code>	size of <code>array_d_work[j]</code> , returned by <code>cusolverMgGetrf_bufferSize</code> . <code>lwork</code> denotes number of elements, not number of bytes.
<code>info</code>	<code>host</code>	<code>output</code>	if <code>info = 0</code> , the LU factorization is successful. if <code>info = -i</code> , the <code>i</code> -th parameter is wrong (not counting handle). if <code>info = i</code> , the <code>U(i,i) = 0</code> .

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>M,N &lt; 0</code> ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

### 3.4.5. cusolverMgGetrs()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverMgGetrs_bufferSize(
    cusolverMgHandle_t handle,
    cublasOperation_t TRANS,
    int N,
    int NRHS,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    int *array_d_IPIV[],
    void *array_d_B[],
    int IB,
    int JB,
    cudaLibMgMatrixDesc_t descrB,
    cudaDataType_t computeType,
    int64_t *lwork);
```

```
cusolverStatus_t
cusolverMgGetrs(
    cusolverMgHandle_t handle,
    cublasOperation_t TRANS,
    int N,
    int NRHS,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    int *array_d_IPIV[],
    void *array_d_B[],
    int IB,
    int JB,
    cudaLibMgMatrixDesc_t descrB,
    cudaDataType_t computeType,
    void *array_d_work[],
    int64_t lwork,
    int *info );
```

This function solves a linear system of multiple right-hand sides

$$\text{op}(\mathbf{A}) * \mathbf{X} = \mathbf{B}$$

where  $\mathbf{A}$  is a  $\mathbf{N} \times \mathbf{N}$  matrix, and was LU-factored by `getrf`, that is, lower triangular part of  $\mathbf{A}$  is  $\mathbf{L}$ , and upper triangular part (including diagonal elements) of  $\mathbf{A}$  is  $\mathbf{U}$ .  $\mathbf{B}$  is a  $\mathbf{N} \times \mathbf{NRHS}$  right-hand side matrix. The solution matrix  $\mathbf{X}$  overwrites the right-hand-side matrix  $\mathbf{B}$ .

The input parameter **TRANS** is defined by

$$\text{op}(\mathbf{A}) = \begin{cases} \mathbf{A} & \text{if TRANS == CUBLAS_OP_N} \\ \mathbf{A}^T & \text{if TRANS == CUBLAS_OP_T} \\ \mathbf{A}^H & \text{if TRANS == CUBLAS_OP_C} \end{cases}$$

The user has to provide device working space in `array_d_work`. `array_d_work` is a host pointer array of dimension `G`, where `G` is number of devices. `array_d_work[j]` is a device pointer pointing to a device memory in `j`-th device. The data type of `array_d_work[j]` is `computeType`. The size of `array_d_work[j]` is `lwork` which is number of elements per device, returned by `cusolverMgGetrs_bufferSize()`.

If `array_d_IPIV` is null, no pivoting is performed. Otherwise, `array_d_IPIV` is an output of `getrf`. It contains pivot indices, which are used to permute right-hand sides.

If output parameter `info = -i` (less than zero), the `i`-th parameter is wrong (not counting handle).

The generic API has three different types, `dataTypeA` is data type of the matrix `A`, `dataTypeB` is data type of the matrix `B`, and `computeType` is compute type of the operation and data type of the workspace (`array_d_work`). `descrA` contains `dataTypeA`, so there is no explicit parameter of `dataTypeA`. `descrB` contains `dataTypeB`, so there is no explicit parameter of `dataTypeB`. `cusolverMgGetrs` only supports the following four combinations.

#### valid combination of data type and compute type

DataTypeA	DataTypeB	ComputeType	Meaning
CUDA_R_32F	CUDA_R_32F	CUDA_R_32F	SGETRS
CUDA_R_64F	CUDA_R_64F	CUDA_R_64F	DGETRS
CUDA_C_32F	CUDA_C_32F	CUDA_C_32F	CGETRS
CUDA_C_64F	CUDA_C_64F	CUDA_C_64F	ZGETRS

Remark 1: tile size `TA` must be less or equal to 512.

Remark 2: only support `TRANS=CUBLAS_OP_N`.

Appendix I provides an example of `cusolverMgGetrs`.

#### API of getrs

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverMG library context.
<code>TRANS</code>	host	input	operation <code>op</code> ( <code>A</code> ) that is non- or (conj.) transpose.
<code>N</code>	host	input	number of rows and columns of matrix <code>sub(A)</code> .
<code>NRHS</code>	host	input	number of columns of matrix <code>sub(B)</code> .
<code>array_d_A</code>	host	input	a host pointer array of dimension <code>G</code> . It contains a distributed <type> array containing <code>sub(A)</code> of dimension <code>M * N</code> . <code>sub(A)</code> contains the factors <code>L</code> and <code>U</code> .
<code>IA</code>	host	input	The row index in the global array <code>A</code> indicating the first row of <code>sub(A)</code> .
<code>JA</code>	host	input	The column index in the global array <code>A</code> indicating the first column of <code>sub(A)</code> .

<code>descrA</code>	<code>host</code>	<code>input</code>	matrix descriptor for the distributed matrix A.
<code>array_d_IPIV</code>	<code>host</code>	<code>input</code>	a host pointer array of dimension <code>G</code> . it contains a distributed integer array containing <code>sub(IPIV)</code> of dimension <code>min(M,N)</code> . <code>sub(IPIV)</code> contains pivot indices.
<code>array_d_B</code>	<code>host</code>	<code>in/out</code>	a host pointer array of dimension <code>G</code> . It contains a distributed <code>&lt;type&gt;</code> array containing <code>sub(B)</code> of dimension <code>N * NRHS</code> .
<code>IB</code>	<code>host</code>	<code>input</code>	The row index in the global array B indicating the first row of <code>sub(B)</code> .
<code>JB</code>	<code>host</code>	<code>input</code>	The column index in the global array B indicating the first column of <code>sub(B)</code> .
<code>descrB</code>	<code>host</code>	<code>input</code>	matrix descriptor for the distributed matrix B.
<code>computeType</code>	<code>host</code>	<code>input</code>	Data type used for computation.
<code>array_d_work</code>	<code>host</code>	<code>in/out</code>	a host pointer array of dimension <code>G</code> . <code>array_d_work[j]</code> points to a device working space in <code>j</code> -th device, <code>&lt;type&gt;</code> array of size <code>lwork</code> .
<code>lwork</code>	<code>host</code>	<code>input</code>	size of <code>array_d_work[j]</code> , returned by <code>cusolverMgGetrs_bufferSize</code> . <code>lwork</code> denotes number of elements, not number of bytes.
<code>info</code>	<code>host</code>	<code>output</code>	if <code>info = 0</code> , the operation is successful. if <code>info = -i</code> , the <code>i</code> -th parameter is wrong (not counting handle).

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( <code>N&lt;0</code> or <code>NRHS&lt;0</code> ).
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

## 3.5. Dense Eigenvalue Solver Reference

This chapter describes eigenvalue solver API of `cuSolverMG`.

### 3.5.1. cusolverMgSyevd()

The helper functions below can calculate the sizes needed for pre-allocated buffer.

```
cusolverStatus_t
cusolverMgSyevd_bufferSize(
    cusolverMgHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    void *W,
    cudaDataType_t dataTypeW,
    cudaDataType_t computeType,
    int64_t *lwork
);
```

```
cusolverStatus_t
cusolverMgSyevd(
    cusolverMgHandle_t handle,
    cusolverEigMode_t jobz,
    cublasFillMode_t uplo,
    int N,
    void *array_d_A[],
    int IA,
    int JA,
    cudaLibMgMatrixDesc_t descrA,
    void *W,
    cudaDataType_t dataTypeW,
    cudaDataType_t computeType,
    void *array_d_work[],
    int64_t lwork,
    int *info );
```

This function computes eigenvalues and eigenvectors of a symmetric (Hermitian)  $\mathbf{N} \times \mathbf{N}$  matrix  $\mathbf{A}$ . The standard symmetric eigenvalue problem is

$$\mathbf{A} * \mathbf{V} = \mathbf{V} * \mathbf{\Lambda}$$

where  $\mathbf{\Lambda}$  is a real  $\mathbf{N} \times \mathbf{N}$  diagonal matrix.  $\mathbf{V}$  is an  $\mathbf{N} \times \mathbf{N}$  unitary matrix. The diagonal elements of  $\mathbf{\Lambda}$  are the eigenvalues of  $\mathbf{A}$  in ascending order.

**cusolverMgSyevd** returns the eigenvalues in  $\mathbf{W}$  and overwrites the eigenvectors in  $\mathbf{A}$ .  $\mathbf{W}$  is a host  $1 \times \mathbf{N}$  vector.

The generic API has three different types, **dataTypeA** is data type of the matrix  $\mathbf{A}$ , **dataTypeW** is data type of the vector  $\mathbf{W}$ , and **computeType** is compute type of the operation and data type of the workspace (**array\_d\_work**) **descrA** contains **dataTypeA**, so there is no explicit parameter of **dataTypeA**. **cusolverMgSyevd** only supports the following four combinations.

**valid combination of data type and compute type**

DataTypeA	DataTypeW	ComputeType	Meaning
CUDA_R_32F	CUDA_R_32F	CUDA_R_32F	SSYEVD
CUDA_R_64F	CUDA_R_64F	CUDA_R_64F	DSYEVD

CUDA_C_32F	CUDA_R_32F	CUDA_C_32F	CHEEVD
CUDA_C_64F	CUDA_R_64F	CUDA_C_64F	ZHEEVD

The user has to provide device working space in `array_d_work`. `array_d_work` is a host pointer array of dimension `G`, where `G` is number of devices. `array_d_work[j]` is a device pointer pointing to a device memory in `j`-th device. The data type of `array_d_work[j]` is `computeType`. The size of `array_d_work[j]` is `lwork` which is number of elements per device, returned by `cusolverMgSyevd_bufferSize()`.

`array_d_A` is also a host pointer array of dimension `G`. `array_d_A[j]` is a device pointer pointing to a device memory in `j`-th device. The data type of `array_d_A[j]` is `dataTypeA`. The size of `array_d_A[j]` is about `N*TA*(blocks per device)`. The user has to prepare `array_d_A` manually (please check the samples in Appendix H).

If output parameter `info = -i` (less than zero), the `i`-th parameter is wrong (not counting handle). If `info = i` (greater than zero), `i` off-diagonal elements of an intermediate tridiagonal form did not converge to zero.

if `jobz = CUSOLVER_EIG_MODE_VECTOR`, `A` contains the orthonormal eigenvectors of the matrix `A`. The eigenvectors are computed by a divide and conquer algorithm.

Remark 1: only `CUBLAS_FILL_MODE_LOWER` is supported, so the user has to prepare lower triangle of `A`.

Remark 2: only `IA=1` and `JA=1` are supported.

Remark 3: tile size `TA` must be less or equal to 1024. To achieve best performance, `TA` should be 256 or 512.

Appendix H provides three examples of `cusolverMgSyevd`.

### API of `syevd`

parameter	Memory	In/out	Meaning
<code>handle</code>	host	input	handle to the cuSolverMG library context.
<code>jobz</code>	host	input	specifies options to either compute eigenvalue only or compute eigen-pair: <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> : Compute eigenvalues only; <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> : Compute eigenvalues and eigenvectors.
<code>uplo</code>	host	input	specifies which part of <code>A</code> is stored. <code>uplo = CUBLAS_FILL_MODE_LOWER</code> : Lower triangle of <code>A</code> is stored. <code>uplo = CUBLAS_FILL_MODE_UPPER</code> : Upper triangle of <code>A</code> is stored. Only <code>CUBLAS_FILL_MODE_LOWER</code> is supported.
<code>N</code>	host	input	number of rows (or columns) of matrix <code>sub(A)</code> .
<code>array_d_A</code>	host	in/out	a host pointer array of dimension <code>G</code> . It contains a distributed <type> array containing <code>sub(A)</code> of dimension <code>N * N</code> . If <code>uplo = CUBLAS_FILL_MODE_UPPER</code> , the leading <code>N</code> -by- <code>N</code> upper triangular part

			of $\text{sub}(\mathbf{A})$ contains the upper triangular part of the matrix $\text{sub}(\mathbf{A})$ . If <code>uplo = CUBLAS_FILL_MODE_LOWER</code> , the leading N-by-N lower triangular part of $\text{sub}(\mathbf{A})$ contains the lower triangular part of the matrix $\text{sub}(\mathbf{A})$ . On exit, if <code>jobz = CUSOLVER_EIG_MODE_VECTOR</code> , and <code>info = 0</code> , $\text{sub}(\mathbf{A})$ contains the orthonormal eigenvectors of the matrix $\text{sub}(\mathbf{A})$ . If <code>jobz = CUSOLVER_EIG_MODE_NOVECTOR</code> , the contents of $\mathbf{A}$ are destroyed.
<code>IA</code>	host	input	The row index in the global array $\mathbf{A}$ indicating the first row of $\text{sub}(\mathbf{A})$ .
<code>JA</code>	host	input	The column index in the global array $\mathbf{A}$ indicating the first column of $\text{sub}(\mathbf{A})$ .
<code>descrA</code>	host	input	matrix descriptor for the distributed matrix $\mathbf{A}$ .
<code>W</code>	host	output	a real array of dimension $\mathbf{N}$ . The eigenvalue values of $\text{sub}(\mathbf{A})$ , in ascending order $i$ e, sorted so that $W(i) \leq W(i+1)$ .
<code>dataTypeW</code>	host	input	Data type of the vector $W$ .
<code>computeType</code>	host	input	Data type used for computation.
<code>array_d_work</code>	host	in/out	a host pointer array of dimension $\mathbf{G}$ . <code>array_d_work[j]</code> points to a device working space in $j$ -th device, <code>&lt;type&gt;</code> array of size <code>lwork</code> .
<code>lwork</code>	host	input	size of <code>array_d_work[j]</code> , returned by <code>cusolverMgSyevd_bufferSize</code> . <code>lwork</code> denotes number of elements, not number of bytes.
<code>info</code>	host	output	if <code>info = 0</code> , the operation is successful. if <code>info = -i</code> , the $i$ -th parameter is wrong (not counting handle). if <code>info = i (&gt; 0)</code> , <code>info</code> indicates $i$ off-diagonal elements of an intermediate tridiagonal form did not converge to zero;

### Status Returned

<code>CUSOLVER_STATUS_SUCCESS</code>	the operation completed successfully.
<code>CUSOLVER_STATUS_INVALID_VALUE</code>	invalid parameters were passed ( $\mathbf{N} < 0$ , or $\text{lda} < \max(1, \mathbf{N})$ , or <code>jobz</code> is not <code>CUSOLVER_EIG_MODE_NOVECTOR</code> or <code>CUSOLVER_EIG_MODE_VECTOR</code> , or <code>uplo</code> is not <code>CUBLAS_FILL_MODE_LOWER</code> , or <code>IA</code> and <code>JA</code> are not 1, or $\mathbf{N}$ is bigger than dimension of global $\mathbf{A}$ , or the combination of <code>dataType</code> and <code>computeType</code> is not valid.
<code>CUSOLVER_STATUS_INTERNAL_ERROR</code>	an internal operation failed.

# Appendix A. CUSOLVERRF EXAMPLES

## A.1. cuSolverRF In-memory Example

This is an example in the C programming language of how to use the standard routines in the cuSolverRF library. We focus on solving the set of linear systems

$$A_i x_i = f_i$$

but we change the indexing from one- to zero-based to follow the C programming language. The example begins with the usual includes and main()

```
#include <stdio.h>
#include <stdlib.h>
#include <cuda_runtime.h>
#include "cusolverRf.h"

#define TEST_PASSED 0
#define TEST_FAILED 1

int main (void){
    /* matrix A */
    int n;
    int nnzA;
    int *Ap=NULL;
    int *Ai=NULL;
    double *Ax=NULL;
    int *d_Ap=NULL;
    int *d_Ai=NULL;
    double *d_rAx=NULL;
    /* matrices L and U */
    int nnzL, nnzU;
    int *Lp=NULL;
    int *Li=NULL;
    double* Lx=NULL;
    int *Up=NULL;
    int *Ui=NULL;
    double* Ux=NULL;
    /* reordering matrices */
    int *P=NULL;
    int *Q=NULL;
    int *d_P=NULL;
    int *d_Q=NULL;
    /* solution and rhs */
    int nrhs; // # of rhs for each system (currently only =1 is supported)
    double *d_X=NULL;
    double *d_T=NULL;
    /* cuda */
    cudaError_t cudaStatus;
    /* cuolverRf */
    cusolverRfHandle_t gH=NULL;
    cusolverStatus_t status;
    /* host sparse direct solver */
    /* ... */
    /* other variables */
    int tnnzL, tnnzU;
    int *tLp=NULL;
    int *tLi=NULL;
    double *tLx=NULL;
    int *tUp=NULL;
    int *tUi=NULL;
    double *tUx=NULL;
    double t1, t2;
```

Then we initialize the library.

```

/* ASSUMPTION: recall that we are solving a set of linear systems
A_{i} x_{i} = f_{i} for i=0,...,k-1
where the sparsity pattern of the coefficient matrices A_{i}
as well as the reordering to minimize fill-in and the pivoting
used during the LU factorization remain the same. */

/* Step 1: solve the first linear system (i=0) on the host,
using host sparse direct solver, which involves
full LU factorization and solve. */
/* ... */

/* Step 2: interface to the library by extracting the following
information from the first solve:
a) triangular factors L and U
b) pivoting and reordering permutations P and Q
c) also, allocate all the necessary memory */
/* ... */

/* Step 3: use the library to solve subsequent (i=1,...,k-1) linear systems
a) the library setup (called only once) */
//create handle
status = cusolverRfCreate(&gH);
if (status != CUSOLVER_STATUS_SUCCESS){
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}

//set fast mode
status = cusolverRfSetResetValuesFastMode(gH, GLU_RESET_VALUES_FAST_MODE_ON);
if (status != CUSOLVER_STATUS_SUCCESS){
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}

```

## Call refactorization and solve.

```

//assemble internal data structures (you should use the coefficient matrix A
//corresponding to the second (i=1) linear system in this call)
t1 = cusolver_test_seconds();
status = cusolverRfSetupHost(n, nnzA, Ap, Ai, Ax,
                             nnzL, Lp, Li, Lx, nnzU, Up, Ui, Ux, P, Q, gH);
cudaStatus = cudaDeviceSynchronize();
t2 = cusolver_test_seconds();
if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess)) {
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
printf("cusolverRfSetupHost time = %f (s)\n", t2-t1);

//analyze available parallelism
t1 = cusolver_test_seconds();
status = cusolverRfAnalyze(gH);
cudaStatus = cudaDeviceSynchronize();
t2 = cusolver_test_seconds();
if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess)) {
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
printf("cusolverRfAnalyze time = %f (s)\n", t2-t1);

/* b) The library subsequent (i=1,...,k-1) LU re-factorization
and solve (called multiple times). */
for (i=1; i<k; i++){
    //LU re-factorization
    t1 = cusolver_test_seconds();
    status = cusolverRfRefactor(gH);
    cudaStatus = cudaDeviceSynchronize();
    t2 = cusolver_test_seconds();
    if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess))
    {
        printf ("[cusolverRF status %d]\n",status);
        return TEST_FAILED;
    }
    printf("cuSolverReRefactor time = %f (s)\n", t2-t1);

    //forward and backward solve
    t1 = cusolver_test_seconds();
    status = cusolverRfSolve(gH, d_P, d_Q, nrhs, d_T, n, d_X, n);
    cudaStatus = cudaDeviceSynchronize();
    t2 = cusolver_test_seconds();
    if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess))
    {
        printf ("[cusolverRf status %d]\n",status);
        return TEST_FAILED;
    }
    printf("cusolverRfSolve time = %f (s)\n", t2-t1);
}

```

Extract the results and return.

```

// extract the factors (if needed)
status = cusolverRfExtractSplitFactorsHost(gH, &tnnzL, &tLp, &tLi,
&tLx,
                                &tnnzU, &tUp, &tUi, &tUx);
if(status != CUSOLVER_STATUS_SUCCESS){
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
/*
//print
int row, j;
printf("printing L\n");
for (row=0; row<n; row++){
    for (j=tLp[row]; j<tLp[row+1]; j++){
        printf("%d,%d,%f\n",row,tLi[j],tLx[j]);
    }
}
printf("printing U\n");
for (row=0; row<n; row++){
    for (j=tUp[row]; j<tUp[row+1]; j++){
        printf("%d,%d,%f\n",row,tUi[j],tUx[j]);
    }
}
*/

/* perform any other operations based on the solution */
/* ... */

/* check if done */
/* ... */

/* proceed to solve the next linear system */
// update the coefficient matrix using reset values
// (assuming that the new linear system, in other words,
// new values are already on the GPU in the array d_rAx)
t1 = cusolver_test_seconds();
status = cusolverRfResetValues(n,nnzA,d_Ap,d_Ai,d_rAx,d_P,d_Q,gH);
cudaStatus = cudaDeviceSynchronize();
t2 = cusolver_test_seconds();
if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess))
{
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
printf("cusolverRfResetValues time = %f (s)\n", t2-t1);
}

/* free memory and exit */
/* ... */
return TEST_PASSED;
}

```

## A.2. cuSolverRF-batch Example

This chapter provides an example in the C programming language of how to use the batched routines in the cuSolverRF library. We focus on solving the set of linear systems

$$A_i x_i = f_i$$

but we change the indexing from one- to zero-based to follow the C programming language. The first part is the usual includes and main definition

```
#include <stdio.h>
#include <stdlib.h>
#include <cuda_runtime.h>
#include "cusolverRf.h"

#define TEST_PASSED 0
#define TEST_FAILED 1

int main (void){
    /* matrix A */
    int batchSize;
    int n;
    int nnzA;
    int *Ap=NULL;
    int *Ai=NULL;
    //array of pointers to the values of each matrix in the batch (of size
    //batchSize) on the host
    double **Ax_array=NULL;
    //For example, if Ax_batch is the array (of size batchSize*nnzA) containing
    //the values of each matrix in the batch written contiguously one matrix
    //after another on the host, then Ax_array[j] = &Ax_batch[nnzA*j];
    //for j=0,...,batchSize-1.
    double *Ax_batch=NULL;
    int *d_Ap=NULL;
    int *d_Ai=NULL;
    //array of pointers to the values of each matrix in the batch (of size
    //batchSize) on the device
    double **d_Ax_array=NULL;
    //For example, if d_Ax_batch is the array (of size batchSize*nnzA)
    containing
    //the values of each matrix in the batch written contiguously one matrix
    //after another on the device, then d_Ax_array[j] = &d_Ax_batch[nnzA*j];
    //for j=0,...,batchSize-1.
    double *d_Ax_batch=NULL;
    /* matrices L and U */
    int nnzL, nnzU;
    int *Lp=NULL;
    int *Li=NULL;
    double* Lx=NULL;
    int *Up=NULL;
    int *Ui=NULL;
    double* Ux=NULL;
    /* reordering matrices */
    int *P=NULL;
    int *Q=NULL;
    int *d_P=NULL;
    int *d_Q=NULL;
}
```

Next we initialize the data needed and the create library handles

```

/* solution and rhs */
int nrhs; //# of rhs for each system (currently only =1 is supported)
//temporary storage (of size 2*batchSize*n*nrhs)
double *d_T=NULL;
//array (of size batchSize*n*nrhs) containing the values of each rhs in
//the batch written contiguously one rhs after another on the device
double **d_X_array=NULL;
//array (of size batchSize*n*nrhs) containing the values of each rhs in
//the batch written contiguously one rhs after another on the host
double **X_array=NULL;
/* cuda */
cudaError_t cudaStatus;
/* cusolverRf */
cusolverRfHandle_t gH=NULL;
cusolverStatus_t status;
/* host sparse direct solver */
...
/* other variables */
double t1, t2;

/* ASSUMPTION:
recall that we are solving a batch of linear systems
 $A_{\{j\}} x_{\{j\}} = f_{\{j\}}$  for  $j=0, \dots, \text{batchSize}-1$ 
where the sparsity pattern of the coefficient matrices  $A_{\{j\}}$ 
as well as the reordering to minimize fill-in and the pivoting
used during the LU factorization remain the same. */

/* Step 1: solve the first linear system (j=0) on the host,
using host sparse direct solver, which involves
full LU factorization and solve. */
/* ... */

/* Step 2: interface to the library by extracting the following
information from the first solve:
a) triangular factors L and U
b) pivoting and reordering permutations P and Q
c) also, allocate all the necessary memory */
/* ... */

/* Step 3: use the library to solve the remaining (j=1, ..., batchSize-1)
linear systems.
a) the library setup (called only once) */
//create handle
status = cusolverRfcreate(&gH);
if (status != CUSOLVER_STATUS_SUCCESS){
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}

```

We call the batch solve method and return.

```

//assemble internal data structures
t1 = cusolver_test_seconds();
status = cusolverRfBatchSetupHost(batchSize, n, nnzA, Ap, Ai, Ax_array,
                                nnzL, Lp, Li, Lx, nnzU, Up, Ui, Ux, P, Q,
gH);
cudaStatus = cudaDeviceSynchronize();
t2 = cusolver_test_seconds();
if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess)) {
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
printf("cusolverRfBatchSetupHost time = %f (s)\n", t2-t1);

//analyze available parallelism
t1 = cusolver_test_seconds();
status = cusolverRfBatchAnalyze(gH);
cudaStatus = cudaDeviceSynchronize();
t2 = cusolver_test_seconds();
if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess)) {
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
printf("cusolverRfBatchAnalyze time = %f (s)\n", t2-t1);

/* b) The library subsequent (j=1,...,batchSize-1) LU re-factorization
and solve (may be called multiple times). For the subsequent batches
the values can be reset using cusolverRfBatch_reset_values_routine. */
//LU re-factorization
t1 = cusolver_test_seconds();
status = cusolverRfBatchRefactor(gH);
cudaStatus = cudaDeviceSynchronize();
t2 = cusolver_test_seconds();
if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess)) {
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
printf("cusolverRfBatchRefactor time = %f (s)\n", t2-t1);

//forward and backward solve
t1 = cusolver_test_seconds();
status = cusolverRfBatchSolve(gH, d_P, d_Q, nrhs, d_T, n, d_X_array, n);
cudaStatus = cudaDeviceSynchronize();
t2 = cusolver_test_seconds();
if ((status != CUSOLVER_STATUS_SUCCESS) || (cudaStatus != cudaSuccess)) {
    printf ("[cusolverRf status %d]\n",status);
    return TEST_FAILED;
}
printf("cusolverRfBatchSolve time = %f (s)\n", t2-t1);

/* free memory and exit */
/* ... */
return TEST_PASSED;
}

```

# Appendix B.

## CSR QR BATCH EXAMPLES

### B.1. Batched Sparse QR example 1

This chapter provides a simple example in the C programming language of how to use batched sparse QR to solve a set of linear systems

$$A_i x_i = b_i$$

All matrices  $A_i$  are small perturbations of

$$A = \begin{pmatrix} 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 2.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 3.0 & 0.0 \\ 0.1 & 0.1 & 0.1 & 4.0 \end{pmatrix}$$

All right-hand side vectors  $b_i$  are small perturbation of the Matlab vector 'ones(4,1)'.  
We assume device memory is big enough to compute all matrices in one pass.

## The usual includes and main definition

```

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>

#include <cusolverSp.h>
#include <cuda_runtime_api.h>

int main(int argc, char*argv[])
{
    cusolverSpHandle_t cusolverH = NULL;
    // GPU does batch QR
    csrqrInfo_t info = NULL;
    cusparseMatDescr_t descrA = NULL;

    cusparseStatus_t cusparse_status = CUSPARSE_STATUS_SUCCESS;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;

    // GPU does batch QR
    // d_A is CSR format, d_csrValA is of size nnzA*batchSize
    // d_x is a matrix of size batchSize * m
    // d_b is a matrix of size batchSize * m
    int *d_csrRowPtrA = NULL;
    int *d_csrColIndA = NULL;
    double *d_csrValA = NULL;
    double *d_b = NULL; // batchSize * m
    double *d_x = NULL; // batchSize * m

    size_t size_qr = 0;
    size_t size_internal = 0;
    void *buffer_qr = NULL; // working space for numerical factorization

    /*
    *   | 1   |
    * A = |     2   |
    *     |     3   |
    *     | 0.1 0.1 0.1 4 |
    *   CSR of A is based-1
    *
    * b = [1 1 1 1]
    */

```

## Set up the library handle and data

```

const int m = 4 ;
const int nnzA = 7;
const int csrRowPtrA[m+1] = { 1, 2, 3, 4, 8};
const int csrColIndA[nnzA] = { 1, 2, 3, 1, 2, 3, 4};
const double csrValA[nnzA] = { 1.0, 2.0, 3.0, 0.1, 0.1, 0.1, 4.0};
const double b[m] = {1.0, 1.0, 1.0, 1.0};
const int batchSize = 17;

double *csrValABatch = (double*)malloc(sizeof(double)*nnzA*batchSize);
double *bBatch       = (double*)malloc(sizeof(double)*m*batchSize);
double *xBatch       = (double*)malloc(sizeof(double)*m*batchSize);
assert( NULL != csrValABatch );
assert( NULL != bBatch );
assert( NULL != xBatch );

// step 1: prepare Aj and bj on host
// Aj is a small perturbation of A
// bj is a small perturbation of b
// csrValABatch = [A0, A1, A2, ...]
// bBatch = [b0, b1, b2, ...]
for(int colidx = 0 ; colidx < nnzA ; colidx++){
    double Areg = csrValA[colidx];
    for (int batchId = 0 ; batchId < batchSize ; batchId++){
        double eps = ((double)((rand() % 100) + 1)) * 1.e-4;
        csrValABatch[batchId*nnzA + colidx] = Areg + eps;
    }
}

for(int j = 0 ; j < m ; j++){
    double breg = b[j];
    for (int batchId = 0 ; batchId < batchSize ; batchId++){
        double eps = ((double)((rand() % 100) + 1)) * 1.e-4;
        bBatch[batchId*m + j] = breg + eps;
    }
}

// step 2: create cusolver handle, qr info and matrix descriptor
cusolver_status = cusolverSpCreate(&cusolverH);
assert( cusolver_status == CUSOLVER_STATUS_SUCCESS);

cusparse_status = cusparseCreateMatDescr(&descrA);
assert(cusparse_status == CUSPARSE_STATUS_SUCCESS);

cusparseSetMatType(descrA, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatIndexBase(descrA, CUSPARSE_INDEX_BASE_ONE); // base-1

cusolver_status = cusolverSpCreateCsrqrInfo(&info);
assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

```

## Call the solver

```

// step 3: copy Aj and bj to device
  cudaStat1 = cudaMalloc ((void**) &d_csrValA , sizeof(double) * nnzA *
batchSize);
  cudaStat2 = cudaMalloc ((void**) &d_csrColIndA, sizeof(int) * nnzA);
  cudaStat3 = cudaMalloc ((void**) &d_csrRowPtrA, sizeof(int) * (m+1));
  cudaStat4 = cudaMalloc ((void**) &d_b , sizeof(double) * m *
batchSize);
  cudaStat5 = cudaMalloc ((void**) &d_x , sizeof(double) * m *
batchSize);
  assert(cudaStat1 == cudaSuccess);
  assert(cudaStat2 == cudaSuccess);
  assert(cudaStat3 == cudaSuccess);
  assert(cudaStat4 == cudaSuccess);
  assert(cudaStat5 == cudaSuccess);

  cudaStat1 = cudaMemcpy(d_csrValA , csrValABatch, sizeof(double) * nnzA *
batchSize, cudaMemcpyHostToDevice);
  cudaStat2 = cudaMemcpy(d_csrColIndA, csrColIndA, sizeof(int) * nnzA,
cudaMemcpyHostToDevice);
  cudaStat3 = cudaMemcpy(d_csrRowPtrA, csrRowPtrA, sizeof(int) * (m+1),
cudaMemcpyHostToDevice);
  cudaStat4 = cudaMemcpy(d_b, bBatch, sizeof(double) * m * batchSize,
cudaMemcpyHostToDevice);
  assert(cudaStat1 == cudaSuccess);
  assert(cudaStat2 == cudaSuccess);
  assert(cudaStat3 == cudaSuccess);
  assert(cudaStat4 == cudaSuccess);

// step 4: symbolic analysis
  cusolver_status = cusolverSpXcsrqrAnalysisBatched(
    cusolverH, m, m, nnzA,
    descrA, d_csrRowPtrA, d_csrColIndA,
    info);
  assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

// step 5: prepare working space
  cusolver_status = cusolverSpDcsrqrBufferInfoBatched(
    cusolverH, m, m, nnzA,
    descrA, d_csrValA, d_csrRowPtrA, d_csrColIndA,
    batchSize,
    info,
    &size_internal,
    &size_qr);
  assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

  printf("numerical factorization needs internal data %lld bytes\n",
(long long) size_internal);
  printf("numerical factorization needs working space %lld bytes\n",
(long long) size_qr);

  cudaStat1 = cudaMalloc((void**) &buffer_qr, size_qr);
  assert(cudaStat1 == cudaSuccess);

```

## Get results back

```

// step 6: numerical factorization
// assume device memory is big enough to compute all matrices.
cusolver_status = cusolverSpDcsrqrsvBatched(
    cusolverH, m, m, nnzA,
    descrA, d_csrValA, d_csrRowPtrA, d_csrColIndA,
    d_b, d_x,
    batchSize,
    info,
    buffer_qr);
assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

// step 7: check residual
// xBatch = [x0, x1, x2, ...]
cudaStat1 = cudaMemcpy(xBatch, d_x, sizeof(double)*m*batchSize,
    cudaMemcpyDeviceToHost);
assert(cudaStat1 == cudaSuccess);

const int baseA = (CUSPARSE_INDEX_BASE_ONE ==
    cusparseGetMatIndexBase(descrA)) ? 1:0 ;

for(int batchId = 0 ; batchId < batchSize; batchId++){
    // measure |bj - Aj*xj|
    double *csrValAj = csrValABatch + batchId * nnzA;
    double *xj = xBatch + batchId * m;
    double *bj = bBatch + batchId * m;
    // sup| bj - Aj*xj|
    double sup_res = 0;
    for(int row = 0 ; row < m ; row++){
        const int start = csrRowPtrA[row] - baseA;
        const int end = csrRowPtrA[row+1] - baseA;
        double Ax = 0.0; // Aj(row,:)*xj
        for(int colIdx = start ; colIdx < end ; colIdx++){
            const int col = csrColIndA[colIdx] - baseA;
            const double Areg = csrValAj[colIdx];
            const double xreg = xj[col];
            Ax = Ax + Areg * xreg;
        }
        double r = bj[row] - Ax;
        sup_res = (sup_res > fabs(r)) ? sup_res : fabs(r);
    }
    printf("batchId %d: sup|bj - Aj*xj| = %E \n", batchId, sup_res);
}

for(int batchId = 0 ; batchId < batchSize; batchId++){
    double *xj = xBatch + batchId * m;
    for(int row = 0 ; row < m ; row++){
        printf("x%d[%d] = %E\n", batchId, row, xj[row]);
    }
    printf("\n");
}

return 0;
}

```

## B.2. Batched Sparse QR example 2

This is the same as example 1 in appendix C except that we assume device memory is not enough, so we need to cut 17 matrices into several chunks and compute each chunk by batched sparse QR.

## The usual includes and main definitions

```

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cusolverSp.h>
#include <cuda_runtime_api.h>

#define imin( x, y ) ((x)<(y)) ? (x) : (y)

int main(int argc, char*argv[])
{
    cusolverSpHandle_t cusolverH = NULL;
    // GPU does batch QR
    csrqrInfo_t info = NULL;
    cusparseMatDescr_t descrA = NULL;

    cusparseStatus_t cusparse_status = CUSPARSE_STATUS_SUCCESS;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;

    // GPU does batch QR
    // d_A is CSR format, d_csrValA is of size nnzA*batchSize
    // d_x is a matrix of size batchSize * m
    // d_b is a matrix of size batchSize * m
    int *d_csrRowPtrA = NULL;
    int *d_csrColIndA = NULL;
    double *d_csrValA = NULL;
    double *d_b = NULL; // batchSize * m
    double *d_x = NULL; // batchSize * m

    size_t size_qr = 0;
    size_t size_internal = 0;
    void *buffer_qr = NULL; // working space for numerical factorization

    /*
     * | 1 |
     * A = | 2 |
     * | 3 |
     * | 0.1 0.1 0.1 4 |
     * CSR of A is based-1
     *
     * b = [1 1 1 1]
     */

```

## Create the library handle

```

const int m = 4 ;
const int nnzA = 7;
const int csrRowPtrA[m+1] = { 1, 2, 3, 4, 8};
const int csrColIndA[nnzA] = { 1, 2, 3, 1, 2, 3, 4};
const double csrValA[nnzA] = { 1.0, 2.0, 3.0, 0.1, 0.1, 0.1, 4.0};
const double b[m] = {1.0, 1.0, 1.0, 1.0};
const int batchSize = 17;

double *csrValABatch = (double*)malloc(sizeof(double)*nnzA*batchSize);
double *bBatch       = (double*)malloc(sizeof(double)*m*batchSize);
double *xBatch       = (double*)malloc(sizeof(double)*m*batchSize);
assert( NULL != csrValABatch );
assert( NULL != bBatch );
assert( NULL != xBatch );

// step 1: prepare Aj and bj on host
// Aj is a small perturbation of A
// bj is a small perturbation of b
// csrValABatch = [A0, A1, A2, ...]
// bBatch = [b0, b1, b2, ...]
for(int colidx = 0 ; colidx < nnzA ; colidx++){
    double Areg = csrValA[colidx];
    for (int batchId = 0 ; batchId < batchSize ; batchId++){
        double eps = ((double)((rand() % 100) + 1)) * 1.e-4;
        csrValABatch[batchId*nnzA + colidx] = Areg + eps;
    }
}

for(int j = 0 ; j < m ; j++){
    double breg = b[j];
    for (int batchId = 0 ; batchId < batchSize ; batchId++){
        double eps = ((double)((rand() % 100) + 1)) * 1.e-4;
        bBatch[batchId*m + j] = breg + eps;
    }
}

// step 2: create cusolver handle, qr info and matrix descriptor
cusolver_status = cusolverSpCreate(&cusolverH);
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);

cusparse_status = cusparseCreateMatDescr(&descrA);
assert(cusparse_status == CUSPARSE_STATUS_SUCCESS);

cusparseSetMatType(descrA, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatIndexBase(descrA, CUSPARSE_INDEX_BASE_ONE); // base-1

cusolver_status = cusolverSpCreateCsrqrInfo(&info);
assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

```

## Set up the data

```

// step 3: copy Aj and bj to device
  cudaStat1 = cudaMalloc ((void**) &d_csrValA , sizeof(double) * nnzA *
batchSize);
  cudaStat2 = cudaMalloc ((void**) &d_csrColIndA, sizeof(int) * nnzA);
  cudaStat3 = cudaMalloc ((void**) &d_csrRowPtrA, sizeof(int) * (m+1));
  cudaStat4 = cudaMalloc ((void**) &d_b , sizeof(double) * m *
batchSize);
  cudaStat5 = cudaMalloc ((void**) &d_x , sizeof(double) * m *
batchSize);
  assert(cudaStat1 == cudaSuccess);
  assert(cudaStat2 == cudaSuccess);
  assert(cudaStat3 == cudaSuccess);
  assert(cudaStat4 == cudaSuccess);
  assert(cudaStat5 == cudaSuccess);

// don't copy csrValABatch and bBatch because device memory may be big enough
  cudaStat1 = cudaMemcpy(d_csrColIndA, csrColIndA, sizeof(int) * nnzA,
cudaMemcpyHostToDevice);
  cudaStat2 = cudaMemcpy(d_csrRowPtrA, csrRowPtrA, sizeof(int) * (m+1),
cudaMemcpyHostToDevice);
  assert(cudaStat1 == cudaSuccess);
  assert(cudaStat2 == cudaSuccess);

// step 4: symbolic analysis
  cusolver_status = cusolverSpXcsrqrAnalysisBatched(
  cusolverH, m, m, nnzA,
  descrA, d_csrRowPtrA, d_csrColIndA,
  info);
  assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

// step 5: find "proper" batchSize
// get available device memory
size_t free_mem = 0;
size_t total_mem = 0;
cudaStat1 = CudaMemGetInfo( &free_mem, &total_mem );
assert( cudaSuccess == cudaStat1 );

int batchSizeMax = 2;
while(batchSizeMax < batchSize){
  printf("batchSizeMax = %d\n", batchSizeMax);
  cusolver_status = cusolverSpDcsrqrBufferInfoBatched(
  cusolverH, m, m, nnzA,
  // d_csrValA is don't care
  descrA, d_csrValA, d_csrRowPtrA, d_csrColIndA,
  batchSizeMax, // WARNING: use batchSizeMax
  info,
  &size_internal,
  &size_qr);
  assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

  if ( (size_internal + size_qr) > free_mem ){
    // current batchSizeMax exceeds hardware limit, so cut it by half.
    batchSizeMax /= 2; break;
  }
  batchSizeMax *= 2; // double batchSizeMax and try it again.
}
// correct batchSizeMax such that it is not greater than batchSize.
batchSizeMax = imin(batchSizeMax, batchSize);
printf("batchSizeMax = %d\n", batchSizeMax);

// Assume device memory is not big enough, and batchSizeMax = 2
batchSizeMax = 2;

```

## Perform analysis and call solve

```

// step 6: prepare working space
// [necessary]
// Need to call cusolverDcsrqrBufferInfoBatched again with batchSizeMax
// to fix batchSize used in numerical factorization.
    cusolver_status = cusolverSpDcsrqrBufferInfoBatched(
        cusolverH, m, m, nnzA,
        // d_csrValA is don't care
        descrA, d_csrValA, d_csrRowPtrA, d_csrColIndA,
        batchSizeMax, // WARNING: use batchSizeMax
        info,
        &size_internal,
        &size_qr);
    assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);

    printf("numerical factorization needs internal data %lld bytes\n",
(long long)size_internal);
    printf("numerical factorization needs working space %lld bytes\n",
(long long)size_qr);

    cudaStat1 = cudaMalloc((void**)&buffer_qr, size_qr);
    assert(cudaStat1 == cudaSuccess);

// step 7: solve  $A_j x_j = b_j$ 
    for(int idx = 0 ; idx < batchSize; idx += batchSizeMax){
        // current batchSize 'cur_batchSize' is the batchSize used in numerical
        factorization
        const int cur_batchSize = imin(batchSizeMax, batchSize - idx);
        printf("current batchSize = %d\n", cur_batchSize);
        // copy part of  $A_j$  and  $b_j$  to device
        cudaStat1 = cudaMemcpy(d_csrValA, csrValABatch + idx*nnzA,
            sizeof(double) * nnzA * cur_batchSize, cudaMemcpyHostToDevice);
        cudaStat2 = cudaMemcpy(d_b, bBatch + idx*m,
            sizeof(double) * m * cur_batchSize, cudaMemcpyHostToDevice);
        assert(cudaStat1 == cudaSuccess);
        assert(cudaStat2 == cudaSuccess);
        // solve part of  $A_j x_j = b_j$ 
        cusolver_status = cusolverSpDcsrqrsvBatched(
            cusolverH, m, m, nnzA,
            descrA, d_csrValA, d_csrRowPtrA, d_csrColIndA,
            d_b, d_x,
            cur_batchSize, // WARNING: use current batchSize
            info,
            buffer_qr);
        assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);
        // copy part of  $x_j$  back to host
        cudaStat1 = cudaMemcpy(xBatch + idx*m, d_x,
            sizeof(double) * m * cur_batchSize, cudaMemcpyDeviceToHost);
        assert(cudaStat1 == cudaSuccess);
    }

```

## Check results

```

// step 7: check residual
// xBatch = [x0, x1, x2, ...]
const int baseA = (CUSPARSE_INDEX_BASE_ONE ==
  cusparseGetMatIndexBase(descrA)) ? 1:0 ;

for(int batchId = 0 ; batchId < batchSize; batchId++){
  // measure |bj - Aj*xj|
  double *csrValAj = csrValABatch + batchId * nnzA;
  double *xj = xBatch + batchId * m;
  double *bj = bBatch + batchId * m;
  // sup| bj - Aj*xj|
  double sup_res = 0;
  for(int row = 0 ; row < m ; row++){
    const int start = csrRowPtrA[row ] - baseA;
    const int end   = csrRowPtrA[row+1] - baseA;
    double Ax = 0.0; // Aj(row,:)*xj
    for(int colIdx = start ; colIdx < end ; colIdx++){
      const int col = csrColIndA[colIdx] - baseA;
      const double Areg = csrValAj[colIdx];
      const double xreg = xj[col];
      Ax = Ax + Areg * xreg;
    }
    double r = bj[row] - Ax;
    sup_res = (sup_res > fabs(r)) ? sup_res : fabs(r);
  }
  printf("batchId %d: sup|bj - Aj*xj| = %E \n", batchId, sup_res);
}

for(int batchId = 0 ; batchId < batchSize; batchId++){
  double *xj = xBatch + batchId * m;
  for(int row = 0 ; row < m ; row++){
    printf("x%d[%d] = %E\n", batchId, row, xj[row]);
  }
  printf("\n");
}

return 0;
}

```

# Appendix C.

## QR EXAMPLES

### C.1. QR Factorization Dense Linear Solver

This chapter provides a simple example in the C programming language of how to use a dense QR factorization to solve a linear system

$$Ax = b$$

A is a 3x3 dense matrix, nonsingular.

$$A = \begin{pmatrix} 1.0 & 2.0 & 3.0 \\ 4.0 & 5.0 & 6.0 \\ 2.0 & 1.0 & 1.0 \end{pmatrix}$$

The following code uses three steps:

Step 1:  $A = Q^*R$  by `geqrf`.

Step 2:  $B := Q^T B$  by `ormqr`.

Step 3: solve  $R^*X = B$  by `trsm`.

## The usual includes and main definition

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include ormqr_example.cpp
 *   nvcc -o -fopenmp a.out ormqr_example.o -L/usr/local/cuda/lib64 -lcudart -
lcublas -lcusolver
 *
 */

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>

#include <cuda_runtime.h>

#include <cublas_v2.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cublasHandle_t cublasH = NULL;
    cublasStatus_t cublas_status = CUBLAS_STATUS_SUCCESS;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    const int m = 3;
    const int lda = m;
    const int ldb = m;
    const int nrhs = 1; // number of right hand side vectors
/*
 *   | 1 2 3 |
 *   A = | 4 5 6 |
 *       | 2 1 1 |
 *
 *   x = (1 1 1)'
 *   b = (6 15 4)'
 */
}

```

## Create the library handle and load the data

```

    double A[lda*m] = { 1.0, 4.0, 2.0, 2.0, 5.0, 1.0, 3.0, 6.0, 1.0};
//    double X[ldb*nrhs] = { 1.0, 1.0, 1.0}; // exact solution
    double B[ldb*nrhs] = { 6.0, 15.0, 4.0};
    double XC[ldb*nrhs]; // solution matrix from GPU
/* device memory */
    double *d_A = NULL;
    double *d_tau = NULL;
    double *d_B = NULL;
    int *devInfo = NULL;
    double *d_work = NULL;
    int lwork_geqrf = 0;
    int lwork_ormqr = 0;
    int lwork = 0;

    int info_gpu = 0;

    const double one = 1;

    printf("A = (matlab base-1)\n");
    printMatrix(m, m, A, lda, "A");
    printf("=====\n");
    printf("B = (matlab base-1)\n");
    printMatrix(m, nrhs, B, ldb, "B");
    printf("=====\n");

/* step 1: create cudense/cublas handle */
    cusolver_status = cusolverDnCreate(&cusolverH);
    assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

    cublas_status = cublasCreate(&cublasH);
    assert(CUBLAS_STATUS_SUCCESS == cublas_status);

/* step 2: copy A and B to device */
    cudaStat1 = cudaMalloc ((void**)&d_A , sizeof(double) * lda * m);
    cudaStat2 = cudaMalloc ((void**)&d_tau, sizeof(double) * m);
    cudaStat3 = cudaMalloc ((void**)&d_B , sizeof(double) * ldb * nrhs);
    cudaStat4 = cudaMalloc ((void**)&devInfo, sizeof(int));
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    assert(cudaSuccess == cudaStat4);

    cudaStat1 = cudaMemcpy(d_A, A, sizeof(double) * lda * m ,
        cudaMemcpyHostToDevice);
    cudaStat2 = cudaMemcpy(d_B, B, sizeof(double) * ldb * nrhs,
        cudaMemcpyHostToDevice);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);

```

## Call the solver

```

/* step 3: query working space of geqrf and ormqr */
cusolver_status = cusolverDnDgeqrf_bufferSize(
    cudenseH,
    m,
    m,
    d_A,
    lda,
    &lwork_geqrf);
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);

cusolver_status= cusolverDnDormqr_bufferSize(
    cudenseH,
    CUBLAS_SIDE_LEFT,
    CUBLAS_OP_T,
    m,
    nrhs,
    m,
    d_A,
    lda,
    d_tau,
    d_B,
    ldb,
    &lwork_ormqr);
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);

lwork = (lwork_geqrf > lwork_ormqr)? lwork_geqrf : lwork_ormqr;

cudaStat1 = cudaMalloc((void**) &d_work, sizeof(double)*lwork);
assert(cudaSuccess == cudaStat1);

/* step 4: compute QR factorization */
cusolver_status = cusolverDnDgeqrf(
    cusolverH,
    m,
    m,
    d_A,
    lda,
    d_tau,
    d_work,
    lwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

/* check if QR is good or not */
cudaStat1 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
    cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("after geqrf: info_gpu = %d\n", info_gpu);
assert(0 == info_gpu);

```

## Check the results

```

/* step 5: compute Q^T*B */
cusolver_status= cusolverDnDormqr(
    cudenseH,
    CUBLAS_SIDE_LEFT,
    CUBLAS_OP_T,
    m,
    nrhs,
    m,
    d_A,
    lda,
    d_tau,
    d_B,
    ldb,
    d_work,
    lwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);
/* check if QR is good or not */
cudaStat1 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
    cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("after ormqr: info_gpu = %d\n", info_gpu);
assert(0 == info_gpu);

/* step 6: compute x = R \ Q^T*B */
cublas_status = cublasDtrsm(
    cublasH,
    CUBLAS_SIDE_LEFT,
    CUBLAS_FILL_MODE_UPPER,
    CUBLAS_OP_N,
    CUBLAS_DIAG_NON_UNIT,
    m,
    nrhs,
    &one,
    d_A,
    lda,
    d_B,
    ldb);
cudaStat1 = cudaDeviceSynchronize();
assert(CUBLAS_STATUS_SUCCESS == cublas_status);
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(XC, d_B, sizeof(double)*ldb*nrhs,
    cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("X = (matlab base-1)\n");
printMatrix(m, nrhs, XC, ldb, "X");

/* free resources */
if (d_A ) cudaFree(d_A);
if (d_tau ) cudaFree(d_tau);
if (d_B ) cudaFree(d_B);
if (devInfo) cudaFree(devInfo);
if (d_work ) cudaFree(d_work);
if (cublasH ) cublasDestroy(cublasH);
if (cusolverH) cusolverDnDestroy(cusolverH);
cudaDeviceReset();
return 0;
}

```

## C.2. orthogonalization

This chapter provides a simple example in the C programming language of how to do orthogonalization by QR factorization.

$$A = Q * R$$

A is a 3x2 dense matrix,

$$A = \begin{pmatrix} 1.0 & 2.0 \\ 4.0 & 5.0 \\ 2.0 & 1.0 \end{pmatrix}$$

The following code uses three steps:

Step 1:  $A = Q * R$  by `geqrf`.

Step 2: form Q by `orgqr`.

Step 3: check if Q is unitary or not.

## The usual includes and main definition

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 * nvcc -c -I/usr/local/cuda/include orgqr_example.cpp
 * g++ -fopenmp -o a.out orgqr_example.o -I/usr/local/cuda/lib64 -lcudart -
lcublas -lcusolver
 *
 */

#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>

#include <cuda_runtime.h>

#include <cublas_v2.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cublasHandle_t cublasH = NULL;
    cublasStatus_t cublas_status = CUBLAS_STATUS_SUCCESS;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    const int m = 3;
    const int n = 2;
    const int lda = m;
/*
 *   | 1 2 |
 *   A = | 4 5 |
 *       | 2 1 |
 */

```

## Create the library handle and load the data

```

double A[lda*n] = { 1.0, 4.0, 2.0, 2.0, 5.0, 1.0};
double Q[lda*n]; // orthonormal columns
double R[n*n]; // R = I - Q**T*Q

double *d_A = NULL;
double *d_tau = NULL;
int *devInfo = NULL;
double *d_work = NULL;

double *d_R = NULL;

int lwork_geqrf = 0;
int lwork_orgqr = 0;
int lwork = 0;

int info_gpu = 0;

const double h_one = 1;
const double h_minus_one = -1;

printf("A = (matlab base-1)\n");
printMatrix(m, n, A, lda, "A");
printf("=====\n");

// step 1: create cusolverDn/cublas handle
cusolver_status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

cublas_status = cublasCreate(&cublasH);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

// step 2: copy A and B to device
cudaStat1 = cudaMalloc ((void**)&d_A , sizeof(double)*lda*n);
cudaStat2 = cudaMalloc ((void**)&d_tau, sizeof(double)*n);
cudaStat3 = cudaMalloc ((void**)&devInfo, sizeof(int));
cudaStat4 = cudaMalloc ((void**)&d_R , sizeof(double)*n*n);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*n,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

```

## Call the solver

```

// step 3: query working space of geqrf and orgqr
cusolver_status = cusolverDnDgeqrf_bufferSize(
    cusolverH,
    m,
    n,
    d_A,
    lda,
    &lwork_geqrf);
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);
cusolver_status = cusolverDnDorgqr_bufferSize(
    cusolverH,
    m,
    n,
    n,
    d_A,
    lda,
    &lwork_orgqr);
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);
// lwork = max(lwork_geqrf, lwork_orgqr)
lwork = (lwork_geqrf > lwork_orgqr)? lwork_geqrf : lwork_orgqr;

cudaStat1 = cudaMalloc((void**) &d_work, sizeof(double)*lwork);
assert(cudaSuccess == cudaStat1);

// step 4: compute QR factorization
cusolver_status = cusolverDnDgeqrf(
    cusolverH,
    m,
    n,
    d_A,
    lda,
    d_tau,
    d_work,
    lwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

// check if QR is successful or not
cudaStat1 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
    cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("after geqrf: info_gpu = %d\n", info_gpu);
assert(0 == info_gpu);

// step 5: compute Q
cusolver_status = cusolverDnDorgqr(
    cusolverH,
    m,
    n,
    n,
    d_A,
    lda,
    d_tau,
    d_work,
    lwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

```

## Check the results

```

// check if QR is good or not
cudaStat1 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("after orgqr: info_gpu = %d\n", info_gpu);
assert(0 == info_gpu);

cudaStat1 = cudaMemcpy(Q, d_A, sizeof(double)*lda*n,
cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("Q = (matlab base-1)\n");
printMatrix(m, n, Q, lda, "Q");

// step 6: measure R = I - Q**T*Q
memset(R, 0, sizeof(double)*n*n);
for(int j = 0 ; j < n ; j++){
    R[j + n*j] = 1.0; // R(j,j)=1
}

cudaStat1 = cudaMemcpy(d_R, R, sizeof(double)*n*n, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

// R = -Q**T*Q + I
cublas_status = cublasDgemm_v2(
    cublasH,
    CUBLAS_OP_T, // Q**T
    CUBLAS_OP_N, // Q
    n, // number of rows of R
    n, // number of columns of R
    m, // number of columns of Q**T
    &h_minus_one, /* host pointer */
    d_A, // Q**T
    lda,
    d_A, // Q
    lda,
    &h_one, /* hostpointer */
    d_R,
    n);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

double dR_nrm2 = 0.0;
cublas_status = cublasDnrm2_v2(
    cublasH, n*n, d_R, 1, &dR_nrm2);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

printf("||I - Q**T*Q|| = %E\n", dR_nrm2);

```

## free resources

```
// free resources
if (d_A ) cudaFree(d_A);
if (d_tau ) cudaFree(d_tau);
if (dēvInfo) cudaFree(dēvInfo);
if (d_work ) cudaFree(d_work);
if (d_R ) cudaFree(d_R);

if (cublasH ) cublasDestroy(cublasH);
if (cusolverH) cusolverDnDestroy(cusolverH);

cudaDeviceReset();

return 0;
}
```

# Appendix D.

## LU EXAMPLES

### D.1. LU Factorization

This chapter provides a simple example in the C programming language of how to use a dense LU factorization to solve a linear system

$$Ax = b$$

A is a 3x3 dense matrix, nonsingular.

$$A = \begin{pmatrix} 1.0 & 2.0 & 3.0 \\ 4.0 & 5.0 & 6.0 \\ 7.0 & 8.0 & 10.0 \end{pmatrix}$$

The code uses `getrf` to do LU factorization and `getrs` to do backward and forward solve. The parameter `pivot_on` decides whether partial pivoting is performed or not.

...

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include getrf_example.cpp
 *   g++ -fopenmp -o a.out getrf_example.o -L/usr/local/cuda/lib64 -lcusolver -lcudart
 */

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cudaStream_t stream = NULL;

    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    const int m = 3;
    const int lda = m;
    const int ldb = m;
/*
 *   | 1 2 3 |
 *   A = | 4 5 6 |
 *       | 7 8 10 |
 *
 * without pivoting: A = L*U
 *   | 1 0 0 |   | 1 2 3 |
 *   L = | 4 1 0 |, U = | 0 -3 -6 |
 *       | 7 2 1 |   | 0 0 1 |
 *
 * with pivoting: P*A = L*U
 *   | 0 0 1 |
 *   P = | 1 0 0 |
 *       | 0 1 0 |
 *
 *   | 1   0   0 |   | 7 8   10   |
 *   L = | 0.1429 1   0 |, U = | 0 0.8571 1.5714 |
 *       | 0.5714 0.5 1 |   | 0 0   -0.5   |
 */

```

...

```

double A[lda*m] = { 1.0, 4.0, 7.0, 2.0, 5.0, 8.0, 3.0, 6.0, 10.0};
double B[m] = { 1.0, 2.0, 3.0 };
double X[m]; /* X = A\B */
double LU[lda*m]; /* L and U */
int Ipiv[m]; /* host copy of pivoting sequence */
int info = 0; /* host copy of error info */

double *d_A = NULL; /* device copy of A */
double *d_B = NULL; /* device copy of B */
int *d_Ipiv = NULL; /* pivoting sequence */
int *d_info = NULL; /* error info */
int lwork = 0; /* size of workspace */
double *d_work = NULL; /* device workspace for getrf */

const int pivot_on = 0;

printf("example of getrf \n");

if (pivot_on){
    printf("pivot is on : compute P*A = L*U \n");
}else{
    printf("pivot is off: compute A = L*U (not numerically stable)\n");
}

printf("A = (matlab base-1)\n");
printMatrix(m, m, A, lda, "A");
printf("=====\n");

printf("B = (matlab base-1)\n");
printMatrix(m, 1, B, ldb, "B");
printf("=====\n");

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusolverDnSetStream(cusolverH, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: copy A to device */
cudaStat1 = cudaMalloc ((void**)&d_A, sizeof(double) * lda * m);
cudaStat2 = cudaMalloc ((void**)&d_B, sizeof(double) * m);
cudaStat3 = cudaMalloc ((void**)&d_Ipiv, sizeof(int) * m);
cudaStat4 = cudaMalloc ((void**)&d_info, sizeof(int));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*m,
cudaMemcpyHostToDevice);
cudaStat2 = cudaMemcpy(d_B, B, sizeof(double)*m, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

```

...

```

/* step 3: query working space of getrf */
    status = cusolverDnDgetrf_bufferSize(
        cusolverH,
        m,
        m,
        d_A,
        lda,
        &lwork);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    cudaStat1 = cudaMalloc((void**) &d_work, sizeof(double)*lwork);
    assert(cudaSuccess == cudaStat1);

/* step 4: LU factorization */
    if (pivot_on){
        status = cusolverDnDgetrf(
            cusolverH,
            m,
            m,
            d_A,
            lda,
            d_work,
            d_Ipiv,
            d_info);
    }else{
        status = cusolverDnDgetrf(
            cusolverH,
            m,
            m,
            d_A,
            lda,
            d_work,
            NULL,
            d_info);
    }
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    if (pivot_on){
        cudaStat1 = cudaMemcpy(Ipiv , d_Ipiv, sizeof(int)*m,
            cudaMemcpyDeviceToHost);
    }
    cudaStat2 = cudaMemcpy(LU , d_A , sizeof(double)*lda*m,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(&info, d_info, sizeof(int), cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);

    if ( 0 > info ){
        printf("%d-th parameter is wrong \n", -info);
        exit(1);
    }
    if (pivot_on){
        printf("pivoting sequence, matlab base-1\n");
        for(int j = 0 ; j < m ; j++){
            printf("Ipiv(%d) = %d\n", j+1, Ipiv[j]);
        }
    }
    printf("L and U = (matlab base-1)\n");
    printMatrix(m, m, LU, lda, "LU");
    printf("=====\n");

```

...

```

/*
 * step 5: solve A*X = B
 *   B = | 1 |, X = | -0.3333 |
 *       | 2 |,   | 0.6667 |
 *       | 3 |,   | 0      |
 *
 */
if (pivot_on){
    status = cusolverDnDgetrs(
        cusolverH,
        CUBLAS_OP_N,
        m,
        1, /* nrhs */
        d_A,
        lda,
        d_Ipiv,
        d_B,
        ldb,
        d_info);
} else{
    status = cusolverDnDgetrs(
        cusolverH,
        CUBLAS_OP_N,
        m,
        1, /* nrhs */
        d_A,
        lda,
        NULL,
        d_B,
        ldb,
        d_info);
}
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == status);
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(X, d_B, sizeof(double)*m, cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("X = (matlab base-1)\n");
printMatrix(m, 1, X, ldb, "X");
printf("=====\n");

/* free resources */
if (d_A) cudaFree(d_A);
if (d_B) cudaFree(d_B);
if (d_Ipiv) cudaFree(d_Ipiv);
if (d_info) cudaFree(d_info);
if (d_work) cudaFree(d_work);

if (cusolverH) cusolverDnDestroy(cusolverH);
if (stream) cudaStreamDestroy(stream);

cudaDeviceReset();

return 0;
}

```

# Appendix E.

## CHOLESKY EXAMPLES

### E.1. batched Cholesky Factorization

This chapter provides a simple example in the C programming language of how to use a batched dense Cholesky factorization to solve a sequence of linear systems

$$A[i]*x[i]=b[i]$$

each **A[i]** is a 3x3 dense Hermitian matrix. In this example, there are two matrices, **A0** and **A1**. **A0** is positive definite and **A1** is not.

The code uses **potrfBatched** to do Cholesky factorization and **potrsBatched** to do backward and forward solve. **potrfBatched** would report singularity on **A1**.

```

...
/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 * nvcc -c -I/usr/local/cuda/include batchchol_example.cpp
 * g++ -o a.out batchchol_example.o -L/usr/local/cuda/lib64 -lcusolver -
lcudart
 */

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t handle = NULL;
    cudaStream_t stream = NULL;

    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;

    const cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;
    const int batchSize = 2;
    const int nrhs = 1;
    const int m = 3;
    const int lda = m;
    const int ldb = m;

    /*
     *      | 1      2      3 |
     * A0 = | 2      5      5 | = L0 * L0**T
     *      | 3      5     12 |
     *
     *      | 1.0000      0      0 |
     * where L0 = | 2.0000      1.0000      0 |
     *             | 3.0000     -1.0000      1.4142 |
     *
     *      | 1      2      3 |
     * A1 = | 2      4      5 | is not s.p.d., failed at row 2
     *      | 3      5     12 |
     *
     */
}

```

...

```

double A0[lda*m] = { 1.0, 2.0, 3.0, 2.0, 5.0, 5.0, 3.0, 5.0, 12.0 };
double A1[lda*m] = { 1.0, 2.0, 3.0, 2.0, 4.0, 5.0, 3.0, 5.0, 12.0 };
double B0[m] = { 1.0, 1.0, 1.0 };
double X0[m]; /* X0 = A0\B0 */
int infoArray[batchSize]; /* host copy of error info */

double L0[lda*m]; /* cholesky factor of A0 */

double *Aarray[batchSize];
double *Barray[batchSize];

double **d_Aarray = NULL;
double **d_Barray = NULL;
int *d_infoArray = NULL;

printf("example of batched Cholesky \n");

printf("A0 = (matlab base-1)\n");
printMatrix(m, m, A0, lda, "A0");
printf("=====\n");

printf("A1 = (matlab base-1)\n");
printMatrix(m, m, A1, lda, "A1");
printf("=====\n");

printf("B0 = (matlab base-1)\n");
printMatrix(m, 1, B0, ldb, "B0");
printf("=====\n");

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&handle);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusolverDnSetStream(handle, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: copy A to device */
for(int j = 0 ; j < batchSize ; j++){
    cudaStat1 = cudaMalloc ((void*)&Aarray[j], sizeof(double) * lda * m);
    assert(cudaSuccess == cudaStat1);
    cudaStat2 = cudaMalloc ((void*)&Barray[j], sizeof(double) * ldb *
nrhs);
    assert(cudaSuccess == cudaStat2);
}
cudaStat1 = cudaMalloc ((void*)&d_infoArray, sizeof(int)*batchSize);
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMalloc ((void*)&d_Aarray, sizeof(double*) * batchSize);
cudaStat2 = cudaMalloc ((void*)&d_Barray, sizeof(double*) * batchSize);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

cudaStat1 = cudaMemcpy(Aarray[0], A0, sizeof(double) * lda * m,
cudaMemcpyHostToDevice);
cudaStat2 = cudaMemcpy(Aarray[1], A1, sizeof(double) * lda * m,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

```

...

```

    cudaStat1 = cudaMemcpy(Barray[0], B0, sizeof(double) * m,
cudaMemcpyHostToDevice);
    cudaStat2 = cudaMemcpy(Barray[1], B0, sizeof(double) * m,
cudaMemcpyHostToDevice);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);

    cudaStat1 = cudaMemcpy(d_Aarray, Aarray, sizeof(double*)*batchSize,
cudaMemcpyHostToDevice);
    cudaStat2 = cudaMemcpy(d_Barray, Barray, sizeof(double*)*batchSize,
cudaMemcpyHostToDevice);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    cudaDeviceSynchronize();

/* step 3: Cholesky factorization */
    status = cusolverDnDpotrfBatched(
        handle,
        uplo,
        m,
        d_Aarray,
        lda,
        d_infoArray,
        batchSize);
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(infoArray, d_infoArray, sizeof(int)*batchSize,
cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(L0, Aarray[0], sizeof(double) * lda * m,
cudaMemcpyDeviceToHost);
    cudaDeviceSynchronize();
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);

    for(int j = 0 ; j < batchSize ; j++){
        printf("info[%d] = %d\n", j, infoArray[j]);
    }

    assert( 0 == infoArray[0] );
/* A1 is singular */
    assert( 2 == infoArray[1] );

    printf("L = (matlab base-1), upper triangle is don't care \n");
    printMatrix(m, m, L0, lda, "L0");
    printf("=====\n");

/*
 * step 4: solve A0*X0 = B0
 *      | 1 |      | 10.5 |
 *   B0 = | 1 |,   X0 = | -2.5 |
 *      | 1 |      | -1.5 |
 */
    status = cusolverDnDpotrsBatched(
        handle,
        uplo,
        m,
        nrhs, /* only support rhs = 1*/
        d_Aarray,
        lda,
        d_Barray,
        ldb,
        d_infoArray,
        batchSize);

```

...

```

    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(infoArray, d_infoArray, sizeof(int),
        cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(X0 , Barray[0], sizeof(double)*m,
        cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    cudaDeviceSynchronize();

    printf("info = %d\n", infoArray[0]);
    assert( 0 == infoArray[0] );

    printf("X0 = (matlab base-1)\n");
    printMatrix(m, 1, X0, ldb, "X0");
    printf("=====\n");

/* free resources */
    if (d_Aarray ) cudaFree(d_Aarray);
    if (d_Barray ) cudaFree(d_Barray);
    if (d_infoArray ) cudaFree(d_infoArray);

    if (handle ) cusolverDnDestroy(handle);
    if (stream ) cudaStreamDestroy(stream);

    cudaDeviceReset();

    return 0;
}

```

# Appendix F.

## EXAMPLES OF DENSE EIGENVALUE SOLVER

### F.1. Standard Symmetric Dense Eigenvalue Solver

This chapter provides a simple example in the C programming language of how to use `syevd` to compute the spectrum of a dense symmetric system by

$$Ax = \lambda x$$

where  $A$  is a 3x3 dense symmetric matrix

$$A = \begin{pmatrix} 3.5 & 0.5 & 0 \\ 0.5 & 3.5 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

The following code uses `syevd` to compute eigenvalues and eigenvectors, then compare to exact eigenvalues {2,3,4}.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include syevd_example.cpp
 *   g++ -o a.out syevd_example.o -L/usr/local/cuda/lib64 -lcudart -lcusolver
 */

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    const int m = 3;
    const int lda = m;
    /*
     *   | 3.5 0.5 0 |
     *   A = | 0.5 3.5 0 |
     *       | 0   0  2 |
     */
    double A[lda*m] = { 3.5, 0.5, 0, 0.5, 3.5, 0, 0, 0, 2.0};
    double lambda[m] = { 2.0, 3.0, 4.0};

    double V[lda*m]; // eigenvectors
    double W[m]; // eigenvalues

    double *d_A = NULL;
    double *d_W = NULL;
    int *devInfo = NULL;
    double *d_work = NULL;
    int lwork = 0;

    int info_gpu = 0;

    printf("A = (matlab base-1)\n");
    printMatrix(m, m, A, lda, "A");
    printf("=====\n");

```

## call eigenvalue solver

```

// step 1: create cusolver/cublas handle
cusolver_status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

// step 2: copy A and B to device
cudaStat1 = cudaMalloc ((void**)&d_A, sizeof(double) * lda * m);
cudaStat2 = cudaMalloc ((void**)&d_W, sizeof(double) * m);
cudaStat3 = cudaMalloc ((void**)&devInfo, sizeof(int));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);

    cudaStat1 = cudaMemcpy(d_A, A, sizeof(double) * lda * m,
        cudaMemcpyHostToDevice);
    assert(cudaSuccess == cudaStat1);

// step 3: query working space of syevd
cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute eigenvalues
and eigenvectors.
cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;
cusolver_status = cusolverDnDsyevd_bufferSize(
    cusolverH,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_W,
    &lwork);
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);

    cudaStat1 = cudaMalloc((void**)&d_work, sizeof(double)*lwork);
    assert(cudaSuccess == cudaStat1);

// step 4: compute spectrum
cusolver_status = cusolverDnDsyevd(
    cusolverH,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_W,
    d_work,
    lwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(W, d_W, sizeof(double)*m, cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_A, sizeof(double)*lda*m,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
        cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);

```

check the result

```

printf("after syevd: info_gpu = %d\n", info_gpu);
assert(0 == info_gpu);

printf("eigenvalue = (matlab base-1), ascending order\n");
for(int i = 0 ; i < m ; i++){
    printf("W[%d] = %E\n", i+1, W[i]);
}

printf("V = (matlab base-1)\n");
printMatrix(m, m, V, lda, "V");
printf("=====\n");

// step 4: check eigenvalues
double lambda_sup = 0;
for(int i = 0 ; i < m ; i++){
    double error = fabs( lambda[i] - W[i]);
    lambda_sup = (lambda_sup > error)? lambda_sup : error;
}
printf("|lambda - W| = %E\n", lambda_sup);

// free resources
if (d_A ) cudaFree(d_A);
if (d_W ) cudaFree(d_W);
if (devInfo) cudaFree(devInfo);
if (d_work ) cudaFree(d_work);

if (cusolverH) cusolverDnDestroy(cusolverH);

cudaDeviceReset();

return 0;
}

```

## F.2. Standard Symmetric Dense Eigenvalue Solver

This chapter provides a simple example in the C programming language of how to use syevd to compute the spectrum of a dense symmetric system by

$$Ax = \lambda x$$

where A is a 3x3 dense symmetric matrix

$$A = \begin{pmatrix} 3.5 & 0.5 & 0 \\ 0.5 & 3.5 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

The following code uses **syevd** to compute eigenvalues and eigenvectors, then compare to exact eigenvalues {2,3,4}.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include syevd_example.cpp
 *   g++ -o a.out syevd_example.o -L/usr/local/cuda/lib64 -lcudart -lcusolver
 */

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    const int m = 3;
    const int lda = m;
    /*
     *   | 3.5 0.5 0 |
     *   A = | 0.5 3.5 0 |
     *       | 0   0  2 |
     */
    double A[lda*m] = { 3.5, 0.5, 0, 0.5, 3.5, 0, 0, 0, 2.0};
    double lambda[m] = { 2.0, 3.0, 4.0};

    double V[lda*m]; // eigenvectors
    double W[m]; // eigenvalues

    double *d_A = NULL;
    double *d_W = NULL;
    int *devInfo = NULL;
    double *d_work = NULL;
    int lwork = 0;

    int info_gpu = 0;

    printf("A = (matlab base-1)\n");
    printMatrix(m, m, A, lda, "A");
    printf("=====\n");

```

## call eigenvalue solver

```

// step 1: create cusolver/cublas handle
cusolver_status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

// step 2: copy A and B to device
cudaStat1 = cudaMalloc ((void**)&d_A, sizeof(double) * lda * m);
cudaStat2 = cudaMalloc ((void**)&d_W, sizeof(double) * m);
cudaStat3 = cudaMalloc ((void**)&devInfo, sizeof(int));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);

    cudaStat1 = cudaMemcpy(d_A, A, sizeof(double) * lda * m,
        cudaMemcpyHostToDevice);
    assert(cudaSuccess == cudaStat1);

// step 3: query working space of syevd
cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute eigenvalues
and eigenvectors.
cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;
cusolver_status = cusolverDnDsyevd_bufferSize(
    cusolverH,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_W,
    &lwork);
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);

    cudaStat1 = cudaMalloc((void**)&d_work, sizeof(double)*lwork);
    assert(cudaSuccess == cudaStat1);

// step 4: compute spectrum
cusolver_status = cusolverDnDsyevd(
    cusolverH,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_W,
    d_work,
    lwork,
    devInfo);
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(W, d_W, sizeof(double)*m, cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_A, sizeof(double)*lda*m,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
        cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);

```

check the result

```

printf("after syevd: info_gpu = %d\n", info_gpu);
assert(0 == info_gpu);

printf("eigenvalue = (matlab base-1), ascending order\n");
for(int i = 0 ; i < m ; i++){
    printf("W[%d] = %E\n", i+1, W[i]);
}

printf("V = (matlab base-1)\n");
printMatrix(m, m, V, lda, "V");
printf("=====\n");

// step 4: check eigenvalues
double lambda_sup = 0;
for(int i = 0 ; i < m ; i++){
    double error = fabs( lambda[i] - W[i]);
    lambda_sup = (lambda_sup > error)? lambda_sup : error;
}
printf("|lambda - W| = %E\n", lambda_sup);

// free resources
if (d_A ) cudaFree(d_A);
if (d_W ) cudaFree(d_W);
if (devInfo) cudaFree(devInfo);
if (d_work ) cudaFree(d_work);

if (cusolverH) cusolverDnDestroy(cusolverH);

cudaDeviceReset();

return 0;
}

```

## F.3. Generalized Symmetric-Definite Dense Eigenvalue Solver

This chapter provides a simple example in the C programming language of how to use `sygvd` to compute spectrum of a pair of dense symmetric matrices (A,B) by

$$Ax = \lambda Bx$$

where A is a 3x3 dense symmetric matrix

$$A = \begin{pmatrix} 3.5 & 0.5 & 0 \\ 0.5 & 3.5 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

and B is a 3x3 positive definite matrix

$$B = \begin{pmatrix} 10 & 2 & 3 \\ 2 & 10 & 5 \\ 3 & 5 & 10 \end{pmatrix}$$

The following code uses **sygvd** to compute eigenvalues and eigenvectors, then compare to exact eigenvalues {0.158660256604, 0.370751508101882, 0.6}.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include sygvd_example.cpp
 *   g++ -o a.out sygvd_example.o -L/usr/local/cuda/lib64 -lcusolver
 */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    const int m = 3;
    const int lda = m;

    /*
     *   | 3.5 0.5 0 |
     *   A = | 0.5 3.5 0 |
     *       | 0   0  2 |
     *
     *   | 10  2  3 |
     *   B = | 2  10  5 |
     *       | 3  5  10 |
     */
    double A[lda*m] = { 3.5, 0.5, 0, 0.5, 3.5, 0, 0, 0, 2.0};
    double B[lda*m] = { 10.0, 2.0, 3.0, 2.0, 10.0, 5.0, 3.0, 5.0, 10.0};
    double lambda[m] = { 0.158660256604, 0.370751508101882, 0.6};

    double V[lda*m]; // eigenvectors
    double W[m]; // eigenvalues

    double *d_A = NULL;
    double *d_B = NULL;
    double *d_W = NULL;
    int *devInfo = NULL;
    double *d_work = NULL;
    int lwork = 0;
    int info_gpu = 0;

    printf("A = (matlab base-1)\n");
    printMatrix(m, m, A, lda, "A");
    printf("=====\n");

    printf("B = (matlab base-1)\n");
    printMatrix(m, m, B, lda, "B");
    printf("=====\n");
}

```

## call eigenvalue solver

```

// step 1: create cusolver/cublas handle
cusolver_status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

// step 2: copy A and B to device
cudaStat1 = cudaMalloc ((void**)&d_A, sizeof(double) * lda * m);
cudaStat2 = cudaMalloc ((void**)&d_B, sizeof(double) * lda * m);
cudaStat3 = cudaMalloc ((void**)&d_W, sizeof(double) * m);
cudaStat4 = cudaMalloc ((void**)&devInfo, sizeof(int));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double) * lda * m,
cudaMemcpyHostToDevice);
cudaStat2 = cudaMemcpy(d_B, B, sizeof(double) * lda * m,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

// step 3: query working space of sygvd
cusolverEigType_t itype = CUSOLVER_EIG_TYPE_1; // A*x = (lambda)*B*x
cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute eigenvalues
and eigenvectors.
cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;
cusolver_status = cusolverDnDsygvd_bufferSize(
    cusolverH,
    itype,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_B,
    lda,
    d_W,
    &lwork);
assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);
cudaStat1 = cudaMalloc((void**)&d_work, sizeof(double)*lwork);
assert(cudaSuccess == cudaStat1);

// step 4: compute spectrum of (A,B)
cusolver_status = cusolverDnDsygvd(
    cusolverH,
    itype,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_B,
    lda,
    d_W,
    d_work,
    lwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

```

check the result

```

    cudaStat1 = cudaMemcpy(W, d_W, sizeof(double)*m, cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_A, sizeof(double)*lda*m,
cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);

    printf("after sygvd: info_gpu = %d\n", info_gpu);
    assert(0 == info_gpu);

    printf("eigenvalue = (matlab base-1), ascending order\n");
    for(int i = 0 ; i < m ; i++){
        printf("W[%d] = %E\n", i+1, W[i]);
    }

    printf("V = (matlab base-1)\n");
    printMatrix(m, m, V, lda, "V");
    printf("=====\n");

// step 4: check eigenvalues
double lambda_sup = 0;
for(int i = 0 ; i < m ; i++){
    double error = fabs( lambda[i] - W[i]);
    lambda_sup = (lambda_sup > error)? lambda_sup : error;
}
printf("|lambda - W| = %E\n", lambda_sup);

// free resources
if (d_A ) cudaFree(d_A);
if (d_B ) cudaFree(d_B);
if (d_W ) cudaFree(d_W);
if (devInfo) cudaFree(devInfo);
if (d_work ) cudaFree(d_work);

if (cusolverH) cusolverDnDestroy(cusolverH);

cudaDeviceReset();

return 0;
}

```

## F.4. Generalized Symmetric-Definite Dense Eigenvalue Solver

This chapter provides a simple example in the C programming language of how to use sygvd to compute spectrum of a pair of dense symmetric matrices (A,B) by

$$Ax = \lambda Bx$$

where A is a 3x3 dense symmetric matrix

$$A = \begin{pmatrix} 3.5 & 0.5 & 0 \\ 0.5 & 3.5 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

and  $B$  is a 3x3 positive definite matrix

$$B = \begin{pmatrix} 10 & 2 & 3 \\ 2 & 10 & 5 \\ 3 & 5 & 10 \end{pmatrix}$$

The following code uses **sygvd** to compute eigenvalues and eigenvectors, then compare to exact eigenvalues {0.158660256604, 0.370751508101882, 0.6}.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include sygvd_example.cpp
 *   g++ -o a.out sygvd_example.o -L/usr/local/cuda/lib64 -lcusolver
 */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    const int m = 3;
    const int lda = m;

    /*
     *   | 3.5 0.5 0 |
     *   A = | 0.5 3.5 0 |
     *       | 0   0  2 |
     *
     *   | 10  2  3 |
     *   B = | 2  10  5 |
     *       | 3  5  10 |
     */
    double A[lda*m] = { 3.5, 0.5, 0, 0.5, 3.5, 0, 0, 0, 2.0};
    double B[lda*m] = { 10.0, 2.0, 3.0, 2.0, 10.0, 5.0, 3.0, 5.0, 10.0};
    double lambda[m] = { 0.158660256604, 0.370751508101882, 0.6};

    double V[lda*m]; // eigenvectors
    double W[m]; // eigenvalues

    double *d_A = NULL;
    double *d_B = NULL;
    double *d_W = NULL;
    int *devInfo = NULL;
    double *d_work = NULL;
    int lwork = 0;
    int info_gpu = 0;

    printf("A = (matlab base-1)\n");
    printMatrix(m, m, A, lda, "A");
    printf("=====\n");

    printf("B = (matlab base-1)\n");
    printMatrix(m, m, B, lda, "B");
    printf("=====\n");
}

```

## call eigenvalue solver

```

// step 1: create cusolver/cublas handle
cusolver_status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

// step 2: copy A and B to device
cudaStat1 = cudaMalloc ((void**)&d_A, sizeof(double) * lda * m);
cudaStat2 = cudaMalloc ((void**)&d_B, sizeof(double) * lda * m);
cudaStat3 = cudaMalloc ((void**)&d_W, sizeof(double) * m);
cudaStat4 = cudaMalloc ((void**)&devInfo, sizeof(int));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double) * lda * m,
cudaMemcpyHostToDevice);
cudaStat2 = cudaMemcpy(d_B, B, sizeof(double) * lda * m,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

// step 3: query working space of sygvd
cusolverEigType_t itype = CUSOLVER_EIG_TYPE_1; // A*x = (lambda)*B*x
cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute eigenvalues
and eigenvectors.
cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;
cusolver_status = cusolverDnDsygvd_bufferSize(
    cusolverH,
    itype,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_B,
    lda,
    d_W,
    &lwork);
assert(cusolver_status == CUSOLVER_STATUS_SUCCESS);
cudaStat1 = cudaMalloc((void**)&d_work, sizeof(double)*lwork);
assert(cudaSuccess == cudaStat1);

// step 4: compute spectrum of (A,B)
cusolver_status = cusolverDnDsygvd(
    cusolverH,
    itype,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_B,
    lda,
    d_W,
    d_work,
    lwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

```

check the result

```

    cudaStat1 = cudaMemcpy(W, d_W, sizeof(double)*m, cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_A, sizeof(double)*lda*m,
cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);

    printf("after sygvd: info_gpu = %d\n", info_gpu);
    assert(0 == info_gpu);

    printf("eigenvalue = (matlab base-1), ascending order\n");
    for(int i = 0 ; i < m ; i++){
        printf("W[%d] = %E\n", i+1, W[i]);
    }

    printf("V = (matlab base-1)\n");
    printMatrix(m, m, V, lda, "V");
    printf("=====\n");

// step 4: check eigenvalues
double lambda_sup = 0;
for(int i = 0 ; i < m ; i++){
    double error = fabs( lambda[i] - W[i]);
    lambda_sup = (lambda_sup > error)? lambda_sup : error;
}
printf("|lambda - W| = %E\n", lambda_sup);

// free resources
if (d_A ) cudaFree(d_A);
if (d_B ) cudaFree(d_B);
if (d_W ) cudaFree(d_W);
if (devInfo) cudaFree(devInfo);
if (d_work ) cudaFree(d_work);

if (cusolverH) cusolverDnDestroy(cusolverH);

cudaDeviceReset();

return 0;
}

```

## F.5. Standard Symmetric Dense Eigenvalue Solver (via Jacobi method)

This chapter provides a simple example in the C programming language of how to use **syevj** to compute the spectrum of a dense symmetric system by

$$Ax = \lambda x$$

where A is a 3x3 dense symmetric matrix

$$A = \begin{pmatrix} 3.5 & 0.5 & 0 \\ 0.5 & 3.5 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

The following code uses **syevj** to compute eigenvalues and eigenvectors, then compare to exact eigenvalues {2,3,4}.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include syevj_example.cpp
 *   g++ -o syevj_example syevj_example.o -L/usr/local/cuda/lib64 -lcusolver -
lcudart
 */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cudaStream_t stream = NULL;
    syevjInfo_t syevj_params = NULL;

    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    const int m = 3;
    const int lda = m;
/*
 *   | 3.5 0.5 0 |
 *   A = | 0.5 3.5 0 |
 *       | 0   0  2 |
 *
 */
    double A[lda*m] = { 3.5, 0.5, 0, 0.5, 3.5, 0, 0, 0, 2.0};
    double lambda[m] = { 2.0, 3.0, 4.0};

    double V[lda*m]; /* eigenvectors */
    double W[m];     /* eigenvalues */

    double *d_A = NULL; /* device copy of A */
    double *d_W = NULL; /* eigenvalues */
    int *d_info = NULL; /* error info */
    int lwork = 0;     /* size of workspace */
    double *d_work = NULL; /* device workspace for syevj */
    int info = 0;     /* host copy of error info */

/* configuration of syevj */
    const double tol = 1.e-7;
    const int max_sweeps = 15;
    const cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute
eigenvectors.
    const cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;

```

## configure parameters of syevj

```

/* numerical results of syevj */
double residual = 0;
int executed_sweeps = 0;

printf("example of syevj \n");
printf("tol = %E, default value is machine zero \n", tol);
printf("max. sweeps = %d, default value is 100\n", max_sweeps);

printf("A = (matlab base-1)\n");
printMatrix(m, m, A, lda, "A");
printf("=====\n");

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusolverDnSetStream(cusolverH, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: configuration of syevj */
status = cusolverDnCreateSyevjInfo(&syevj_params);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of tolerance is machine zero */
status = cusolverDnXsyevjSetTolerance(
    syevj_params,
    tol);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of max. sweeps is 100 */
status = cusolverDnXsyevjSetMaxSweeps(
    syevj_params,
    max_sweeps);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 3: copy A to device */
cudaStat1 = cudaMalloc ((void**)&d_A, sizeof(double) * lda * m);
cudaStat2 = cudaMalloc ((void**)&d_W, sizeof(double) * m);
cudaStat3 = cudaMalloc ((void**)&d_info, sizeof(int));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*m,
    cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

```

## call eigenvalue solver

```

/* step 4: query working space of syevj */
status = cusolverDnDsyejv_bufferSize(
    cusolverH,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_W,
    &lwork,
    syevj_params);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaMalloc((void**) &d_work, sizeof(double)*lwork);
assert(cudaSuccess == cudaStat1);

/* step 5: compute eigen-pair */
status = cusolverDnDsyejv(
    cusolverH,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_W,
    d_work,
    lwork,
    d_info,
    syevj_params);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == status);
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(W, d_W, sizeof(double)*m, cudaMemcpyDeviceToHost);
cudaStat2 = cudaMemcpy(V, d_A, sizeof(double)*lda*m,
    cudaMemcpyDeviceToHost);
cudaStat3 = cudaMemcpy(&info, d_info, sizeof(int), cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);

if ( 0 == info ){
    printf("syevj converges \n");
}else if ( 0 > info ){
    printf("%d-th parameter is wrong \n", -info);
    exit(1);
}else{
    printf("WARNING: info = %d : syevj does not converge \n", info );
}

printf("Eigenvalue = (matlab base-1), ascending order\n");
for(int i = 0 ; i < m ; i++){
    printf("W[%d] = %E\n", i+1, W[i]);
}

printf("V = (matlab base-1)\n");
printMatrix(m, m, V, lda, "V");
printf("=====\n");

```

check the result

```

/* step 6: check eigenvalues */
double lambda_sup = 0;
for(int i = 0 ; i < m ; i++){
    double error = fabs( lambda[i] - W[i]);
    lambda_sup = (lambda_sup > error)? lambda_sup : error;
}
printf("|lambda - W| = %E\n", lambda_sup);

status = cusolverDnXsyevjGetSweeps(
    cusolverH,
    syevj_params,
    &executed_sweeps);
assert(CUSOLVER_STATUS_SUCCESS == status);

status = cusolverDnXsyevjGetResidual(
    cusolverH,
    syevj_params,
    &residual);
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("residual |A - V*W*V**H|_F = %E \n", residual );
printf("number of executed sweeps = %d \n", executed_sweeps );

/* free resources */
if (d_A ) cudaFree(d_A);
if (d_W ) cudaFree(d_W);
if (d_info ) cudaFree(d_info);
if (d_work ) cudaFree(d_work);

if (cusolverH ) cusolverDnDestroy(cusolverH);
if (stream ) cudaStreamDestroy(stream);
if (syevj_params) cusolverDnDestroySyevjInfo(syevj_params);

cudaDeviceReset();

return 0;
}

```

## F.6. Generalized Symmetric-Definite Dense Eigenvalue Solver (via Jacobi method)

This chapter provides a simple example in the C programming language of how to use **sygvj** to compute spectrum of a pair of dense symmetric matrices (A,B) by

$$Ax = \lambda Bx$$

where A is a 3x3 dense symmetric matrix

$$A = \begin{pmatrix} 3.5 & 0.5 & 0 \\ 0.5 & 3.5 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

and B is a 3x3 positive definite matrix

$$B = \begin{pmatrix} 10 & 2 & 3 \\ 2 & 10 & 5 \\ 3 & 5 & 10 \end{pmatrix}$$

The following code uses `sygvj` to compute eigenvalues and eigenvectors.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include sygvj_example.cpp
 *   g++ -o sygvj_example sygvj_example.o -L/usr/local/cuda/lib64 -lcusolver -
 *   lcudart
 */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cudaStream_t stream = NULL;
    syevjInfo_t syevj_params = NULL;
    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    const int m = 3;
    const int lda = m;

    /*
    *   | 3.5 0.5 0 |
    *   A = | 0.5 3.5 0 |
    *   | 0 0 2 |
    *
    *   | 10 2 3 |
    *   B = | 2 10 5 |
    *   | 3 5 10 |
    */
    double A[lda*m] = { 3.5, 0.5, 0, 0.5, 3.5, 0, 0, 0, 2.0};
    double B[lda*m] = { 10.0, 2.0, 3.0, 2.0, 10.0, 5.0, 3.0, 5.0, 10.0};
    double lambda[m] = { 0.158660256604, 0.370751508101882, 0.6};

    double V[lda*m]; /* eigenvectors */
    double W[m]; /* eigenvalues */

    double *d_A = NULL; /* device copy of A */
    double *d_B = NULL; /* device copy of B */
    double *d_W = NULL; /* numerical eigenvalue */
    int *d_info = NULL; /* error info */
    int lwork = 0; /* size of workspace */
    double *d_work = NULL; /* device workspace for sygvj */
    int info = 0; /* host copy of error info */

```

## configure parameters of Jacobi method

```

/* configuration of sygvj */
const double tol = 1.e-7;
const int max_sweeps = 15;
const cusolverEigType_t itype = CUSOLVER_EIG_TYPE_1; // A*x = (lambda)*B*x
const cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute
eigenvectors.
const cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;

/* numerical results of syevj */
double residual = 0;
int executed_sweeps = 0;

printf("example of sygvj \n");
printf("tol = %E, default value is machine zero \n", tol);
printf("max. sweeps = %d, default value is 100\n", max_sweeps);

printf("A = (matlab base-1)\n");
printMatrix(m, m, A, lda, "A");
printf("=====\n");

printf("B = (matlab base-1)\n");
printMatrix(m, m, B, lda, "B");
printf("=====\n");

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusolverDnSetStream(cusolverH, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: configuration of syevj */
status = cusolverDnCreateSyevjInfo(&syevj_params);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of tolerance is machine zero */
status = cusolverDnXsyevjSetTolerance(
    syevj_params,
    tol);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of max. sweeps is 100 */
status = cusolverDnXsyevjSetMaxSweeps(
    syevj_params,
    max_sweeps);
assert(CUSOLVER_STATUS_SUCCESS == status);

```

## call eigenvalue solver

```

/* step 3: copy A and B to device */
    cudaStat1 = cudaMalloc ((void**) &d_A, sizeof(double) * lda * m);
    cudaStat2 = cudaMalloc ((void**) &d_B, sizeof(double) * lda * m);
    cudaStat3 = cudaMalloc ((void**) &d_W, sizeof(double) * m);
    cudaStat4 = cudaMalloc ((void**) &d_info, sizeof(int));
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    assert(cudaSuccess == cudaStat4);

    cudaStat1 = cudaMemcpy(d_A, A, sizeof(double) * lda * m,
        cudaMemcpyHostToDevice);
    cudaStat2 = cudaMemcpy(d_B, B, sizeof(double) * lda * m,
        cudaMemcpyHostToDevice);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);

/* step 4: query working space of sygvj */
    status = cusolverDnDsygvj_bufferSize(
        cusolverH,
        itype,
        jobz,
        uplo,
        m,
        d_A,
        lda,
        d_B,
        lda, /* ldb */
        d_W,
        &lwork,
        syevj_params);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    cudaStat1 = cudaMalloc((void**) &d_work, sizeof(double) * lwork);
    assert(cudaSuccess == cudaStat1);

/* step 5: compute spectrum of (A,B) */
    status = cusolverDnDsygvj(
        cusolverH,
        itype,
        jobz,
        uplo,
        m,
        d_A,
        lda,
        d_B,
        lda, /* ldb */
        d_W,
        d_work,
        lwork,
        d_info,
        syevj_params);
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(W, d_W, sizeof(double) * m, cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_A, sizeof(double) * lda * m,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(&info, d_info, sizeof(int), cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);

```

## check the result

```

    if ( 0 == info ){
        printf("sygvj converges \n");
    }else if ( 0 > info ){
        printf("Error: %d-th parameter is wrong \n", -info);
        exit(1);
    }else if ( m >= info ){
        printf("Error: leading minor of order %d of B is not positive definite
\n", -info);
        exit(1);
    }else { /* info = m+1 */
        printf("WARNING: info = %d : sygvj does not converge \n", info );
    }

    printf("Eigenvalue = (matlab base-1), ascending order\n");
    for(int i = 0 ; i < m ; i++){
        printf("W[%d] = %E\n", i+1, W[i]);
    }

    printf("V = (matlab base-1)\n");
    printMatrix(m, m, V, lda, "V");
    printf("=====\n");

/* step 6: check eigenvalues */
    double lambda_sup = 0;
    for(int i = 0 ; i < m ; i++){
        double error = fabs( lambda[i] - W[i]);
        lambda_sup = (lambda_sup > error)? lambda_sup : error;
    }
    printf("|lambda - W| = %E\n", lambda_sup);

    status = cusolverDnXsyevjGetSweeps(
        cusolverH,
        syevj_params,
        &executed_sweeps);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    status = cusolverDnXsyevjGetResidual(
        cusolverH,
        syevj_params,
        &residual);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    printf("residual |M - V*W*V**H|_F = %E \n", residual );
    printf("number of executed sweeps = %d \n", executed_sweeps );

/* free resources */
    if (d_A ) cudaFree(d_A);
    if (d_B ) cudaFree(d_B);
    if (d_W ) cudaFree(d_W);
    if (d_info ) cudaFree(d_info);
    if (d_work ) cudaFree(d_work);
    if (cusolverH) cusolverDnDestroy(cusolverH);
    if (stream ) cudaStreamDestroy(stream);
    if (syevj_params) cusolverDnDestroySyevjInfo(syevj_params);

    cudaDeviceReset();
    return 0;
}

```

## F.7. batch eigenvalue solver for dense symmetric matrix

This chapter provides a simple example in the C programming language of how to use `syevjBatched` to compute the spectrum of a sequence of dense symmetric matrices by

$$A_j x = \lambda x$$

where  $A_0$  and  $A_1$  are 3x3 dense symmetric matrices

$$A_0 = \begin{pmatrix} 1 & -1 & 0 \\ -1 & 2 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$A_1 = \begin{pmatrix} 3 & 4 & 0 \\ 4 & 7 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

The following code uses `syevjBatched` to compute eigenvalues and eigenvectors

$$A_j = V_j^* W_j^* V_j^T$$

The user can disable/enable sorting by the function `cusolverDnXsyevjSetSortEig`.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include batchsyevj_example.cpp
 *   g++ -o batchsyevj_example batchsyevj_example.o -L/usr/local/cuda/lib64 -
lcusolver -lcudart
 *
 */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cudaStream_t stream = NULL;
    syevjInfo_t syevj_params = NULL;

    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    const int m = 3; // 1<= m <= 32
    const int lda = m;
    const int batchSize = 2;

/*
 *      |  1  -1  0  |
 *   A0 = | -1  2  0  |
 *      |  0  0  0  |
 *
 *   A0 = V0 * W0 * V0**T
 *
 *   W0 = diag(0, 0.3820, 2.6180)
 *
 *      |  3  4  0  |
 *   A1 = |  4  7  0  |
 *      |  0  0  0  |
 *
 *   A1 = V1 * W1 * V1**T
 *
 *   W1 = diag(0, 0.5279, 9.4721)
 */

```

## setup matrices A0 and A1

```

double A[lda*m*batchSize]; /* A = [A0 ; A1] */
double V[lda*m*batchSize]; /* V = [V0 ; V1] */
double W[m*batchSize];     /* W = [W0 ; W1] */
int info[batchSize];       /* info = [info0 ; info1] */

double *d_A = NULL; /* lda-by-m-by-batchSize */
double *d_W = NULL; /* m-by-batchSize */
int* d_info = NULL; /* batchSize */
int lwork = 0; /* size of workspace */
double *d_work = NULL; /* device workspace for syevjBatched */

const double tol = 1.e-7;
const int max_sweeps = 15;
const int sort_eig = 0; /* don't sort eigenvalues */
const cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; /* compute
eigenvectors */
const cublasFillMode_t uplo = CUBLAS_FILL_MODE_LOWER;

/* residual and executed_sweeps are not supported on syevjBatched */
double residual = 0;
int executed_sweeps = 0;

double *A0 = A;
double *A1 = A + lda*m;
/*
 *      | 1 -1 0 |
 *  A0 = | -1 2 0 |
 *      | 0 0 0 |
 *  A0 is column-major
 */
A0[0 + 0*lda] = 1.0;
A0[1 + 0*lda] = -1.0;
A0[2 + 0*lda] = 0.0;

A0[0 + 1*lda] = -1.0;
A0[1 + 1*lda] = 2.0;
A0[2 + 1*lda] = 0.0;

A0[0 + 2*lda] = 0.0;
A0[1 + 2*lda] = 0.0;
A0[2 + 2*lda] = 0.0;
/*
 *      | 3 4 0 |
 *  A1 = | 4 7 0 |
 *      | 0 0 0 |
 *  A1 is column-major
 */
A1[0 + 0*lda] = 3.0;
A1[1 + 0*lda] = 4.0;
A1[2 + 0*lda] = 0.0;

A1[0 + 1*lda] = 4.0;
A1[1 + 1*lda] = 7.0;
A1[2 + 1*lda] = 0.0;

A1[0 + 2*lda] = 0.0;
A1[1 + 2*lda] = 0.0;
A1[2 + 2*lda] = 0.0;

```

## configure parameters of syevj

```

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusolverDnSetStream(cusolverH, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: configuration of syevj */
status = cusolverDnCreateSyevjInfo(&syevj_params);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of tolerance is machine zero */
status = cusolverDnXsyevjSetTolerance(
    syevj_params,
    tol);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of max. sweeps is 100 */
status = cusolverDnXsyevjSetMaxSweeps(
    syevj_params,
    max_sweeps);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* disable sorting */
status = cusolverDnXsyevjSetSortEig(
    syevj_params,
    sort_eig);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 3: copy A to device */
cudaStat1 = cudaMalloc ((void**)&d_A , sizeof(double) * lda * m *
batchSize);
cudaStat2 = cudaMalloc ((void**)&d_W , sizeof(double) * m * batchSize);
cudaStat3 = cudaMalloc ((void**)&d_info, sizeof(int ) * batchSize);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double) * lda * m * batchSize,
cudaMemcpyHostToDevice);
cudaStat2 = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

/* step 4: query working space of syevjBatched */
status = cusolverDnDsyevjBatched_bufferSize(
    cusolverH,
    jobz,
    uplo,
    m,
    d_A,
    lda,
    d_W,
    &lwork,
    syevj_params,
    batchSize
);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaMalloc((void**)&d_work, sizeof(double)*lwork);
assert(cudaSuccess == cudaStat1);

```

## call eigenvalue solver

```

/* step 5: compute spectrum of A0 and A1 */
    status = cusolverDnDsyevejBatched(
        cusolverH,
        jobz,
        uplo,
        m,
        d_A,
        lda,
        d_W,
        d_work,
        lwork,
        d_info,
        syevj_params,
        batchSize
    );
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(V      , d_A      , sizeof(double) * lda * m * batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(W      , d_W      , sizeof(double) * m * batchSize      ,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(&info, d_info, sizeof(int) * batchSize      ,
        cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);

    for(int i = 0 ; i < batchSize ; i++){
        if ( 0 == info[i] ){
            printf("matrix %d: syevj converges \n", i);
        }else if ( 0 > info[i] ){
/* only info[0] shows if some input parameter is wrong.
 * If so, the error is CUSOLVER_STATUS_INVALID_VALUE.
 */
            printf("Error: %d-th parameter is wrong \n", -info[i] );
            exit(1);
        }else { /* info = m+1 */
/* if info[i] is not zero, Jacobi method does not converge at i-th matrix. */
            printf("WARNING: matrix %d, info = %d : sygvj does not converge \n",
                i, info[i] );
        }
    }

/* Step 6: show eigenvalues and eigenvectors */
    double *W0 = W;
    double *W1 = W + m;
    printf("==== \n");
    for(int i = 0 ; i < m ; i++){
        printf("W0[%d] = %f\n", i, W0[i]);
    }
    printf("==== \n");
    for(int i = 0 ; i < m ; i++){
        printf("W1[%d] = %f\n", i, W1[i]);
    }
    printf("==== \n");

    double *V0 = V;
    double *V1 = V + lda*m;
    printf("V0 = (matlab base-1)\n");
    printMatrix(m, m, V0, lda, "V0");
    printf("V1 = (matlab base-1)\n");
    printMatrix(m, m, V1, lda, "V1");

```

cannot query residual and executed sweeps.

```

/*
 * The following two functions do not support batched version.
 * The error CUSOLVER_STATUS_NOT_SUPPORTED is returned.
 */
    status = cusolverDnXsyevjGetSweeps(
        cusolverH,
        syevj_params,
        &executed_sweeps);
    assert(CUSOLVER_STATUS_NOT_SUPPORTED == status);

    status = cusolverDnXsyevjGetResidual(
        cusolverH,
        syevj_params,
        &residual);
    assert(CUSOLVER_STATUS_NOT_SUPPORTED == status);

/* free resources */
    if (d_A ) cudaFree(d_A);
    if (d_W ) cudaFree(d_W);
    if (d_info ) cudaFree(d_info);
    if (d_work ) cudaFree(d_work);

    if (cusolverH) cusolverDnDestroy(cusolverH);
    if (stream ) cudaStreamDestroy(stream);
    if (syevj_params) cusolverDnDestroySyevjInfo(syevj_params);

    cudaDeviceReset();

    return 0;
}

```

# Appendix G.

## EXAMPLES OF SINGULAR VALUE DECOMPOSITION

### G.1. SVD with singular vectors

This chapter provides a simple example in the C programming language of how to perform singular value decomposition.

$$A = U * \Sigma * V^H$$

A is a 3x2 dense matrix,

$$A = \begin{pmatrix} 1.0 & 2.0 \\ 4.0 & 5.0 \\ 2.0 & 1.0 \end{pmatrix}$$

The following code uses three steps:

Step 1: compute  $A = U * S * V^T$

Step 2: check accuracy of singular value

Step 3: measure residual  $A - U * S * V^T$

```

...

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 * nvcc -c -I/usr/local/cuda/include svd_example.cpp
 * g++ -o a.out svd_example.o -I/usr/local/cuda/lib64 -lcudart -lcublas -
lcusolver
 *
 */

#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cublas_v2.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cublasHandle_t cublasH = NULL;
    cublasStatus_t cublas_status = CUBLAS_STATUS_SUCCESS;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;
    cudaError_t cudaStat6 = cudaSuccess;
    const int m = 3;
    const int n = 2;
    const int lda = m;
    /*
     *   | 1 2 |
     *   A = | 4 5 |
     *       | 2 1 |
     */
    double A[lda*n] = { 1.0, 4.0, 2.0, 2.0, 5.0, 1.0};
    double U[lda*m]; // m-by-m unitary matrix
    double VT[lda*n]; // n-by-n unitary matrix
    double S[n]; // singular value
    double S_exact[n] = {7.065283497082729, 1.040081297712078};

    double *d_A = NULL;
    double *d_S = NULL;
    double *d_U = NULL;
    double *d_VT = NULL;
    int *devInfo = NULL;
    double *d_work = NULL;
    double *d_rwork = NULL;
    double *d_W = NULL; // W = S*VT

    int lwork = 0;
    int info_gpu = 0;
    const double h_one = 1;
    const double h_minus_one = -1;

```

...

```

printf("A = (matlab base-1)\n");
printMatrix(m, n, A, lda, "A");
printf("=====\n");

// step 1: create cusolverDn/cublas handle
cusolver_status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

cublas_status = cublasCreate(&cublasH);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

// step 2: copy A and B to device
cudaStat1 = cudaMalloc ((void**) &d_A , sizeof(double)*lda*n);
cudaStat2 = cudaMalloc ((void**) &d_S , sizeof(double)*n);
cudaStat3 = cudaMalloc ((void**) &d_U , sizeof(double)*lda*m);
cudaStat4 = cudaMalloc ((void**) &d_VT , sizeof(double)*lda*n);
cudaStat5 = cudaMalloc ((void**) &devInfo, sizeof(int));
cudaStat6 = cudaMalloc ((void**) &d_W , sizeof(double)*lda*n);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);
assert(cudaSuccess == cudaStat5);
assert(cudaSuccess == cudaStat6);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*n,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

// step 3: query working space of SVD
cusolver_status = cusolverDnDgesvd_bufferSize(
    cusolverH,
    m,
    n,
    &lwork );
assert (cusolver_status == CUSOLVER_STATUS_SUCCESS);

cudaStat1 = cudaMalloc((void**) &d_work , sizeof(double)*lwork);
assert(cudaSuccess == cudaStat1);

// step 4: compute SVD
signed char jobu = 'A'; // all m columns of U
signed char jobvt = 'A'; // all n columns of VT
cusolver_status = cusolverDnDgesvd (
    cusolverH,
    jobu,
    jobvt,
    m,
    n,
    d_A,
    lda,
    d_S,
    d_U,
    lda, // ldu
    d_VT,
    lda, // ldvt,
    d_work,
    lwork,
    d_rwork,
    devInfo);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

```

...

```

    cudaStat1 = cudaMemcpy(U , d_U , sizeof(double)*lda*m,
cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(VT, d_VT, sizeof(double)*lda*n,
cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(S , d_S , sizeof(double)*n ,
cudaMemcpyDeviceToHost);
    cudaStat4 = cudaMemcpy(&info_gpu, devInfo, sizeof(int),
cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    assert(cudaSuccess == cudaStat4);

    printf("after gesvd: info_gpu = %d\n", info_gpu);
    assert(0 == info_gpu);
    printf("=====\n");

    printf("S = (matlab base-1)\n");
    printMatrix(n, 1, S, lda, "S");
    printf("=====\n");

    printf("U = (matlab base-1)\n");
    printMatrix(m, m, U, lda, "U");
    printf("=====\n");

    printf("VT = (matlab base-1)\n");
    printMatrix(n, n, VT, lda, "VT");
    printf("=====\n");

// step 5: measure error of singular value
double ds_sup = 0;
for(int j = 0; j < n; j++){
    double err = fabs( S[j] - S_exact[j] );
    ds_sup = (ds_sup > err)? ds_sup : err;
}
printf("|S - S_exact| = %E \n", ds_sup);

// step 6: |A - U*S*VT|
// W = S*VT
cublas_status = cublasDdggmm(
    cublasH,
    CUBLAS_SIDE_LEFT,
    n,
    n,
    d_VT,
    lda,
    d_S,
    1,
    d_W,
    lda);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

```

...

```

// A := -U*W + A
cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*n,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cublas_status = cublasDgemv_v2(
    cublasH,
    CUBLAS_OP_N, // U
    CUBLAS_OP_N, // W
    m, // number of rows of A
    n, // number of columns of A
    n, // number of columns of U
    &h_minus_one, /* host pointer */
    d_U, // U
    lda,
    d_W, // W
    lda,
    &h_one, /* hostpointer */
    d_A,
    lda);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

double dR_fro = 0.0;
cublas_status = cublasDnrm2_v2(
    cublasH, lda*n, d_A, 1, &dR_fro);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

printf("|A - U*S*VT| = %E \n", dR_fro);

// free resources
if (d_A ) cudaFree(d_A);
if (d_S ) cudaFree(d_S);
if (d_U ) cudaFree(d_U);
if (d_VT ) cudaFree(d_VT);
if (devInfo) cudaFree(devInfo);
if (d_work ) cudaFree(d_work);
if (d_rwork) cudaFree(d_rwork);
if (d_W ) cudaFree(d_W);

if (cublasH ) cublasDestroy(cublasH);
if (cusolverH) cusolverDnDestroy(cusolverH);

cudaDeviceReset();

return 0;
}

```

## G.2. SVD with singular vectors (via Jacobi method)

This chapter provides a simple example in the C programming language of how to perform singular value decomposition by `gesvdj`.

$$A = U * \Sigma * V^H$$

A is a 3x2 dense matrix,

$$A = \begin{pmatrix} 1.0 & 2.0 \\ 4.0 & 5.0 \\ 2.0 & 1.0 \end{pmatrix}$$

```

...
/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 * nvcc -c -I/usr/local/cuda/include gesvdj_example.cpp
 * g++ -o gesvdj_example gesvdj_example.o -I/usr/local/cuda/lib64 -lcudart -
lcusolver
 */
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %20.16E\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cudaStream_t stream = NULL;
    gesvdjInfo_t gesvdj_params = NULL;

    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;
    const int m = 3;
    const int n = 2;
    const int lda = m;
/*
 * | 1 2 |
 * A = | 4 5 |
 * | 2 1 |
 */
    double A[lda*n] = { 1.0, 4.0, 2.0, 2.0, 5.0, 1.0};
    double U[lda*m]; /* m-by-m unitary matrix, left singular vectors */
    double V[lda*n]; /* n-by-n unitary matrix, right singular vectors */
    double S[n]; /* numerical singular value */
/* exact singular values */
    double S_exact[n] = {7.065283497082729, 1.040081297712078};
    double *d_A = NULL; /* device copy of A */
    double *d_S = NULL; /* singular values */
    double *d_U = NULL; /* left singular vectors */
    double *d_V = NULL; /* right singular vectors */
    int *d_info = NULL; /* error info */
    int lwork = 0; /* size of workspace */
    double *d_work = NULL; /* devie workspace for gesvdj */
    int info = 0; /* host copy of error info */

```

...

```

/* configuration of gesvdj */
const double tol = 1.e-7;
const int max_sweeps = 15;
const cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute
eigenvectors.
const int econ = 0 ; /* econ = 1 for economy size */

/* numerical results of gesvdj */
double residual = 0;
int executed_sweeps = 0;

printf("example of gesvdj \n");
printf("tol = %E, default value is machine zero \n", tol);
printf("max. sweeps = %d, default value is 100\n", max_sweeps);
printf("econ = %d \n", econ);

printf("A = (matlab base-1)\n");
printMatrix(m, n, A, lda, "A");
printf("=====\n");

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusolverDnSetStream(cusolverH, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: configuration of gesvdj */
status = cusolverDnCreateGesvdjInfo(&gesvdj_params);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of tolerance is machine zero */
status = cusolverDnXgesvdjSetTolerance(
    gesvdj_params,
    tol);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of max. sweeps is 100 */
status = cusolverDnXgesvdjSetMaxSweeps(
    gesvdj_params,
    max_sweeps);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 3: copy A and B to device */
cudaStat1 = cudaMalloc ((void**) &d_A , sizeof(double)*lda*n);
cudaStat2 = cudaMalloc ((void**) &d_S , sizeof(double)*n);
cudaStat3 = cudaMalloc ((void**) &d_U , sizeof(double)*lda*m);
cudaStat4 = cudaMalloc ((void**) &d_V , sizeof(double)*lda*n);
cudaStat5 = cudaMalloc ((void**) &d_info, sizeof(int));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);
assert(cudaSuccess == cudaStat5);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*n,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

```

...

```

/* step 4: query workspace of SVD */
    status = cusolverDnDgesvdj_bufferSize(
        cusolverH,
        jobz, /* CUSOLVER_EIG_MODE_NOVECTOR: compute singular values only */
              /* CUSOLVER_EIG_MODE_VECTOR: compute singular value and singular
vectors */
        econ, /* econ = 1 for economy size */
        m,    /* nubmer of rows of A, 0 <= m */
        n,    /* number of columns of A, 0 <= n */
        d_A,  /* m-by-n */
        lda,  /* leading dimension of A */
        d_S,  /* min(m,n) */
              /* the singular values in descending order */
        d_U,  /* m-by-m if econ = 0 */
              /* m-by-min(m,n) if econ = 1 */
        lda,  /* leading dimension of U, ldu >= max(1,m) */
        d_V,  /* n-by-n if econ = 0 */
              /* n-by-min(m,n) if econ = 1 */
        lda,  /* leading dimension of V, ldv >= max(1,n) */
        &lwork,
        gesvdj_params);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    cudaStat1 = cudaMalloc((void**) &d_work, sizeof(double)*lwork);
    assert(cudaSuccess == cudaStat1);

/* step 5: compute SVD */
    status = cusolverDnDgesvdj(
        cusolverH,
        jobz, /* CUSOLVER_EIG_MODE_NOVECTOR: compute singular values only */
              /* CUSOLVER_EIG_MODE_VECTOR: compute singular value and singular
vectors */
        econ, /* econ = 1 for economy size */
        m,    /* nubmer of rows of A, 0 <= m */
        n,    /* number of columns of A, 0 <= n */
        d_A,  /* m-by-n */
        lda,  /* leading dimension of A */
        d_S,  /* min(m,n) */
              /* the singular values in descending order */
        d_U,  /* m-by-m if econ = 0 */
              /* m-by-min(m,n) if econ = 1 */
        lda,  /* leading dimension of U, ldu >= max(1,m) */
        d_V,  /* n-by-n if econ = 0 */
              /* n-by-min(m,n) if econ = 1 */
        lda,  /* leading dimension of V, ldv >= max(1,n) */
        d_work,
        lwork,
        d_info,
        gesvdj_params);
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(U, d_U, sizeof(double)*lda*m,
        cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_V, sizeof(double)*lda*n,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(S, d_S, sizeof(double)*n,
        cudaMemcpyDeviceToHost);
    cudaStat4 = cudaMemcpy(&info, d_info, sizeof(int), cudaMemcpyDeviceToHost);
    cudaStat5 = cudaDeviceSynchronize();
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    assert(cudaSuccess == cudaStat4);
    assert(cudaSuccess == cudaStat5);

```

...

```

if ( 0 == info ){
    printf("gesvdj converges \n");
}else if ( 0 > info ){
    printf("%d-th parameter is wrong \n", -info);
    exit(1);
}else{
    printf("WARNING: info = %d : gesvdj does not converge \n", info );
}

printf("S = singular values (matlab base-1)\n");
printMatrix(n, 1, S, lda, "S");
printf("=====\n");

printf("U = left singular vectors (matlab base-1)\n");
printMatrix(m, m, U, lda, "U");
printf("=====\n");

printf("V = right singular vectors (matlab base-1)\n");
printMatrix(n, n, V, lda, "V");
printf("=====\n");

/* step 6: measure error of singular value */
double ds_sup = 0;
for(int j = 0; j < n; j++){
    double err = fabs( S[j] - S_exact[j] );
    ds_sup = (ds_sup > err)? ds_sup : err;
}
printf("|S - S_exact|_sup = %E \n", ds_sup);

status = cusolverDnXgesvdjGetSweeps(
    cusolverH,
    gesvdj_params,
    &executed_sweeps);
assert(CUSOLVER_STATUS_SUCCESS == status);

status = cusolverDnXgesvdjGetResidual(
    cusolverH,
    gesvdj_params,
    &residual);
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("residual |A - U*S*V**H|_F = %E \n", residual );
printf("number of executed sweeps = %d \n", executed_sweeps );

/* free resources */
if (d_A ) cudaFree(d_A);
if (d_S ) cudaFree(d_S);
if (d_U ) cudaFree(d_U);
if (d_V ) cudaFree(d_V);
if (d_info) cudaFree(d_info);
if (d_work ) cudaFree(d_work);

if (cusolverH) cusolverDnDestroy(cusolverH);
if (stream ) cudaStreamDestroy(stream);
if (gesvdj_params) cusolverDnDestroyGesvdjInfo(gesvdj_params);

cudaDeviceReset();
return 0;
}

```

## G.3. batch dense SVD solver

This chapter provides a simple example in the C programming language of how to use **gesvdjBatched** to compute the SVD of a sequence of dense matrices

$$A_j = U_j \Sigma_j V_j^H$$

where A0 and A1 are 3x2 dense matrices

$$A_0 = \begin{pmatrix} 1 & -1 \\ -1 & 2 \\ 0 & 0 \end{pmatrix}$$

$$A_1 = \begin{pmatrix} 3 & 4 \\ 4 & 7 \\ 0 & 0 \end{pmatrix}$$

The following code uses **gesvdjBatched** to compute singular values and singular vectors.

The user can disable/enable sorting by the function `cusolverDnXgesvdjSetSortEig`.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include gesvdjbatch_example.cpp
 *   g++ -o gesvdjbatch_example gesvdjbatch_example.o -L/usr/local/cuda/lib64 -
lcusolver -lcudart
 */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %20.16E\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cudaStream_t stream = NULL;
    gesvdjInfo_t gesvdj_params = NULL;

    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;
    const int m = 3; /* 1 <= m <= 32 */
    const int n = 2; /* 1 <= n <= 32 */
    const int lda = m; /* lda >= m */
    const int ldu = m; /* ldu >= m */
    const int ldv = n; /* ldv >= n */
    const int batchSize = 2;
    const int minmn = (m < n)? m : n; /* min(m,n) */

    /*
     *   | 1 -1 |
     * A0 = | -1 2 |
     *   | 0 0 |
     *
     * A0 = U0 * S0 * V0**T
     * S0 = diag(2.6180, 0.382)
     *
     *   | 3 4 |
     * A1 = | 4 7 |
     *   | 0 0 |
     *
     * A1 = U1 * S1 * V1**T
     * S1 = diag(9.4721, 0.5279)
     */
}

```

## setup matrices A0 and A1

```

double A[lda*n*batchSize]; /* A = [A0 ; A1] */
double U[ldu*m*batchSize]; /* U = [U0 ; U1] */
double V[ldv*n*batchSize]; /* V = [V0 ; V1] */
double S[minmn*batchSize]; /* S = [S0 ; S1] */
int info[batchSize]; /* info = [info0 ; info1] */

double *d_A = NULL; /* lda-by-n-by-batchSize */
double *d_U = NULL; /* ldu-by-m-by-batchSize */
double *d_V = NULL; /* ldv-by-n-by-batchSize */
double *d_S = NULL; /* minmn-by-batchSize */
int* d_info = NULL; /* batchSize */
int lwork = 0; /* size of workspace */
double *d_work = NULL; /* device workspace for gesvdjBatched */

const double tol = 1.e-7;
const int max_sweeps = 15;
const int sort_svd = 0; /* don't sort singular values */
const cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; /* compute singular
vectors */

/* residual and executed_sweeps are not supported on gesvdjBatched */
double residual = 0;
int executed_sweeps = 0;

double *A0 = A;
double *A1 = A + lda*n; /* Aj is m-by-n */
/*
 *      | 1  -1 |
 *  A0 = | -1  2 |
 *      | 0   0 |
 *  A0 is column-major
 */
A0[0 + 0*lda] = 1.0;
A0[1 + 0*lda] = -1.0;
A0[2 + 0*lda] = 0.0;

A0[0 + 1*lda] = -1.0;
A0[1 + 1*lda] = 2.0;
A0[2 + 1*lda] = 0.0;

/*
 *      | 3  4 |
 *  A1 = | 4  7 |
 *      | 0  0 |
 *  A1 is column-major
 */
A1[0 + 0*lda] = 3.0;
A1[1 + 0*lda] = 4.0;
A1[2 + 0*lda] = 0.0;

A1[0 + 1*lda] = 4.0;
A1[1 + 1*lda] = 7.0;
A1[2 + 1*lda] = 0.0;

printf("example of gesvdjBatched \n");
printf("m = %d, n = %d \n", m, n);
printf("tol = %E, default value is machine zero \n", tol);
printf("max. sweeps = %d, default value is 100\n", max_sweeps);

printf("A0 = (matlab base-1)\n");
printMatrix(m, n, A0, lda, "A0");
printf("A1 = (matlab base-1)\n");
printMatrix(m, n, A1, lda, "A1");
printf("=====\n");

```

## configure parameters of gesvdj

```

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusolverDnSetStream(cusolverH, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: configuration of gesvdj */
status = cusolverDnCreateGesvdjInfo(&gesvdj_params);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of tolerance is machine zero */
status = cusolverDnXgesvdjSetTolerance(
    gesvdj_params,
    tol);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* default value of max. sweeps is 100 */
status = cusolverDnXgesvdjSetMaxSweeps(
    gesvdj_params,
    max_sweeps);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* disable sorting */
status = cusolverDnXgesvdjSetSortEig(
    gesvdj_params,
    sort_svd);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 3: copy A to device */
cudaStat1 = cudaMalloc ((void**) &d_A , sizeof(double)*lda*n*batchSize);
cudaStat2 = cudaMalloc ((void**) &d_U , sizeof(double)*ldu*m*batchSize);
cudaStat3 = cudaMalloc ((void**) &d_V , sizeof(double)*ldv*n*batchSize);
cudaStat4 = cudaMalloc ((void**) &d_S , sizeof(double)*minmn*batchSize);
cudaStat5 = cudaMalloc ((void**) &d_info, sizeof(int )*batchSize);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);
assert(cudaSuccess == cudaStat5);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*n*batchSize,
cudaMemcpyHostToDevice);
cudaStat2 = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

```

## call batched singular value solver

```

/* step 4: query working space of gesvdjBatched */
    status = cusolverDnDgesvdjBatched_bufferSize(
        cusolverH,
        jobz,
        m,
        n,
        d_A,
        lda,
        d_S,
        d_U,
        ldu,
        d_V,
        ldv,
        &lwork,
        gesvdj_params,
        batchSize
    );
    assert(CUSOLVER_STATUS_SUCCESS == status);

    cudaStat1 = cudaMalloc((void**) &d_work, sizeof(double)*lwork);
    assert(cudaSuccess == cudaStat1);

/* step 5: compute singular values of A0 and A1 */
    status = cusolverDnDgesvdjBatched(
        cusolverH,
        jobz,
        m,
        n,
        d_A,
        lda,
        d_S,
        d_U,
        ldu,
        d_V,
        ldv,
        d_work,
        lwork,
        d_info,
        gesvdj_params,
        batchSize
    );
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(U, d_U, sizeof(double)*ldu*m*batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_V, sizeof(double)*ldv*n*batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(S, d_S, sizeof(double)*minmn*batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat4 = cudaMemcpy(&info, d_info, sizeof(int) * batchSize,
        cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    assert(cudaSuccess == cudaStat4);

```

check the result

```

    for(int i = 0 ; i < batchSize ; i++){
        if ( 0 == info[i] ){
            printf("matrix %d: gesvdj converges \n", i);
        }else if ( 0 > info[i] ){
/* only info[0] shows if some input parameter is wrong.
 * If so, the error is CUSOLVER_STATUS_INVALID_VALUE.
 */
            printf("Error: %d-th parameter is wrong \n", -info[i] );
            exit(1);
        }else { /* info = m+1 */
/* if info[i] is not zero, Jacobi method does not converge at i-th matrix. */
            printf("WARNING: matrix %d, info = %d : gesvdj does not converge
\n", i, info[i] );
        }
    }

/* Step 6: show singular values and singular vectors */
double *S0 = S;
double *S1 = S + minmn;
printf("==== \n");
for(int i = 0 ; i < minmn ; i++){
    printf("S0(%d) = %20.16E\n", i+1, S0[i]);
}
printf("==== \n");
for(int i = 0 ; i < minmn ; i++){
    printf("S1(%d) = %20.16E\n", i+1, S1[i]);
}
printf("==== \n");

double *U0 = U;
double *U1 = U + ldu*m; /* Uj is m-by-m */
printf("U0 = (matlab base-1)\n");
printMatrix(m, m, U0, ldu, "U0");
printf("U1 = (matlab base-1)\n");
printMatrix(m, m, U1, ldu, "U1");

double *V0 = V;
double *V1 = V + ldv*n; /* Vj is n-by-n */
printf("V0 = (matlab base-1)\n");
printMatrix(n, n, V0, ldv, "V0");
printf("V1 = (matlab base-1)\n");
printMatrix(n, n, V1, ldv, "V1");

```

cannot query residual and executed sweeps

```

/*
 * The following two functions do not support batched version.
 * The error CUSOLVER_STATUS_NOT_SUPPORTED is returned.
 */
    status = cusolverDnXgesvdjGetSweeps(
        cusolverH,
        gesvdj_params,
        &executed_sweeps);
    assert(CUSOLVER_STATUS_NOT_SUPPORTED == status);

    status = cusolverDnXgesvdjGetResidual(
        cusolverH,
        gesvdj_params,
        &residual);
    assert(CUSOLVER_STATUS_NOT_SUPPORTED == status);

/* free resources */
    if (d_A ) cudaFree(d_A);
    if (d_U ) cudaFree(d_U);
    if (d_V ) cudaFree(d_V);
    if (d_S ) cudaFree(d_S);
    if (d_info ) cudaFree(d_info);
    if (d_work ) cudaFree(d_work);

    if (cusolverH) cusolverDnDestroy(cusolverH);
    if (stream ) cudaStreamDestroy(stream);
    if (gesvdj_params) cusolverDnDestroyGesvdjInfo(gesvdj_params);

    cudaDeviceReset();

    return 0;
}

```

## G.4. SVD approximation

This chapter provides a simple example in the C programming language of how to approximate singular value decomposition by **gesvdaStridedBatched**.

$$A = U * \Sigma * V^H$$

A0 and A1 are a 3x2 dense matrices,

$$A0 = \begin{pmatrix} 1.0 & 2.0 \\ 4.0 & 5.0 \\ 2.0 & 1.0 \end{pmatrix}$$

$$A1 = \begin{pmatrix} 10.0 & 9.0 \\ 8.0 & 7.0 \\ 6.0 & 5.0 \end{pmatrix}$$

```

...
/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 * nvcc -c -I/usr/local/cuda/include gesvda_example.cpp
 * g++ -o gesvda_example gesvda_example.o -I/usr/local/cuda/lib64 -lcudart -
lcusolver
 */

#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const float*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            float Areg = A[row + col*lda];
            printf("%s(%d,%d) = %20.16E\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cudaStream_t stream = NULL;

    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;
    const int batchSize = 2;
    const int m = 3;
    const int n = 2;
    const int lda = m;
    const int ldu = m;
    const int ldv = n;
    const int rank = n;
    const long long int strideA = (long long int)lda*n;
    const long long int strideS = n;
    const long long int strideU = (long long int)ldu*n;
    const long long int strideV = (long long int)ldv*n;
    /*
     * A0 = | 1 2 |, A1 = | 10 9 |
     *      | 4 5 |,      | 8 7 |
     *      | 2 1 |,      | 6 5 |
     */
    float A[strideA*batchSize] = { 1.0, 4.0, 2.0, 2.0, 5.0, 1.0, 10.0, 8.0, 6.0,
    9.0, 7.0, 5.0};
    float U[strideU*batchSize]; /* left singular vectors */
    float V[strideV*batchSize]; /* right singular vectors */
    float S[strideS*batchSize]; /* numerical singular value */

    /* exact singular values */
    float S_exact[strideS*batchSize] = {7.065283497082729, 1.040081297712078,
    18.839649186929730, 0.260035600289472};

```

...

```

float *d_A = NULL; /* device copy of A */
float *d_S = NULL; /* singular values */
float *d_U = NULL; /* left singular vectors */
float *d_V = NULL; /* right singular vectors */
int *d_info = NULL; /* error info */
int lwork = 0; /* size of workspace */
float *d_work = NULL; /* devie workspace for gesvda */
const cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR; // compute
eigenvectors.
double RnormF[batchSize]; /* residual norm */
int info[batchSize]; /* host copy of error info */

printf("example of gesvdaStridedBatched \n");
printf("A = (matlab base-1)\n");
printMatrix(m, n, A, lda, "A0");
printf("=====\n");
printMatrix(m, n, A + strideA, lda, "A1");
printf("=====\n");

/* step 1: create cusolver handle, bind a stream */
status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == status);
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);
status = cusolverDnSetStream(cusolverH, stream);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* step 2: copy A to device */
cudaStat1 = cudaMalloc ((void**) &d_A , sizeof(float)*strideA*batchSize);
cudaStat2 = cudaMalloc ((void**) &d_S , sizeof(float)*strideS*batchSize);
cudaStat3 = cudaMalloc ((void**) &d_U , sizeof(float)*strideU*batchSize);
cudaStat4 = cudaMalloc ((void**) &d_V , sizeof(float)*strideV*batchSize);
cudaStat5 = cudaMalloc ((void**) &d_info, sizeof(int)*batchSize);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);
assert(cudaSuccess == cudaStat5);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(float)*strideA*batchSize,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaDeviceSynchronize(); /* sync with null stream */

/* step 3: query workspace of SVD */
status = cusolverDnSgesvdaStridedBatched_bufferSize(
cusolverH,
jobz, /* CUSOLVER_EIG_MODE_NOVECTOR: compute singular values only */
/* CUSOLVER_EIG_MODE_VECTOR: compute singular value and singular
vectors */
rank, /* number of singular values */
m, /* nubmer of rows of Aj, 0 <= m */
n, /* number of columns of Aj, 0 <= n */
d_A, /* Aj is m-by-n */
lda, /* leading dimension of Aj */
strideA, /* >= lda*n */
d_S, /* Sj is rank-by-1, singular values in descending order */
strideS, /* >= rank */
d_U, /* Uj is m-by-rank */
ldu, /* leading dimension of Uj, ldu >= max(1,m) */
strideU, /* >= ldu*rank */
d_V, /* Vj is n-by-rank */
ldv, /* leading dimension of Vj, ldv >= max(1,n) */
strideV, /* >= ldv*rank */
&lwork,
batchSize /* number of matrices */
);
assert(CUSOLVER_STATUS_SUCCESS == status);

```

...

```

    cudaStat1 = cudaMalloc((void**) &d_work , sizeof(float)*lwork);
    assert(cudaSuccess == cudaStat1);

/* step 4: compute SVD */
    status = cusolverDnSgesvdaStridedBatched(
        cusolverH,
        jobz, /* CUSOLVER_EIG_MODE_NOVECTOR: compute singular values only */
              /* CUSOLVER_EIG_MODE_VECTOR: compute singular value and singular
vectors */
        rank, /* number of singular values */
        m,    /* number of rows of Aj, 0 <= m */
        n,    /* number of columns of Aj, 0 <= n */
        d_A,  /* Aj is m-by-n */
        lda,  /* leading dimension of Aj */
        strideA, /* >= lda*n */
        d_S,  /* Sj is rank-by-1 */
              /* the singular values in descending order */
        strideS, /* >= rank */
        d_U,  /* Uj is m-by-rank */
        ldu,  /* leading dimension of Uj, ldu >= max(1,m) */
        strideU, /* >= ldu*rank */
        d_V,  /* Vj is n-by-rank */
        ldv,  /* leading dimension of Vj, ldv >= max(1,n) */
        strideV, /* >= ldv*rank */
        d_work,
        lwork,
        d_info,
        RnormF,
        batchSize /* number of matrices */
    );
    cudaStat1 = cudaDeviceSynchronize();
    assert(CUSOLVER_STATUS_SUCCESS == status);
    assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(U, d_U, sizeof(float)*strideU*batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(V, d_V, sizeof(float)*strideV*batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(S, d_S, sizeof(float)*strideS*batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat4 = cudaMemcpy(info, d_info, sizeof(int)*batchSize,
        cudaMemcpyDeviceToHost);
    cudaStat5 = cudaDeviceSynchronize();
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    assert(cudaSuccess == cudaStat4);
    assert(cudaSuccess == cudaStat5);

    if ( 0 > info[0] ){
        printf("%d-th parameter is wrong \n", -info[0]);
        exit(1);
    }
    for(int idx = 0 ; idx < batchSize; idx++){
        if ( 0 == info[idx] ){
            printf("%d-th matrix, gesvda converges \n", idx );
        }else{
            printf("WARNING: info[%d] = %d : gesvda does not converge \n", idx,
info[idx] );
        }
    }

    printf("S = singular values (matlab base-1)\n");
    printf("U = left singular vectors (matlab base-1)\n");
    printf("V = right singular vectors (matlab base-1)\n\n");

```

...

```

printMatrix(rank, 1, S, n, "S0");
printf("=====\n");

printMatrix(m, rank, U, ldu, "U0");
printf("=====\n");

printMatrix(n, rank, V, ldv, "V0");
printf("=====\n");

float ds_sup = 0;
for(int j = 0; j < n; j++){
    float err = fabs( S[j] - S_exact[j] );
    ds_sup = (ds_sup > err)? ds_sup : err;
}
printf("|S0 - S0_exact|_sup = %E \n", ds_sup);

printf("residual |A0 - U0*S0*V0**H|_F = %E \n", RnormF[0] );

printMatrix(rank, 1, S + strideS, n, "S1");
printf("=====\n");

printMatrix(m, rank, U + strideU, ldu, "U1");
printf("=====\n");

printMatrix(n, rank, V + strideV, ldv, "V1");
printf("=====\n");

ds_sup = 0;
for(int j = 0; j < n; j++){
    float err = fabs( S[strideS + j] - S_exact[strideS + j] );
    ds_sup = (ds_sup > err)? ds_sup : err;
}
printf("|S1 - S1_exact|_sup = %E \n", ds_sup);

printf("residual |A1 - U1*S1*V1**H|_F = %E \n", RnormF[1] );

/* free resources */
if (d_A ) cudaFree(d_A);
if (d_S ) cudaFree(d_S);
if (d_U ) cudaFree(d_U);
if (d_V ) cudaFree(d_V);
if (d_info ) cudaFree(d_info);
if (d_work ) cudaFree(d_work);

if (cusolverH) cusolverDnDestroy(cusolverH);
if (stream ) cudaStreamDestroy(stream);

cudaDeviceReset();

return 0;
}

```

## G.5. 64-bit GESVD

This chapter provides a simple example in the C programming language of how to use `cusolverDnGesvd`.

$$A = U * \Sigma * V^H$$

A is a 3x2 dense matrix,

$$A = \begin{pmatrix} 1.0 & 2.0 \\ 4.0 & 5.0 \\ 2.0 & 1.0 \end{pmatrix}$$

The following code uses three steps:

Step 1: compute  $A = U*S*V^T$

Step 2: check accuracy of singular value

Step 3: measure residual  $A-U*S*V^T$

```

...
/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 * nvcc -c -I/usr/local/cuda/include svd64_example.cpp
 * nvcc -o a.out svd64_example.o -L/usr/local/cuda/lib64 -lcudart -lcublas -
lcusolver
 */

#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cublas_v2.h>
#include <cusolverDn.h>

void printMatrix(int m, int n, const double*A, int lda, const char* name)
{
    for(int row = 0 ; row < m ; row++){
        for(int col = 0 ; col < n ; col++){
            double Areg = A[row + col*lda];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusolverDnHandle_t cusolverH = NULL;
    cublasHandle_t cublasH = NULL;
    cublasStatus_t cublas_status = CUBLAS_STATUS_SUCCESS;
    cusolverStatus_t cusolver_status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;
    cudaError_t cudaStat6 = cudaSuccess;
    const int m = 3;
    const int n = 2;
    const int lda = m;
    /*
     *   | 1 2 |
     *   A = | 4 5 |
     *       | 2 1 |
     */
    double A[lda*n] = { 1.0, 4.0, 2.0, 2.0, 5.0, 1.0};
    double U[lda*m]; /* m-by-m unitary matrix */
    double VT[lda*n]; /* n-by-n unitary matrix */
    double S[n]; /* singular value */

    double S_exact[n] = {7.065283497082729, 1.040081297712078};

    double *d_A = NULL;
    double *d_S = NULL;
    double *d_U = NULL;
    double *d_VT = NULL;
    int *d_info = NULL;
    double *pBuffer = NULL;
    double *d_W = NULL; /* W = S*VT */
    size_t workspaceInBytes = 0 ;
    int info_gpu = 0;
    const double h_one = 1;
    const double h_minus_one = -1;

```

...

```

printf("A = (matlab base-1)\n");
printMatrix(m, n, A, lda, "A");
printf("=====\n");

/* step 1: create cusolverDn/cublas handle */
cusolver_status = cusolverDnCreate(&cusolverH);
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);

cublas_status = cublasCreate(&cublasH);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

/* step 2: copy A and B to device */
cudaStat1 = cudaMalloc ((void**) &d_A , sizeof(double)*lda*n);
cudaStat2 = cudaMalloc ((void**) &d_S , sizeof(double)*n);
cudaStat3 = cudaMalloc ((void**) &d_U , sizeof(double)*lda*m);
cudaStat4 = cudaMalloc ((void**) &d_VT , sizeof(double)*lda*n);
cudaStat5 = cudaMalloc ((void**) &d_info, sizeof(int));
cudaStat6 = cudaMalloc ((void**) &d_W , sizeof(double)*lda*n);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);
assert(cudaSuccess == cudaStat5);
assert(cudaSuccess == cudaStat6);

cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*n,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaDeviceSynchronize(); /* wait until d_A is ready */

/* step 3: query working space of SVD */
signed char jobu = 'A'; /* all m columns of U */
signed char jobvt = 'A'; /* all n columns of VT */
cusolver_status = cusolverDnGesvd_bufferSize (
    cusolverH,
    NULL, /* params */
    jobu,
    jobvt,
    (int64_t)m,
    (int64_t)n,
    CUDA_R_64F, /* dataTypeA */
    d_A,
    (int64_t)lda,
    CUDA_R_64F, /* dataTypeS */
    d_S,
    CUDA_R_64F, /* dataTypeU */
    d_U,
    (int64_t)lda, /* ldu */
    CUDA_R_64F, /* dataTypeVT */
    d_VT,
    (int64_t)lda, /* ldvt */
    CUDA_R_64F, /* computeType */
    &workspaceInBytes);

cudaStat1 = cudaMalloc((void**) &pBuffer , workspaceInBytes);
assert(cudaSuccess == cudaStat1);

```

...

```

/* step 4: compute SVD */
cusolver_status = cusolverDnGesvd(
    cusolverH,
    NULL, /* params */
    jobu,
    jobvt,
    (int64_t)m,
    (int64_t)n,
    CUDA_R_64F, /* dataTypeA */
    d_A,
    (int64_t)lda,
    CUDA_R_64F, /* dataTypeS */
    d_S,
    CUDA_R_64F, /* dataTypeU */
    d_U,
    (int64_t)lda, /* ldu */
    CUDA_R_64F, /* dataTypeVT */
    d_VT,
    (int64_t)lda, /* ldvt */
    CUDA_R_64F, /* computeType */
    pBuffer,
    workspaceInBytes,
    d_info);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSOLVER_STATUS_SUCCESS == cusolver_status);
assert(cudaSuccess == cudaStat1);

    cudaStat1 = cudaMemcpy(U , d_U , sizeof(double)*lda*m,
        cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(VT, d_VT, sizeof(double)*lda*n,
        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(S , d_S , sizeof(double)*n ,
        cudaMemcpyDeviceToHost);
    cudaStat4 = cudaMemcpy(&info_gpu, d_info, sizeof(int),
        cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    assert(cudaSuccess == cudaStat4);
    cudaDeviceSynchronize(); /* wait until host data is ready */

printf("after gesvd: info_gpu = %d\n", info_gpu);
assert(0 == info_gpu);
printf("=====\n");

printf("S = (matlab base-1)\n");
printMatrix(n, 1, S, lda, "S");
printf("=====\n");

printf("U = (matlab base-1)\n");
printMatrix(m, m, U, lda, "U");
printf("=====\n");

printf("VT = (matlab base-1)\n");
printMatrix(n, n, VT, lda, "VT");
printf("=====\n");

/* step 5: measure error of singular value */
double ds_sup = 0;
for(int j = 0; j < n; j++){
    double err = fabs( S[j] - S_exact[j] );
    ds_sup = (ds_sup > err)? ds_sup : err;
}
printf("|S - S_exact| = %E \n", ds_sup);

```

...

```

/* step 6: |A - U*S*VT| */
/* W = S*VT */
cublas_status = cublasDdggmm(
    cublasH,
    CUBLAS_SIDE_LEFT,
    n,
    n,
    d_VT,
    lda,
    d_S,
    1,
    d_W,
    lda);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);
/* A := -U*W + A */
cudaStat1 = cudaMemcpy(d_A, A, sizeof(double)*lda*n,
    cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaDeviceSynchronize(); /* wait until d_A is ready */
cublas_status = cublasDgemv_v2(
    cublasH,
    CUBLAS_OP_N,
    CUBLAS_OP_N,
    m, /* number of rows of A */
    n, /* number of columns of A */
    n, /* number of columns of U */
    &h_minus_one,
    d_U,
    lda,
    d_W,
    lda,
    &h_one,
    d_A,
    lda);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

double dR_fro = 0.0;
cublas_status = cublasDnrm2_v2(
    cublasH, lda*n, d_A, 1, &dR_fro);
assert(CUBLAS_STATUS_SUCCESS == cublas_status);

printf("|A - U*S*VT| = %E \n", dR_fro);

/* free resources */
if (d_A ) cudaFree(d_A);
if (d_S ) cudaFree(d_S);
if (d_U ) cudaFree(d_U);
if (d_VT ) cudaFree(d_VT);
if (d_info) cudaFree(d_info);
if (pBuffer ) cudaFree(pBuffer);
if (d_W ) cudaFree(d_W);

if (cublasH ) cublasDestroy(cublasH);
if (cusolverH) cusolverDnDestroy(cusolverH);

cudaDeviceReset();

return 0;
}

```

# Appendix H.

## EXAMPLES OF MULTIGPU EIGENVALUE SOLVER

This chapter provides three examples to perform multiGPU symmetric eigenvalue solver. The difference among them is how to generate the testing matrix. The testing matrix is a tridiagonal matrix, from standard 3-point stencil of Laplacian operator with Dirichlet boundary condition, so each row has (-1, 2, -1) signature.

The spectrum has analytic formula, we can check the accuracy of eigenvalues easily. The user can change the dimension of the matrix to measure the performance of eigenvalue solver.

The example code enables peer-to-peer access to take advantage of NVLINK. The user can check the performance by on/off peer-to-peer access.

The procedures of these three examples are 1) to prepare a tridiagonal matrix in distributed sense, 2) to query size of the workspace and to allocate the workspace for each device, 3) to compute eigenvalues and eigenvectors, and 4) to check accuracy of eigenvalues.

The example 1 allocates distributed matrix by calling `createMat`. It generates the matrix on host memory and copies it to distributed device memory via `memcpyH2D`.

The example 2 allocates distributed matrix manually, generates the matrix on host memory and copies it to distributed device memory manually. This example is for the users who are familiar with data layout of ScaLAPACK.

The example 3 allocates distributed matrix by calling `createMat` and generates the matrix element-by-element on distributed matrix via `memcpyH2D`. The user needs not to know the data layout of ScaLAPACK. It is useful when the matrix is sparse.

## H.1. SYEVD of 1D Laplacian operator (example 1)

...

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *     nvcc -ccbin gcc -I/usr/local/cuda/include -c main.cpp -o main.o
 *     nvcc -cudart static main.o -lcusolverMg
 */
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include "cusolverMg.h"
#include "util.hxx"

// #define SHOW_FORMAT

#ifndef IDX2F
#define IDX2F(i,j,lda) (((j)-1)*((size_t)lda))+((i)-1)
#endif /* IDX2F */

#ifndef IDX1F
#define IDX1F(i) ((i)-1)
#endif /* IDX1F */

static void print_matrix(
    int m,
    int n,
    const double *A,
    int lda,
    const char* name)
{
    printf("%s = matlab base-1, %d-by-%d matrix\n", name, m, n);
    for(int row = 1 ; row <= m ; row++){
        for(int col = 1 ; col <= n ; col++){
            double Aij = A[IDX2F(row, col, lda)];
            printf("%s(%d,%d) = %20.16E\n", name, row, col, Aij );
        }
    }
}

static void gen_1d_laplacian(
    int N,
    double *A,
    int lda)
{
    memset(A, 0, sizeof(double)*lda*N);
    for(int J = 1 ; J <= N; J++){
        /* A(J,J) = 2 */
        A[ IDX2F( J, J, lda ) ] = 2.0;
        if ( (J-1) >= 1 ){
            /* A(J, J-1) = -1 */
            A[ IDX2F( J, J-1, lda ) ] = -1.0;
        }
        if ( (J+1) <= N ){
            /* A(J, J+1) = -1 */
            A[ IDX2F( J, J+1, lda ) ] = -1.0;
        }
    }
}

```

...

```

int main( int argc, char* argv[])
{
    cusolverMgHandle_t handle = NULL;
    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat = cudaSuccess;

    /* maximum number of GPUs */
    const int MAX_NUM_DEVICES = 16;

    int nbGpus = 0;
    int deviceList[MAX_NUM_DEVICES];

    const int N    = 2111;
    const int IA   = 1;
    const int JA   = 1;
    const int T_A  = 256; /* tile size */
    const int lda  = N;

    double *A = NULL; /* A is N-by-N */
    double *D = NULL; /* D is 1-by-N */
    int info = 0;

    cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR;

    cudaLibMgMatrixDesc_t descrA;
    cudaLibMgGrid_t gridA;
    cusolverMgGridMapping_t mapping = CUDALIBMG_GRID_MAPPING_COL_MAJOR;

    double **array_d_A = NULL;

    int64_t lwork = 0; /* workspace: number of elements per device */
    double **array_d_work = NULL;

    printf("test 1D Laplacian of order %d\n", N);

    printf("step 1: create Mg handle and select devices \n");
    status = cusolverMgCreate(&handle);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    cudaStat = cudaGetDeviceCount( &nbGpus );
    assert( cudaSuccess == cudaStat );

    nbGpus = (nbGpus < MAX_NUM_DEVICES)? nbGpus : MAX_NUM_DEVICES;
    printf("\tthere are %d GPUs \n", nbGpus);
    for(int j = 0 ; j < nbGpus ; j++){
        deviceList[j] = j;
        cudaDeviceProp prop;
        cudaGetDeviceProperties(&prop, j);
        printf("\tdevice %d, %s, cc %d.%d \n",j, prop.name, prop.major,
prop.minor);
    }

    status = cusolverMgDeviceSelect(
        handle,
        nbGpus,
        deviceList);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    printf("step 2: Enable peer access.\n");
    assert( 0 == enablePeerAccess( nbGpus, deviceList ) );

```

...

```

printf("step 3: allocate host memory A \n");
A = (double *)malloc (sizeof(double)*lda*N);
D = (double *)malloc (sizeof(double)*N);
assert( NULL != A );
assert( NULL != D );

printf("step 4: prepare 1D Laplacian \n");
gen_ld_laplacian(
    N,
    &A[ IDX2F( IA, JA, lda ) ],
    lda
);

#ifdef SHOW_FORMAT
    print_matrix( N, N, A, lda, "A");
#endif

printf("step 5: create matrix descriptors for A and D \n");

status = cusolverMgCreateDeviceGrid(&gridA, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
/* (global) A is N-by-N */
status = cusolverMgCreateMatrixDesc(
    &descrA,
    N, /* nubmer of rows of (global) A */
    N, /* number of columns of (global) A */
    N, /* number or rows in a tile */
    T_A, /* number of columns in a tile */
    CUDA_R_64F,
    gridA );
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 6: allocate distributed matrices A and D \n");

array_d_A = (double**)malloc(sizeof(double*)*nbGpus);
assert(NULL != array_d_A);
/* A := 0 */
createMat<double>(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A
);

printf("step 7: prepare data on devices \n");
memcpyH2D<double>(
    nbGpus,
    deviceList,
    N,
    N,
/* input */
    A,
    lda,
/* output */
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A, /* host pointer array of dimension nbGpus */
    IA,
    JA
);

```

...

```

printf("step 8: allocate workspace space \n");
status = cusolverMgSyevd_bufferSize(
    handle,
    (cusolverEigMode_t)jobz,
    CUBLAS_FILL_MODE_LOWER, /* only support lower mode */
    N,
    (void**)array_d_A,
    IA, /* base-1 */
    JA, /* base-1 */
    descrA,
    (void*)D,
    CUDA_R_64F,
    CUDA_R_64F,
    &lwork);
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("\tallocate device workspace, lwork = %lld \n", (long long)lwork);
array_d_work = (double**)malloc(sizeof(double)*nbGpus);
assert( NULL != array_d_work);
/* array_d_work[j] points to device workspace of device j */
workspaceAlloc(
    nbGpus,
    deviceList,
    sizeof(double)*lwork, /* number of bytes per device */
    (void**)array_d_work
);

/* sync all devices */
cudaStat = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat);

printf("step 9: compute eigenvalues and eigenvectors \n");
status = cusolverMgSyevd(
    handle,
    (cusolverEigMode_t)jobz,
    CUBLAS_FILL_MODE_LOWER, /* only support lower mode */
    N,
    (void**)array_d_A, /* exit: eigenvectors */
    IA,
    JA,
    descrA,
    (void**)D, /* exit: eigenvalues */
    CUDA_R_64F,
    CUDA_R_64F,
    (void**)array_d_work,
    lwork,
    &info /* host */
);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* sync all devices */
cudaStat = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat);

/* check if SYEVD converges */
assert(0 == info);

```

...

```

printf("step 10: copy eigenvectors to A and eigenvalues to D\n");

memcpyD2H<double>(
    nbGpus,
    deviceList,
    N,
    N,
/* input */
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A,
    IA,
    JA,
/* output */
    A, /* N-y-N eigenvectors */
    lda
);

#ifdef SHOW_FORMAT
printf("eigenvalue D = \n");
/* D is 1-by-N */
print_matrix(1, N, D, 1, "D");
#endif

printf("step 11: verify eigenvalues \n");
printf("    lambda(k) = 4 * sin(pi/2 * k/(N+1))^2 for k = 1:N \n");
double max_err_D = 0;
for(int k = 1; k <= N ; k++){
    const double pi = 4*atan(1.0);
    const double h = 1.0/((double)N+1);
    const double factor = sin(pi/2.0 * ((double)k)*h);
    const double lambda = 4.0*factor*factor;
    const double err = fabs(D[IDX1F(k)] - lambda);
    max_err_D = (max_err_D > err)? max_err_D : err;
//    printf("k = %d, D = %E, lambda = %E, err = %E\n", k, D[IDX1F(k)],
lambda, err);
}
printf("\n|D - lambda|_inf = %E\n\n", max_err_D);

printf("step 12: free resources \n");
destroyMat(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    (void**)array_d_A );

workspaceFree( nbGpus, deviceList, (void**)array_d_work );

if (NULL != A) free(A);
if (NULL != D) free(D);

if (NULL != array_d_A ) free(array_d_A);
if (NULL != array_d_work) free(array_d_work);

return 0;
}

```

...



```

...
/* util.hxx */
/*
 * nbGpus : (int) number of gpus in deviceList array.
 * deviceList : (*int) list of device ids.
 *
 * The function restores the input device before leaving.
 */
static int enablePeerAccess (const int nbGpus, const int *deviceList )
{
    int currentDevice = 0;
    cudaGetDevice( &currentDevice );

    /* Remark: access granted by this cudaDeviceEnablePeerAccess is
    unidirectional */
    /* Rows and columns represents a connectivity matrix between GPUs in the
    system */
    for(int row=0; row < nbGpus; row++) {
        cudaSetDevice(row);
        for(int col=0; col < nbGpus; col++) {
            if( row != col ){
                cudaError_t cudaStat1 = cudaSuccess;
                cudaError_t cudaStat2 = cudaSuccess;
                int canAccessPeer = 0;
                cudaStat1 = cudaDeviceCanAccessPeer( &canAccessPeer, row, col );
                if ( canAccessPeer ){
                    printf("\t Enable peer access from gpu %d to gpu %d\n", row,
col );
                        cudaStat2 = cudaDeviceEnablePeerAccess( col, 0 );
                    }
                assert(cudaStat1 == cudaSuccess);
                assert(cudaStat2 == cudaSuccess);
            }
        }
        cudaSetDevice( currentDevice );
        return 0;
    }

static int workspaceFree(
    int num_devices,
    const int *deviceIdA, /* <int> dimension num_devices */
    void **array_d_work /* <t> num_devices, host array */
    j */
    )
{
    int currentDev = 0; /* record current device ID */
    cudaGetDevice( &currentDev );

    for(int idx = 0 ; idx < num_devices ; idx++){
        int deviceId = deviceIdA[idx];
        /* WARNING: we need to set device before any runtime API */
        cudaSetDevice( deviceId );

        if (NULL != array_d_work[idx]){
            cudaFree(array_d_work[idx]);
        }
    }
    cudaSetDevice( currentDev );
    return 0;
}

```

...

```

static int workspaceAlloc(
    int num_devices,
    const int *deviceIdA, /* <int> dimension num_devices */
    size_t sizeInBytes, /* number of bytes per device */
    void **array_d_work /* <t> num_devices, host array */
    /* array_d_work[j] points to device workspace of device
    j */
)
{
    cudaError_t cudaStat1 = cudaSuccess;

    int currentDev = 0; /* record current device ID */
    cudaGetDevice( &currentDev );

    memset(array_d_work, 0, sizeof(void*)*num_devices);
    for(int idx = 0 ; idx < num_devices ; idx++){
        int deviceId = deviceIdA[idx];
/* WARNING: we need to set device before any runtime API */
        cudaSetDevice( deviceId );

        void *d_workspace = NULL;

        cudaStat1 = cudaMalloc(&d_workspace, sizeInBytes);
        assert( cudaSuccess == cudaStat1 );
        array_d_work[idx] = d_workspace;
    }
    cudaSetDevice(currentDev);
    return 0;
}

/* create a empty matrix A with A := 0 */
template <typename T_ELEM>
int createMat(
    int num_devices,
    const int *deviceIdA, /* <int> dimension num_devices */
    int N_A, /* number of columns of global A */
    int T_A, /* number of columns per column tile */
    int LLD_A, /* leading dimension of local A */
    T_ELEM **array_d_A /* host pointer array of dimension num_devices */
)
{
    cudaError_t cudaStat1 = cudaSuccess;
    int currentDev = 0; /* record current device id */
    cudaGetDevice( &currentDev );
    cudaDeviceSynchronize();
    const int A_num_blks = ( N_A + T_A - 1 ) / T_A;
    const int max_A_num_blks_per_device = (A_num_blks + num_devices-1)/
num_devices;
/* Allocate base pointers */
    memset(array_d_A, 0, sizeof(T_ELEM*) * num_devices);
    for( int p = 0 ; p < num_devices ; p++){
        cudaStat1 = cudaSetDevice(deviceIdA[p]);
        assert(cudaSuccess == cudaStat1);
/* Allocate max_A_num_blks_per_device blocks per device */
        cudaStat1 =
        cudaMalloc( &(array_d_A[p]), sizeof(T_ELEM)*LLD_A*T_A*max_A_num_blks_per_device );
        assert(cudaSuccess == cudaStat1);
/* A := 0 */
        cudaStat1 = cudaMemset( array_d_A[p],
0, sizeof(T_ELEM)*LLD_A*T_A*max_A_num_blks_per_device );
        assert(cudaSuccess == cudaStat1);
    }
    cudaDeviceSynchronize();
    cudaSetDevice(currentDev);
    return 0;
}

```

...

```

static int destroyMat (
    int num_devices,
    const int *deviceIdA, /* <int> dimension num devices */
    int N_A, /* number of columns of global A */
    int T_A, /* number of columns per column tile */
    void **array_d_A) /* host pointer array of dimension num_devices */
{
    cudaError_t cudaStat = cudaSuccess;

    int currentDev = 0; /* record current device id */
    cudaGetDevice( &currentDev );

    const int num_blks = (N_A + T_A - 1) / T_A;
    for( int p = 0 ; p < num_devices ; p++){
        cudaStat = cudaSetDevice(deviceIdA[p]);
        assert(cudaSuccess == cudaStat);

        if ( NULL != array_d_A[p] ){
            cudaFree( array_d_A[p] );
        }
    }
    memset(array_d_A, 0, sizeof(void*)*num_devices);
    cudaSetDevice(currentDev);
    return 0;
}

template <typename T_ELEM>
static int mat_pack2unpack(
    int num_devices,
    int N_A, /* number of columns of global A */
    int T_A, /* number of columns per column tile */
    int LLD_A, /* leading dimension of local A */
    T_ELEM **array_d_A_packed, /* host pointer array of dimension num_devices
    */
    /* output */
    T_ELEM **array_d_A_unpacked /* host pointer array of dimension num_blks */
)
{
    const int num_blks = ( N_A + T_A - 1) / T_A;

    for(int p_a = 0 ; p_a < num_devices ; p_a++){
        T_ELEM *d_A = array_d_A_packed[p_a];
        int nz_blks = 0;
        for(int JA_blk_id = p_a ; JA_blk_id < num_blks ; JA_blk_id+=num_devices)
        {
            array_d_A_unpacked[JA_blk_id] = d_A + (size_t)LLD_A * T_A *
nz_blks ;
            nz_blks++;
        }
    }
    return 0;
}

```

...

```

/*
 * A(IA:IA+M-1, JA:JA+N-1) := B(1:M, 1:N)
 */
template <typename T_ELEM>
static int memcpyH2D(
    int num_devices,
    const int *deviceIdA, /* <int> dimension num_devices */
    int M, /* number of rows in local A, B */
    int N, /* number of columns in local A, B */
    /* input */
    const T_ELEM *h_B, /* host array, h_B is M-by-N with leading dimension ldb
    */
    int ldb,
    /* output */
    int N_A, /* number of columns of global A */
    int T_A, /* number of columns per column tile */
    int LLD_A, /* leading dimension of local A */
    T_ELEM **array_d_A_packed, /* host pointer array of dimension num_devices */
    int IA, /* base-1 */
    int JA /* base-1 */
    )
{
    cudaError_t cudaStat1 = cudaSuccess;

    int currentDev = 0; /* record current device id */

    /* Quick return if possible */
    if ( (0 >= M) || (0 >= N) ){
        return 0;
    }

    /* consistent checking */
    if ( ldb < M ){
        return 1;
    }

    cudaGetDevice( &currentDev );
    cudaDeviceSynchronize();

    const int num_blks = ( N_A + T_A - 1 ) / T_A;

    T_ELEM **array_d_A_unpacked = (T_ELEM**)malloc(sizeof(T_ELEM*)*num_blks);
    assert(NULL != array_d_A_unpacked);

    mat_pack2unpack<T_ELEM>(
        num_devices,
        N_A, /* number of columns of global A */
        T_A, /* number of columns per column tile */
        LLD_A, /* leading dimension of local A */
        array_d_A_packed, /* host pointer array of size num_devices */
    /* output */
        array_d_A_unpacked /* host pointer array of size num_blks */
    );

    /* region of interest is A(IA:IA+N-1, JA:JA+N-1) */
    const int N_hat = (JA-1) + N; /* JA is base-1 */

    const int JA_start_blk_id = (JA-1)/T_A;
    const int JA_end_blk_id = (N_hat-1)/T_A;

```

...

```

    for(int p_a = 0 ; p_a < num_devices ; p_a++){
/* region of interest: JA_start_blk_id:1:JA_end_blk_id */
        for(int JA_blk_id = p_a; JA_blk_id <= JA_end_blk_id ; JA_blk_id
+=num_devices){
            if ( JA_blk_id < JA_start_blk_id ) { continue; }
/*
 * process column block of A
 * A(A_start_row:M_A, A_start_col : (A_start_col + IT_A-1) )
 */
            const int IBX_A = (1 + JA_blk_id*T_A); /* base-1 */
            const int A_start_col = imax( JA, IBX_A ); /* base-1 */
            const int A_start_row = IA; /* base-1 */

            const int bdd = imin( N_hat, (IBX_A + T_A -1) );
            const int IT_A = imin( T_A, (bdd - A_start_col + 1) );

            const int loc_A_start_row = A_start_row; /* base-1 */
            const int loc_A_start_col = (A_start_col-IBX_A)+1; /* base-1 */

            T_ELEM *d_A = array_d_A_unpacked[JA_blk_id] +
            IDX2F( loc_A_start_row, loc_A_start_col, LLD_A );
            const T_ELEM *h_A = h_B + IDX2F( A_start_row - IA + 1, A_start_col -
            JA + 1, ldb );

            cudaStat1 = cudaMemcpy2D(
                d_A, /* dst */
                (size_t)LLD_A * sizeof(T_ELEM),
                h_A, /* src */
                (size_t)ldb * sizeof(T_ELEM),
                (size_t)M * sizeof(T_ELEM),
                (size_t)IT_A,
                cudaMemcpyHostToDevice
            );
            assert( cudaSuccess == cudaStat1 );
        } /* for each tile per device */
    } /* for each device */
    cudaDeviceSynchronize();
    cudaSetDevice(currentDev);

    if ( NULL != array_d_A_unpacked ) { free(array_d_A_unpacked); }
    return 0;
}

/*
 * B(1:M, 1:N) := A(IA:IA+M-1, JA:JA+N-1)
 */
template <typename T_ELEM>
static int memcpyD2H(
    int num_devices,
    const int *deviceIdA, /* <int> dimension num_devices */
    int M, /* number of rows in local A, B */
    int N, /* number of columns in local A, B */
    /* input */
    int N_A, /* number of columns of global A */
    int T_A, /* number of columns per column tile */
    int LLD_A, /* leading dimension of local A */
    T_ELEM **array_d_A_packed, /* host pointer array of dimension num_devices */
    int IA, /* base-1 */
    int JA, /* base-1 */
    /* output */
    T_ELEM *h_B, /* host array, h_B is M-by-N with leading dimension ldb */
    int ldb
)
{

```

...

```

    cudaError_t cudaStat1 = cudaSuccess;
    int currentDev = 0; /* record current device id */

/* Quick return if possible */
    if ( (0 >= M) || (0 >= N) ){
        return 0;
    }
/* consistent checking */
    if ( ldb < M ){
        return 1;
    }
    cudaGetDevice( &currentDev );
    cudaDeviceSynchronize();

    const int num_blks = ( N_A + T_A - 1 ) / T_A;
    T_ELEM **array_d_A_unpacked = (T_ELEM**)malloc(sizeof(T_ELEM)*num_blks);
    assert(NULL != array_d_A_unpacked);

    mat_pack2unpack<T_ELEM>(
        num_devices,
        N_A, /* number of columns of global A */
        T_A, /* number of columns per column tile */
        LLD_A, /* leading dimension of local A */
        array_d_A_packed, /* host pointer array of size num_devices */
        array_d_A_unpacked /* host pointer array of size num_blks */
    );
/* region of interest is A(IA:IA+N-1, JA:JA+N-1) */
    const int N_hat = (JA-1) + N; /* JA is base-1 */
    const int JA_start_blk_id = (JA-1)/T_A;
    const int JA_end_blk_id = (N_hat-1)/T_A;
    for(int p_a = 0 ; p_a < num_devices ; p_a++){
/* region of interest: JA_start_blk_id:1:JA_end_blk_id */
        for(int JA_blk_id = p_a; JA_blk_id <= JA_end_blk_id ; JA_blk_id
+=num_devices){
            if ( JA_blk_id < JA_start_blk_id ) { continue; }
/* process column block, A(A_start_row:M_A, A_start_col : (A_start_col +
IT_A-1) ) */
            const int IBX_A = (1 + JA_blk_id*T_A); /* base-1 */
            const int A_start_col = imax( JA, IBX_A ); /* base-1 */
            const int A_start_row = IA; /* base-1 */
            const int bdd = imin( N_hat, (IBX_A + T_A -1) );
            const int IT_A = imin( T_A, (bdd - A_start_col + 1) );
            const int loc_A_start_row = A_start_row; /* base-1 */
            const int loc_A_start_col = (A_start_col-IBX_A)+1; /* base-1 */
            const T_ELEM *d_A = array_d_A_unpacked[JA_blk_id] +
IDX2F( loc_A_start_row, loc_A_start_col, LLD_A );
            T_ELEM *h_A = h_B + IDX2F( A_start_row - IA + 1, A_start_col - JA +
1, ldb );
            cudaStat1 = cudaMemcpy2D(
                h_A, /* dst */
                (size_t)ldb * sizeof(T_ELEM),
                d_A, /* src */
                (size_t)LLD_A * sizeof(T_ELEM),
                (size_t)M * sizeof(T_ELEM),
                (size_t)IT_A,
                cudaMemcpyDeviceToHost
            );
            assert( cudaSuccess == cudaStat1 );
        } /* for each tile per device */
    } /* for each device */
    cudaDeviceSynchronize();
    cudaSetDevice(currentDev);
    if ( NULL != array_d_A_unpacked ) { free(array_d_A_unpacked); }
    return 0;
}

```

## H.2. SYEVD of 1D Laplacian operator (example 2)

...

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *      nvcc -ccbin gcc -I/usr/local/cuda/include -c main.cpp -o main.o
 *      nvcc -cudart static main.o -lcusolverMg
 */
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include "cusolverMg.h"
#include "util.hxx"

// #define SHOW_FORMAT

#ifndef IDX2F
#define IDX2F(i,j,lda) (((j)-1)*((size_t)lda))+((i)-1)
#endif /* IDX2F */

#ifndef IDX1F
#define IDX1F(i) ((i)-1)
#endif /* IDX1F */

#define imin(x,y) ((x) < (y)) ? (x) : (y)

static void print_matrix(
    int m,
    int n,
    const double *A,
    int lda,
    const char* name)
{
    printf("%s = matlab base-1, %d-by-%d matrix\n", name, m, n);
    for(int row = 1 ; row <= m ; row++){
        for(int col = 1 ; col <= n ; col++){
            double Aij = A[IDX2F(row, col, lda)];
            printf("%s(%d,%d) = %20.16E\n", name, row, col, Aij );
        }
    }
}

static void gen_1d_laplacian(
    int N,
    double *A,
    int lda)
{
    memset(A, 0, sizeof(double)*lda*N);
    for(int J = 1 ; J <= N; J++){
        /* A(J,J) = 2 */
        A[ IDX2F( J, J, lda ) ] = 2.0;
        if ( (J-1) >= 1 ){
            /* A(J, J-1) = -1 */
            A[ IDX2F( J, J-1, lda ) ] = -1.0;
        }
        if ( (J+1) <= N ){
            /* A(J, J+1) = -1 */
            A[ IDX2F( J, J+1, lda ) ] = -1.0;
        }
    }
}

```

...

```

int main( int argc, char* argv[])
{
    cusolverMgHandle_t handle = NULL;
    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat = cudaSuccess;
    /* maximum number of GPUs */
    const int MAX_NUM_DEVICES = 16;

    int nbGpus = 0;
    int deviceList[MAX_NUM_DEVICES];

    const int N = 2111;
    const int IA = 1;
    const int JA = 1;
    const int T_A = 256; /* tile size */
    const int lda = N;

    double *A = NULL; /* A is N-by-N */
    double *D = NULL; /* D is 1-by-N */
    int info = 0;

    cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR;

    cudaLibMgMatrixDesc_t descrA;
    cudaLibMgGrid_t gridA;
    cusolverMgGridMapping_t mapping = CUDALIBMG_GRID_MAPPING_COL_MAJOR;

    double **array_d_A = NULL;

    int64_t lwork = 0 ; /* workspace: number of elements per device */
    double **array_d_work = NULL;

    printf("test 1D Laplacian of order %d\n", N);

    printf("step 1: create Mg handle and select devices \n");
    status = cusolverMgCreate(&handle);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    cudaStat = cudaGetDeviceCount( &nbGpus );
    assert( cudaSuccess == cudaStat );

    nbGpus = (nbGpus < MAX_NUM_DEVICES)? nbGpus : MAX_NUM_DEVICES;
    printf("\tthere are %d GPUs \n", nbGpus);
    for(int j = 0 ; j < nbGpus ; j++){
        deviceList[j] = j;
        cudaDeviceProp prop;
        cudaGetDeviceProperties(&prop, j);
        printf("\tdevice %d, %s, cc %d.%d \n",j, prop.name, prop.major,
prop.minor);
    }

    status = cusolverMgDeviceSelect(
        handle,
        nbGpus,
        deviceList);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    printf("step 2: Enable peer access \n");
    assert( 0 == enablePeerAccess( nbGpus, deviceList ) );

```

...

```

printf("step 3: allocate host memory A \n");
A = (double *)malloc (sizeof(double)*lda*N);
D = (double *)malloc (sizeof(double)*N);
assert( NULL != A );
assert( NULL != D );

printf("step 4: prepare 1D Laplacian \n");
gen_ld_laplacian(
    N,
    &A[ IDX2F( IA, JA, lda ) ],
    lda
);

#ifdef SHOW_FORMAT
    print_matrix( N, N, A, lda, "A");
#endif

printf("step 5: create matrix descriptors for A and D \n");

status = cusolverMgCreateDeviceGrid(&gridA, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
/* (global) A is N-by-N */
status = cusolverMgCreateMatrixDesc(
    &descrA,
    N, /* nubmer of rows of (global) A */
    N, /* number of columns of (global) A */
    N, /* number or rows in a tile */
    T_A, /* number of columns in a tile */
    CUDA_R_64F,
    gridA );
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 6: allocate distributed matrices A and D \n");

array_d_A = (double**) malloc (sizeof(double*) * nbGpus );
assert( NULL != array_d_A );

const int A_num_blks = ( N + T_A - 1 ) / T_A;
const int blks_per_device = (A_num_blks + nbGpus-1)/nbGpus;

for( int p = 0 ; p < nbGpus ; p++){
    cudaSetDevice(deviceList[p]);
    cudaStat =
    cudaMalloc( &(array_d_A[p]), sizeof(double)*lda*T_A*blks_per_device );
    assert(cudaSuccess == cudaStat);
}

printf("step 7: prepare data on devices \n");
/* The following setting only works for IA = JA = 1 */
for( int k = 0 ; k < A_num_blks ; k++){
/* k = ibx * nbGpus + p */
    const int p = (k % nbGpus);
    const int ibx = (k / nbGpus);
    double *h_Ak = A + (size_t)lda*T_A*k;
    double *d_Ak = array_d_A[p] + (size_t)lda*T_A*ibx;
    const int width = imin( T_A, (N - T_A*k) );
    cudaStat = cudaMemcpy(d_Ak, h_Ak, sizeof(double)*lda*width,
    cudaMemcpyHostToDevice);
    assert(cudaSuccess == cudaStat);
}
/* sync all devices */
cudaStat = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat);

```

...

```

printf("step 8: allocate workspace space \n");
status = cusolverMgSyevd_bufferSize(
    handle,
    (cusolverEigMode_t)jobz,
    CUBLAS_FILL_MODE_LOWER, /* only support lower mode */
    N,
    (void**)array_d_A,
    IA, /* base-1 */
    JA, /* base-1 */
    descrA,
    (void*)D,
    CUDA_R_64F,
    CUDA_R_64F,
    &lwork);
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("\tallocate device workspace, lwork = %lld \n", (long long)lwork);
array_d_work = (double**)malloc(sizeof(double)*nbGpus);
assert(NULL != array_d_work);
/* array_d_work[j] points to device workspace of device j */
workspaceAlloc(
    nbGpus,
    deviceList,
    sizeof(double)*lwork, /* number of bytes per device */
    (void**)array_d_work
);

/* sync all devices */
cudaStat = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat);

printf("step 9: compute eigenvalues and eigenvectors \n");
status = cusolverMgSyevd(
    handle,
    (cusolverEigMode_t)jobz,
    CUBLAS_FILL_MODE_LOWER, /* only support lower mode */
    N,
    (void**)array_d_A, /* exit: eigenvectors */
    IA,
    JA,
    descrA,
    (void*)D, /* exit: eigenvalues */
    CUDA_R_64F,
    CUDA_R_64F,
    (void**)array_d_work,
    lwork,
    &info /* host */
);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* sync all devices */
cudaStat = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat);

/* check if SYEVD converges */
assert(0 == info);

```

...

```

printf("step 10: copy eigenvectors to A and eigenvalues to D\n");

memcpyD2H<double>(
    nbGpus,
    deviceList,
    N,
    N,
/* input */
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A,
    IA,
    JA,
/* output */
    A, /* N-y-N eigenvectors */
    lda
);

#ifdef SHOW_FORMAT
printf("eigenvalue D = \n");
/* D is 1-by-N */
print_matrix(1, N, D, 1, "D");
#endif

printf("step 11: verify eigenvalues \n");
printf("    lambda(k) = 4 * sin(pi/2 * k/(N+1))^2 for k = 1:N \n");
double max_err_D = 0;
for(int k = 1; k <= N ; k++){
    const double pi = 4*atan(1.0);
    const double h = 1.0/((double)N+1);
    const double factor = sin(pi/2.0 * ((double)k)*h);
    const double lambda = 4.0*factor*factor;
    const double err = fabs(D[IDX1F(k)] - lambda);
    max_err_D = (max_err_D > err)? max_err_D : err;
//    printf("k = %d, D = %E, lambda = %E, err = %E\n", k, D[IDX1F(k)],
lambda, err);
}
printf("\n|D - lambda|_inf = %E\n\n", max_err_D);

printf("step 12: free resources \n");
destroyMat(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    (void**)array_d_A );

workspaceFree( nbGpus, deviceList, (void**)array_d_work );

if (NULL != A) free(A);
if (NULL != D) free(D);

if (NULL != array_d_A ) free(array_d_A);
if (NULL != array_d_work) free(array_d_work);

return 0;
}

```

## H.3. SYEVD of 1D Laplacian operator (example 3)

...

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *      nvcc -ccbin gcc -I/usr/local/cuda/include -c main.cpp -o main.o
 *      nvcc -cudart static main.o -lcusolverMg
 */
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include "cusolverMg.h"
#include "util.hxx"

// #define SHOW_FORMAT

#ifndef IDX2F
#define IDX2F(i,j,lda) (((j)-1)*((size_t)lda))+((i)-1)
#endif /* IDX2F */
#ifndef IDX1F
#define IDX1F(i) ((i)-1)
#endif /* IDX1F */

static void print_matrix(
    int m,
    int n,
    const double *A,
    int lda,
    const char* name)
{
    printf("%s = matlab base-1, %d-by-%d matrix\n", name, m, n);
    for(int row = 1 ; row <= m ; row++){
        for(int col = 1 ; col <= n ; col++){
            double Aij = A[IDX2F(row, col, lda)];
            printf("%s(%d,%d) = %20.16E\n", name, row, col, Aij );
        }
    }
}

/* the caller must set A = 0 */
static void gen_1d_laplacian(
    cusolverMgHandle_t handle,
    int nbGpus,
    const int *deviceList,
    int N, /* number of columns of global A */
    int T_A, /* number of columns per column tile */
    int LLD_A, /* leading dimension of local A */
    double **array_d_A /* host pointer array of dimension nbGpus */
)
{
    double two = 2.0;
    double minus_one = -1.0;
    for(int J = 1 ; J <= N; J++ ){
        /* A(J,J) = 2 */
        memcpyH2D<double>(nbGpus, deviceList, 1, 1, &two, 1, N, T_A, LLD_A,
array_d_A, J, J);
        if ( (J-1) >= 1 ){
            /* A(J, J-1) = -1 */
            memcpyH2D<double>(nbGpus, deviceList, 1, 1, &minus_one, 1, N, T_A,
LLD_A, array_d_A, J, J-1);
        }
        if ( (J+1) <= N ){
            /* A(J, J+1) = -1 */
            memcpyH2D<double>(nbGpus, deviceList, 1, 1, &minus_one, 1, N, T_A,
LLD_A, array_d_A, J, J+1);
        }
    }
}

```

...

```

int main( int argc, char* argv[])
{
    cusolverMgHandle_t handle = NULL;
    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat = cudaSuccess;

    /* maximum number of GPUs */
    const int MAX_NUM_DEVICES = 16;

    int nbGpus = 0;
    int deviceList[MAX_NUM_DEVICES];

    const int N    = 2111;
    const int IA   = 1;
    const int JA   = 1;
    const int T_A  = 256; /* tile size */
    const int lda  = N;

    double *A = NULL; /* A is N-by-N */
    double *D = NULL; /* D is 1-by-N */
    int info = 0;

    cusolverEigMode_t jobz = CUSOLVER_EIG_MODE_VECTOR;

    cudaLibMgMatrixDesc_t descrA;
    cudaLibMgGrid_t gridA;
    cusolverMgGridMapping_t mapping = CUDALIBMG_GRID_MAPPING_COL_MAJOR;

    double **array_d_A = NULL;

    int64_t lwork = 0; /* workspace: number of elements per device */
    double **array_d_work = NULL;

    printf("test 1D Laplacian of order %d\n", N);

    printf("step 1: create Mg handle and select devices \n");
    status = cusolverMgCreate(&handle);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    cudaStat = cudaGetDeviceCount( &nbGpus );
    assert( cudaSuccess == cudaStat );

    nbGpus = (nbGpus < MAX_NUM_DEVICES)? nbGpus : MAX_NUM_DEVICES;
    printf("\tthere are %d GPUs \n", nbGpus);
    for(int j = 0 ; j < nbGpus ; j++){
        deviceList[j] = j;
        cudaDeviceProp prop;
        cudaGetDeviceProperties(&prop, j);
        printf("\tdevice %d, %s, cc %d.%d \n",j, prop.name, prop.major,
prop.minor);
    }

    status = cusolverMgDeviceSelect(
        handle,
        nbGpus,
        deviceList);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    printf("step 2: Enable peer access.\n");
    assert( 0 == enablePeerAccess( nbGpus, deviceList ) );

```

...

```

printf("step 3: allocate host memory A \n");
A = (double *)malloc (sizeof(double)*lda*N);
D = (double *)malloc (sizeof(double)*N);
assert( NULL != A );
assert( NULL != D );

printf("step 4: create matrix descriptors for A and D \n");
status = cusolverMgCreateDeviceGrid(&gridA, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
/* (global) A is N-by-N */
status = cusolverMgCreateMatrixDesc(
&descrA,
N, /* nubmer of rows of (global) A */
N, /* number of columns of (global) A */
N, /* number or rows in a tile */
T_A, /* number of columns in a tile */
CUDA_R_64F,
gridA );
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 5: allocate distributed matrices A and D, A = 0 and D = 0 \n");

array_d_A = (double**)malloc(sizeof(double*)*nbGpus);
assert(NULL != array_d_A);

/* A := 0 */
createMat<double>(
nbGpus,
deviceList,
N, /* number of columns of global A */
T_A, /* number of columns per column tile */
lda, /* leading dimension of local A */
array_d_A );

printf("step 6: prepare 1D Laplacian on devices \n");
gen_1d_laplacian(
handle,
nbGpus,
deviceList,
N, /* number of columns of global A */
T_A, /* number of columns per column tile */
lda, /* leading dimension of local A */
array_d_A);

printf("step 7: allocate workspace space \n");
status = cusolverMgSyevd_bufferSize(
handle,
(cusolverEigMode_t)jobz,
CUBLAS_FILL_MODE_LOWER, /* only support lower mode */
N,
(void**)array_d_A,
IA, /* base-1 */
JA, /* base-1 */
descrA,
(void*)D,
CUDA_R_64F,
CUDA_R_64F,
&lwork);
assert(CUSOLVER_STATUS_SUCCESS == status);

```

...

```

printf("\tallocate device workspace, lwork = %lld \n", (long long)lwork);
array_d_work = (double**)malloc(sizeof(double*)*nbGpus);
assert( NULL != array_d_work);
/* array_d_work[j] points to device workspace of device j */
workspaceAlloc(
    nbGpus,
    deviceList,
    sizeof(double)*lwork, /* number of bytes per device */
    (void**)array_d_work
);

/* sync all devices */
cudaStat = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat);

printf("step 8: compute eigenvalues and eigenvectors \n");
status = cusolverMgSyevd(
    handle,
    (cusolverEigMode_t)jobz,
    CUBLAS_FILL_MODE_LOWER, /* only support lower mode */
    N,
    (void**)array_d_A, /* exit: eigenvectors */
    IA,
    JA,
    descrA,
    (void*)D, /* exit: eigenvalues */
    CUDA_R_64F,
    CUDA_R_64F,
    (void**)array_d_work,
    lwork,
    &info /* host */
);
assert(CUSOLVER_STATUS_SUCCESS == status);

/* sync all devices */
cudaStat = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat);

/* check if SYEVD converges */
assert(0 == info);

printf("step 9: copy eigenvectors to A and eigenvalues to D\n");
memcpyD2H<double>(
    nbGpus,
    deviceList,
    N,
    N,
/* input */
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A,
    IA,
    JA,
/* output */
    A, /* N-y-N eigenvectors */
    lda
);
#ifdef SHOW_FORMAT
printf("eigenvalue D = \n");
/* D is 1-by-N */
print_matrix(1, N, D, 1, "D");
#endif

```

...

```

printf("step 10: verify eigenvalues \n");
printf("    lambda(k) = 4 * sin(pi/2 *k/(N+1))^2 for k = 1:N \n");
double max_err_D = 0;
for(int k = 1; k <= N ; k++){
    const double pi = 4*atan(1.0);
    const double h = 1.0/((double)N+1);
    const double factor = sin(pi/2.0 * ((double)k)*h);
    const double lambda = 4.0*factor*factor;
    const double err = fabs(D[IDX1F(k)] - lambda);
    max_err_D = (max_err_D > err)? max_err_D : err;
//    printf("k = %d, D = %E, lambda = %E, err = %E\n", k, D[IDX1F(k)],
lambda, err);
}
printf("\n|D - lambda|_inf = %E\n\n", max_err_D);

printf("step 11: free resources \n");
destroyMat(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    (void**)array_d_A );

workspaceFree( nbGpus, deviceList, (void**)array_d_work );

if (NULL != A) free(A);
if (NULL != D) free(D);

if (NULL != array_d_A ) free(array_d_A);
if (NULL != array_d_work) free(array_d_work);

return 0;
}

```

# Appendix I.

## EXAMPLES OF MULTIGPU LINEAR SOLVER

This chapter provides examples to perform multiGPU linear solver.

The example code enables peer-to-peer access to take advantage of NVLINK. The user can check the performance by on/off peer-to-peer access.

The example 1 solves linear system by LU with partial pivoting (**getrf** and **getrs**). It allocates distributed matrix by calling **createMat**. Then generates the matrix on host memory and copies it to distributed device memory via **memcpyH2D**.

The example 1 solves linear system by Cholesky factorization (**potrf** and **potrs**). It allocates distributed matrix by calling **createMat**. Then generates the matrix on host memory and copies it to distributed device memory via **memcpyH2D**.

The example 1 solves linear system using the inverse of an Hermitian positive definite matrix using (**potrf** and **potri**). It allocates distributed matrix by calling **createMat**. Then generates the matrix on host memory and copies it to distributed device memory via **memcpyH2D**.

## I.1. GETRF and GETRS of 1D Laplacian operator (example 1)

Please refer H.1 for util.hxx.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *     nvcc -ccbin gcc -I/usr/local/cuda/include -c main.cpp -o main.o
 *     nvcc -cudart static main.o -lcusolverMg
 */
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include "cusolverMg.h"
#include "util.hxx"

// #define SHOW_FORMAT

#ifndef IDX2F
#define IDX2F(i,j,lda) (((j)-1)*((size_t)lda))+((i)-1)
#endif /* IDX2F */

#ifndef IDX1F
#define IDX1F(i) ((i)-1)
#endif /* IDX1F */

static void print_matrix(
    int m,
    int n,
    const double *A,
    int lda,
    const char* name)
{
    printf("%s = matlab base-1, %d-by-%d matrix\n", name, m, n);
    for(int row = 1 ; row <= m ; row++){
        for(int col = 1 ; col <= n ; col++){
            double Aij = A[IDX2F(row, col, lda)];
            printf("%s(%d,%d) = %20.16E\n", name, row, col, Aij );
        }
    }
}

/* compute |x|_inf */
static double vec_nrm_inf(
    int n,
    const double *x)
{
    double max_nrm = 0;
    for(int row = 1; row <= n ; row++){
        double xi = x[ IDX1F(row) ];
        max_nrm = ( max_nrm > fabs(xi) )? max_nrm : fabs(xi);
    }
    return max_nrm;
}

```

```

...
/* A is 1D laplacian, return A(N:-1:1, :) */
static void gen_ld_laplacian_perm(
    int N,
    double *A,
    int lda)
{
    memset(A, 0, sizeof(double)*lda*N);
    for(int J = 1 ; J <= N; J++){
        A[ IDX2F( N-J+1, J, lda ) ] = 2.0;
        if ( (J-1) >= 1 ){
            A[ IDX2F( N-J+1, J-1, lda ) ] = -1.0;
        }
        if ( (J+1) <= N ){
            A[ IDX2F( N-J+1, J+1, lda ) ] = -1.0;
        }
    }
}

int main( int argc, char* argv[])
{
    cusolverMgHandle_t handle = NULL;
    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat = cudaSuccess;
    /* maximum number of GPUs */
    const int MAX_NUM_DEVICES = 16;

    int nbGpus = 0;
    int deviceList[MAX_NUM_DEVICES];

    const int N = 611;
    const int IA = 1;
    const int JA = 1;
    const int T_A = 256; /* tile size of A */
    const int lda = N;

    const int IB = 1;
    const int JB = 1;
    const int T_B = 100; /* tile size of B */
    const int ldb = N;

    double *A = NULL; /* A is N-by-N */
    double *B = NULL; /* B is N-by-1, right-hand-side vector */
    double *X = NULL; /* X is N-by-1, solution vector */
    int *IPIV = NULL; /* IPIV is 1-by-N, pivoting sequence */
    int info = 0;

    cudaLibMgMatrixDesc_t descrA;
    cudaLibMgMatrixDesc_t descrB;
    cudaLibMgGrid_t gridA;
    cudaLibMgGrid_t gridB;
    cusolverMgGridMapping_t mapping = CUDALIBMG_GRID_MAPPING_COL_MAJOR;

    double **array_d_A = NULL;
    double **array_d_B = NULL;
    int **array_d_IPIV = NULL;

    int64_t lwork_getrf = 0 ;
    int64_t lwork_getrs = 0 ;
    int64_t lwork = 0 ; /* workspace: number of elements per device */
    double **array_d_work = NULL;

    printf("test permuted 1D Laplacian of order %d\n", N);
}

```

...

```

printf("step 1: create Mg handle and select devices \n");
status = cusolverMgCreate(&handle);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat = cudaGetDeviceCount( &nbGpus );
assert( cudaSuccess == cudaStat );

nbGpus = (nbGpus < MAX_NUM_DEVICES)? nbGpus : MAX_NUM_DEVICES;
printf("\tthere are %d GPUS \n", nbGpus);
for(int j = 0 ; j < nbGpus ; j++){
    deviceList[j] = j;
    cudaDeviceProp prop;
    cudaGetDeviceProperties(&prop, j);
    printf("\tdevice %d, %s, cc %d.%d \n",j, prop.name, prop.major,
prop.minor);
}

status = cusolverMgDeviceSelect(
    handle,
    nbGpus,
    deviceList);
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 2: Enable peer access.\n");
assert( 0 == enablePeerAccess( nbGpus, deviceList ) );

printf("step 3: allocate host memory A \n");
A = (double *)malloc (sizeof(double)*lda*N);
B = (double *)malloc (sizeof(double)*ldb*1);
X = (double *)malloc (sizeof(double)*ldb*1);
IPIV = (int *)malloc (sizeof(int)*N);
assert( NULL != A );
assert( NULL != B );
assert( NULL != X );
assert( NULL != IPIV );

/* permute 1D Laplacian to enable pivoting */
printf("step 4: prepare permuted 1D Laplacian for A and B = ones(N,1) \n");
gen_ld_laplacian_perm(
    N,
    &A[ IDX2F( IA, JA, lda ) ],
    lda
);
#ifdef SHOW_FORMAT
    print_matrix( N, N, A, lda, "A");
#endif
/* B = ones(N,1) */
for(int row = 1 ; row <= N ; row++){
    B[IDX1F(row)] = 1.0;
}
printf("step 5: create matrix descriptors for A and B \n");
status = cusolverMgCreateDeviceGrid(&gridA, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
status = cusolverMgCreateDeviceGrid(&gridB, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
/* (global) A is N-by-N */
status = cusolverMgCreateMatrixDesc(
    &descrA,
    N, /* number of rows of (global) A */
    N, /* number of columns of (global) A */
    N, /* number or rows in a tile */
    T_A, /* number of columns in a tile */
    CUDA_R_64F,
    gridA );
assert(CUSOLVER_STATUS_SUCCESS == status);

```

...

```

/* (global) B is N-by-1 */
status = cusolverMgCreateMatrixDesc(
    &descrB,
    N, /* number of rows of (global) B */
    1, /* number of columns of (global) B */
    N, /* number of rows in a tile */
    T_B, /* number of columns in a tile */
    CUDA_R_64F,
    gridB );
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 6: allocate distributed matrices A, B and IPIV \n");
array_d_A = (double**)malloc(sizeof(double*)*nbGpus);
assert(NULL != array_d_A);
array_d_B = (double**)malloc(sizeof(double*)*nbGpus);
assert(NULL != array_d_B);
array_d_IPIV = (int**)malloc(sizeof(int*)*nbGpus);
assert(NULL != array_d_IPIV);

/* A := 0 */
createMat<double>(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A
);
/* B := 0 */
createMat<double>(
    nbGpus,
    deviceList,
    1, /* number of columns of global B */
    T_B, /* number of columns per column tile */
    ldb, /* leading dimension of local B */
    array_d_B
);
/* IPIV := 0, IPIV is consistent with A */
createMat<int>(
    nbGpus,
    deviceList,
    N, /* number of columns of global IPIV */
    T_A, /* number of columns per column tile */
    1, /* leading dimension of local IPIV */
    array_d_IPIV
);
printf("step 7: prepare data on devices \n");
/* distribute A to array_d_A */
memcpyH2D<double>(
    nbGpus,
    deviceList,
    N,
    N,
/* input */
    A,
    lda,
/* output */
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A, /* host pointer array of dimension nbGpus */
    IA,
    JA
);

```

...

```

/* distribute B to array_d_B */
memcpyH2D<double>(
    nbGpus,
    deviceList,
    N,
    1,
/* input */
    B,
    ldb,
/* output */
    1, /* number of columns of global B */
    T_B, /* number of columns per column tile */
    ldb, /* leading dimension of local B */
    array_d_B, /* host pointer array of dimension nbGpus */
    IB,
    JB
);

printf("step 8: allocate workspace space \n");
status = cusolverMgGetrf_bufferSize(
    handle,
    N,
    N,
    (void**)array_d_A,
    IA, /* base-1 */
    JA, /* base-1 */
    descrA,
    array_d_IPIV,
    CUDA_R_64F,
    &lwork_getrf);
assert(CUSOLVER_STATUS_SUCCESS == status);

status = cusolverMgGetrs_bufferSize(
    handle,
    CUBLAS_OP_N,
    N,
    1, /* NRHS */
    (void**)array_d_A,
    IA,
    JA,
    descrA,
    array_d_IPIV,
    (void**)array_d_B,
    IB,
    JB,
    descrB,
    CUDA_R_64F,
    &lwork_getrs);
assert(CUSOLVER_STATUS_SUCCESS == status);

lwork = (lwork_getrf > lwork_getrs)? lwork_getrf : lwork_getrs;
printf("\tallocate device workspace, lwork = %lld \n", (long long)lwork);
array_d_work = (double**)malloc(sizeof(double)*nbGpus);
assert( NULL != array_d_work);

/* array_d_work[j] points to device workspace of device j */
workspaceAlloc(
    nbGpus,
    deviceList,
    sizeof(double)*lwork, /* number of bytes per device */
    (void**)array_d_work
);
cudaStat = cudaDeviceSynchronize(); /* sync all devices */
assert(cudaSuccess == cudaStat);

```

...

```

printf("step 9: solve A*X = B by GETRF and GETRS \n");
status = cusolverMgGetrf(
    handle,
    N,
    N,
    (void**)array_d_A,
    IA,
    JA,
    descrA,
    array_d_IPIV,
    CUDA_R_64F,
    (void**)array_d_work,
    lwork,
    &info /* host */
);
assert(CUSOLVER_STATUS_SUCCESS == status);
cudaStat = cudaDeviceSynchronize(); /* sync all devices */
assert(cudaSuccess == cudaStat);
assert(0 == info); /* check if A is singular */

status = cusolverMgGetrs(
    handle,
    CUBLAS_OP_N,
    N,
    1, /* NRHS */
    (void**)array_d_A,
    IA,
    JA,
    descrA,
    array_d_IPIV,
    (void**)array_d_B,
    IB,
    JB,
    descrB,
    CUDA_R_64F,
    (void**)array_d_work,
    lwork,
    &info /* host */
);

assert(CUSOLVER_STATUS_SUCCESS == status);
cudaStat = cudaDeviceSynchronize(); /* sync all devices */
assert(cudaSuccess == cudaStat);
assert(0 == info); /* check if parameters are valid */

printf("step 10: retrieve IPIV and solution vector X \n");
memcpyD2H<double>(
    nbGpus,
    deviceList,
    N,
    1,
/* input */
    1, /* number of columns of global B */
    T_B, /* number of columns per column tile */
    ldb, /* leading dimension of local B */
    array_d_B,
    IB,
    JB,
/* output */
    X, /* N-by-1 */
    ldb
);

```

...

```

/* IPIV is consistent with A, use JA and T_A */
mempcpyD2H<int>(
    nbGpus,
    deviceList,
    1,
    N,
/* input */
    N, /* number of columns of global IPIV */
    T_A, /* number of columns per column tile */
    1, /* leading dimension of local IPIV */
    array_d_IPIV,
    1,
    JA,
/* output */
    IPIV, /* 1-by-N */
    1
);

#ifdef SHOW_FORMAT
/* X is N-by-1 */
print_matrix(N, 1, X, ldb, "X");
#endif

#ifdef SHOW_FORMAT
/* IPIV is 1-by-N */
printf("IPIV = matlab base-1, 1-by-%d matrix\n", N);
for(int row = 1; row <= N ; row++){
    printf("IPIV(%d) = %d \n", row, IPIV[ IDX1F(row) ]);
}
#endif

printf("step 11: measure residual error |b - A*x| \n");
double max_err = 0;
for(int row = 1; row <= N ; row++){
    double sum = 0.0;
    for(int col = 1; col <= N ; col++){
        double Aij = A[ IDX2F( row, col, lda ) ];
        double xj = X[ IDX1F(col) ];
        sum += Aij*xj;
    }
    double bi = B[ IDX1F(row) ];
    double err = fabs( bi - sum );

    max_err = ( max_err > err )? max_err : err;
}
double x_nrm_inf = vec_nrm_inf(N, X);
double b_nrm_inf = vec_nrm_inf(N, B);
double A_nrm_inf = 4.0;
double rel_err = max_err/(A_nrm_inf * x_nrm_inf + b_nrm_inf);
printf("\n|b - A*x|_inf = %E\n", max_err);
printf("|x|_inf = %E\n", x_nrm_inf);
printf("|b|_inf = %E\n", b_nrm_inf);
printf("|A|_inf = %E\n", A_nrm_inf);
/* relative error is around machine zero */
/* the user can use |b - A*x|/(N*|A|*|x|+|b|) as well */
printf("|b - A*x|/(|A|*|x|+|b|) = %E\n\n", rel_err);

```

...

```

printf("step 12: free resources \n");
destroyMat(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    (void**)array_d_A );
destroyMat(
    nbGpus,
    deviceList,
    1, /* number of columns of global B */
    T_B, /* number of columns per column tile */
    (void**)array_d_B );
destroyMat(
    nbGpus,
    deviceList,
    N, /* number of columns of global IPIV */
    T_A, /* number of columns per column tile */
    (void**)array_d_IPIV );

workspaceFree( nbGpus, deviceList, (void**)array_d_work );

if (NULL != A) free(A);
if (NULL != B) free(B);
if (NULL != X) free(X);
if (NULL != IPIV) free(IPIV);

if (NULL != array_d_A ) free(array_d_A);
if (NULL != array_d_B ) free(array_d_B);
if (NULL != array_d_IPIV) free(array_d_IPIV);
if (NULL != array_d_work) free(array_d_work);

return 0;
}

```

## I.2. POTRF and POTRS of 1D Laplacian operator (example 2)

Please refer H.1 for util.hxx.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *     nvcc -ccbin gcc -I/usr/local/cuda/include -c main.cpp -o main.o
 *     nvcc -cudart static main.o -lcusolverMg
 */
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include "cusolverMg.h"
#include "util.hxx"

// #define SHOW_FORMAT

#ifndef IDX2F
#define IDX2F(i,j,lda) (((j)-1)*((size_t)lda))+((i)-1)
#endif /* IDX2F */

#ifndef IDX1F
#define IDX1F(i) ((i)-1)
#endif /* IDX1F */

static void print_matrix(
    int m,
    int n,
    const double *A,
    int lda,
    const char* name)
{
    printf("%s = matlab base-1, %d-by-%d matrix\n", name, m, n);
    for(int row = 1 ; row <= m ; row++){
        for(int col = 1 ; col <= n ; col++){
            double Aij = A[IDX2F(row, col, lda)];
            printf("%s(%d,%d) = %20.16E\n", name, row, col, Aij );
        }
    }
}

/* compute |x|_inf */
static double vec_nrm_inf(
    int n,
    const double *x)
{
    double max_nrm = 0;
    for(int row = 1; row <= n ; row++){
        double xi = x[ IDX1F(row) ];
        max_nrm = ( max_nrm > fabs(xi) )? max_nrm : fabs(xi);
    }
    return max_nrm;
}

```

...

```

/* A is 1D laplacian, return A(N:-1:1, :) */
static void gen_ld_laplacian(
    int N,
    double *A,
    int lda)
{
    memset(A, 0, sizeof(double)*lda*N);
    for(int J = 1 ; J <= N; J++){
        A[ IDX2F( J, J, lda ) ] = 2.0;
        if ( (J-1) >= 1 ){
            A[ IDX2F( J, J-1, lda ) ] = -1.0;
        }
        if ( (J+1) <= N ){
            A[ IDX2F( J, J+1, lda ) ] = -1.0;
        }
    }
}

/* Generate matrix B := A * X */
static void gen_ref_B(
    int N,
    int NRHS,
    double *A,
    int lda,
    double *X,
    int ldx,
    double *B,
    int ldb)
{
    memset(B, 0, sizeof(double)*lda*NRHS);

    for(int J = 1 ; J <= NRHS; J++){
        for(int I = 1 ; I <= N; I++){
            for(int K = 1 ; K <= N; K++){
                double Aik = A[ IDX2F( I, K, lda ) ];
                double Xk = X[ IDX2F( K, J, ldx ) ];
                B[ IDX2F( I, J, ldb ) ] += (Aik * Xk);
            }
        }
    }
}

int main( int argc, char* argv[])
{
    cusolverMgHandle_t handle = NULL;
    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat = cudaSuccess;
    /* maximum local number of GPUs, set by user */
    const int MAX_NUM_DEVICES = 2;

    int nbGpus = 0;
    int deviceList[MAX_NUM_DEVICES];

```

...

```

const int NRHS = 2;
const int N    = 8;

const int IA   = 1;
const int JA   = 1;
const int T_A  = 256; /* tile size of A */
const int lda  = N;

const int IB   = 1;
const int JB   = 1;
const int T_B  = 10; /* tile size of B */
const int ldb  = N;

double *A = NULL; /* A is N-by-N */
double *B = NULL; /* B is N-by-NRHS */
double *X = NULL; /* X is N-by-NRHS */
int info = 0;

cudaLibMgMatrixDesc_t descrA;
cudaLibMgMatrixDesc_t descrB;
cudaLibMgGrid_t gridA;
cudaLibMgGrid_t gridB;
cusolverMgGridMapping_t mapping = CUDALIBMG_GRID_MAPPING_COL_MAJOR;

double **array_d_A = NULL;
double **array_d_B = NULL;

int64_t lwork_potrf = 0 ;
int64_t lwork_potrs = 0 ;
int64_t lwork = 0 ; /* workspace: number of elements per device */
double **array_d_work = NULL;

printf("test 1D Laplacian of order %d\n", N);

printf("\tstep 1: create Mg handle and select devices \n");
status = cusolverMgCreate(&handle);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat = cudaGetDeviceCount( &nbGpus );
assert( cudaSuccess == cudaStat );

nbGpus = (nbGpus < MAX_NUM_DEVICES)? nbGpus : MAX_NUM_DEVICES;
printf("\tthere are %d GPUS \n", nbGpus);
for(int j = 0 ; j < nbGpus ; j++){
    deviceList[j] = j;
    cudaDeviceProp prop;
    cudaGetDeviceProperties(&prop, j);
    printf("\tdevice %d, %s, cc %d.%d \n",j, prop.name, prop.major,
prop.minor);
}

```

...

```

status = cusolverMgDeviceSelect(
    handle,
    nbGpus,
    deviceList);
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 2: Enable peer access.\n");
assert( 0 == enablePeerAccess( nbGpus, deviceList ) );

printf("step 3: allocate host memory A \n");
A = (double *)malloc (sizeof(double)*lda*N);
B = (double *)malloc (sizeof(double)*ldb*NRHS);
X = (double *)malloc (sizeof(double)*ldb*NRHS);
assert( NULL != A );
assert( NULL != B );
assert( NULL != X );

/* permute 1D Laplacian to enable pivoting */
printf("step 4: prepare 1D Laplacian for A and X = ones(N,NRHS) \n");
gen_1d_laplacian(
    N,
    &A[ IDX2F( IA, JA, lda ) ],
    lda
);

#ifdef SHOW_FORMAT
    print_matrix( N, N, A, lda, "A");
#endif

/* X = ones(N,1) */
for(int row = 1 ; row <= N ; row++){
    for(int col = 1 ; col <= NRHS ; col++){
        X[IDX2F(row, col, ldb)] = 1.0;
    }
}

#ifdef SHOW_FORMAT
    print_matrix( N, NRHS, X, ldb, "Reference solution (X)");
#endif

/* Set B := A * X */
printf("step 5: create rhs for reference solution on host B = A*X \n");
gen_ref_B (
    N,
    NRHS,
    A, /* input */
    lda,
    X, /* input */
    ldb, /* same leading dimension as B */
    B, /* output */
    ldb);

#ifdef SHOW_FORMAT
    print_matrix( N, NRHS, B, ldb, "Generated rhs (B)");
#endif

```

...

```

printf("step 6: create matrix descriptors for A and B \n");
status = cusolverMgCreateDeviceGrid(&gridA, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
status = cusolverMgCreateDeviceGrid(&gridB, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
/* (global) A is N-by-N */
status = cusolverMgCreateMatrixDesc(
&descrA,
N, /* nubmer of rows of (global) A */
N, /* number of columns of (global) A */
N, /* number or rows in a tile */
T_A, /* number of columns in a tile */
CUDA_R_64F,
gridA );
assert(CUSOLVER_STATUS_SUCCESS == status);
/* (global) B is N-by-1 */
status = cusolverMgCreateMatrixDesc(
&descrB,
N, /* nubmer of rows of (global) B */
NRHS, /* number of columns of (global) B */
N, /* number or rows in a tile */
T_B, /* number of columns in a tile */
CUDA_R_64F,
gridB );
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 7: allocate distributed matrices A and B \n");
array_d_A = (double**)malloc(sizeof(double*)*nbGpus);
assert(NULL != array_d_A);
array_d_B = (double**)malloc(sizeof(double*)*nbGpus);
assert(NULL != array_d_B);

/* A := 0 */
createMat<double>(
nbGpus,
deviceList,
N, /* number of columns of global A */
T_A, /* number of columns per column tile */
lda, /* leading dimension of local A */
array_d_A
);
/* B := 0 */
createMat<double>(
nbGpus,
deviceList,
NRHS, /* number of columns of global B */
T_B, /* number of columns per column tile */
ldb, /* leading dimension of local B */
array_d_B
);

```

...

```

printf("step 8: prepare data on devices \n");
/* distribute A to array_d_A */
memcpyH2D<double>(
    nbGpus,
    deviceList,
    N,
    N,
/* input */
    A,
    lda,
/* output */
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A, /* host pointer array of dimension nbGpus */
    IA,
    JA
);
/* distribute B to array_d_B */
memcpyH2D<double>(
    nbGpus,
    deviceList,
    N,
    NRHS, /* number of columns of global B */
/* input */
    B,
    ldb,
/* output */
    NRHS, /* number of columns of global B */
    T_B, /* number of columns per column tile */
    ldb, /* leading dimension of local B */
    array_d_B, /* host pointer array of dimension nbGpus */
    IB,
    JB
);

printf("step 9: allocate workspace space \n");
status = cusolverMgPotrf_bufferSize(
    handle,
    CUBLAS_FILL_MODE_LOWER,
    N,
    (void**)array_d_A,
    IA, /* base-1 */
    JA, /* base-1 */
    descrA,
    CUDA_R_64F,
    &lwork_potrf);
assert(CUSOLVER_STATUS_SUCCESS == status);

```

...

```

    status = cusolverMgPotrs_bufferSize(
        handle,
        CUBLAS_FILL_MODE_LOWER,
        N,
        NRHS, /* number of columns of B */
        (void**)array_d_A,
        IA,
        JA,
        descrA,
        (void**)array_d_B,
        IB,
        JB,
        descrB,
        CUDA_R_64F,
        &lwork_potrs);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    lwork = (lwork_potrf > lwork_potrs)? lwork_potrf : lwork_potrs;
    printf("\tallocate device workspace, lwork = %lld \n", (long long)lwork);

    array_d_work = (double**)malloc(sizeof(double*)*nbGpus);
    assert( NULL != array_d_work);
    /* array_d_work[j] points to device workspace of device j */
    workspaceAlloc(
        nbGpus,
        deviceList,
        sizeof(double)*lwork, /* number of bytes per device */
        (void**)array_d_work
    );
    cudaStat = cudaDeviceSynchronize(); /* sync all devices */
    assert(cudaSuccess == cudaStat);

    printf("step 10: solve A*X = B by POTRF and POTRS \n");
    status = cusolverMgPotrf(
        handle,
        CUBLAS_FILL_MODE_LOWER,
        N,
        (void**)array_d_A,
        IA,
        JA,
        descrA,
        CUDA_R_64F,
        (void**)array_d_work,
        lwork,
        &info /* host */
    );
    assert(CUSOLVER_STATUS_SUCCESS == status);
    cudaStat = cudaDeviceSynchronize(); /* sync all devices */
    assert(cudaSuccess == cudaStat);
    assert(0 == info); /* check if A is singular */

```

...

```

    status = cusolverMgPotrs(
        handle,
        CUBLAS_FILL_MODE_LOWER,
        N,
        NRHS, /* number of columns of B */
        (void**)array_d_A,
        IA,
        JA,
        descrA,
        (void**)array_d_B,
        IB,
        JB,
        descrB,
        CUDA_R_64F,
        (void**)array_d_work,
        lwork,
        &info /* host */
    );
    assert(CUSOLVER_STATUS_SUCCESS == status);
    cudaStat = cudaDeviceSynchronize(); /* sync all devices */
    assert(cudaSuccess == cudaStat);
    assert(0 == info); /* check if parameters are valid */

    printf("step 11: solution vector B \n");
    memcpyD2H<double>(
        nbGpus,
        deviceList,
        N,
        NRHS,
/* input */
        NRHS, /* number of columns of global B */
        T_B, /* number of columns per column tile */
        ldb, /* leading dimension of local B */
        array_d_B,
        IB,
        JB,
/* output */
        B, /* N-by-1 */
        ldb
    );

```

...

```

#ifdef SHOW_FORMAT
    /* X is N-by-NRHS */
    print_matrix(N, NRHS, B, ldb, "Computed solution (B)");
    print_matrix(N, NRHS, X, ldb, "Reference solution (X)");
#endif
printf("step 12: measure residual error |x - x_ref| \n");
double max_err = 0;
for(int col = 1; col <= NRHS ; col++){
    printf("errors for X[:,%d] \n", col);
    double err = 0.0; /* absolute error per column */
    for(int row = 1; row <= N ; row++){
        double xij = X[ IDX2F(row, col, ldb) ];
        double bij = B[ IDX2F(row, col, ldb) ];
        double err = fabs(xij - bij);
        max_err = ( err > max_err ) ? err : max_err;
    }
    double x_nrm_inf = vec_nrm_inf(N, &X[ IDX2F( 1, col, ldb)]);
    double b_nrm_inf = vec_nrm_inf(N, &B[ IDX2F( 1, col, ldb)]);
    double A_nrm_inf = 4.0;
    double rel_err = max_err/(A_nrm_inf * x_nrm_inf + b_nrm_inf);
    printf("\t|b - A*x|_inf = %E\n", max_err);
    printf("\t|x|_inf = %E\n", x_nrm_inf);
    printf("\t|b|_inf = %E\n", b_nrm_inf);
    printf("\t|A|_inf = %E\n", A_nrm_inf);
    /* relative error is around machine zero */
    /* the user can use |b - A*x|/(N*|A|*|x|+|b|) as well */
    printf("\t|b - A*x|/(|A|*|x|+|b|) = %E\n\n", rel_err);
}

printf("step 13: free resources \n");
destroyMat(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    (void**)array_d_A );
destroyMat(
    nbGpus,
    deviceList,
    1, /* number of columns of global B */
    T_B, /* number of columns per column tile */
    (void**)array_d_B );

workspaceFree( nbGpus, deviceList, (void**)array_d_work );

if (NULL != A) free(A);
if (NULL != B) free(B);
if (NULL != X) free(X);

if (NULL != array_d_A ) free(array_d_A);
if (NULL != array_d_B ) free(array_d_B);
if (NULL != array_d_work) free(array_d_work);

return 0;
}

```

## I.3. POTRF and POTRI of 1D Laplacian operator (example 3)

Please refer H.1 for util.hxx.

```

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *     nvcc -ccbin gcc -I/usr/local/cuda/include -c main.cpp -o main.o
 *     nvcc -cudart static main.o -lcusolverMg
 */
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <assert.h>
#include <cuda_runtime.h>
#include "cusolverMg.h"
#include "util.hxx"

// #define SHOW_FORMAT

#ifndef IDX2F
#define IDX2F(i,j,lda) (((j)-1)*((size_t)lda))+((i)-1)
#endif /* IDX2F */

#ifndef IDX1F
#define IDX1F(i) ((i)-1)
#endif /* IDX1F */

static void print_matrix(
    int m,
    int n,
    const double *A,
    int lda,
    const char* name)
{
    printf("%s = matlab base-1, %d-by-%d matrix\n", name, m, n);
    for(int row = 1 ; row <= m ; row++){
        for(int col = 1 ; col <= n ; col++){
            double Aij = A[IDX2F(row, col, lda)];
            printf("%s(%d,%d) = %20.16E\n", name, row, col, Aij );
        }
    }
}

/* compute |x|_inf */
static double vec_nrm_inf(
    int n,
    const double *x)
{
    double max_nrm = 0;
    for(int row = 1; row <= n ; row++){
        double xi = x[ IDX1F(row) ];
        max_nrm = ( max_nrm > fabs(xi) )? max_nrm : fabs(xi);
    }
    return max_nrm;
}

```

...

```

/* A is 1D laplacian, return A(N:-1:1, :) */
static void gen_ld_laplacian(
    int N,
    double *A,
    int lda)
{
    memset(A, 0, sizeof(double)*lda*N);
    for(int J = 1 ; J <= N; J++){
        A[ IDX2F( J, J, lda ) ] = 2.0;
        if ( (J-1) >= 1 ){
            A[ IDX2F( J, J-1, lda ) ] = -1.0;
        }
        if ( (J+1) <= N ){
            A[ IDX2F( J, J+1, lda ) ] = -1.0;
        }
    }
}

/* Generate matrix B := A * X */
static void gen_ref_B(
    int N,
    int NRHS,
    double *A,
    int lda,
    double *X,
    int ldx,
    double *B,
    int ldb)
{
    memset(B, 0, sizeof(double)*lda*NRHS);

    for(int J = 1 ; J <= NRHS; J++){
        for(int I = 1 ; I <= N; I++){
            for(int K = 1 ; K <= N; K++){
                double Aik = A[ IDX2F( I, K, lda ) ];
                double Xk = X[ IDX2F( K, J, ldx ) ];
                B[ IDX2F( I, J, ldb ) ] += (Aik * Xk);
            }
        }
    }
}

```

...

```

/* Apply inverse to RHS matrix */
static void solve_system_with_invA(
    int N,
    int NRHS,
    double *A,
    int lda,
    double *B,
    int ldb,
    double *X,
    int ldx)
{
    /* Extend lower triangular A to full matrix */
    for(int I = 1 ; I <= N; I++){
        for(int J = (I +1) ; J <= N; J++){
            A[ IDX2F( I, J, lda) ] = A[ IDX2F( J, I, lda) ];
        }
    }

#ifdef SHOW_FORMAT
    print_matrix( N, N, A, lda, "full inv(A)");
#endif

    /* Reset input matrix to 0 */
    memset(X, 0, sizeof(double)*lda*NRHS);

    /* Apply full inv(A) by matrix-matrix multiplication */

    for(int J = 1 ; J <= NRHS; J++){
        for(int I = 1 ; I <= N; I++){
            for(int K = 1 ; K <= N; K++){
                double Aik = A[ IDX2F( I, K, lda) ];
                double Bk = B[ IDX2F( K, J, ldx) ];
                X[ IDX2F( I, J, ldx) ] += (Aik * Bk);
            }
        }
    }
};

int main( int argc, char* argv[])
{
    cusolverMgHandle_t handle = NULL;
    cusolverStatus_t status = CUSOLVER_STATUS_SUCCESS;
    cudaError_t cudaStat = cudaSuccess;
    /* maximum local number of GPUs, set by user */
    const int MAX_NUM_DEVICES = 2;

    int nbGpus = 0;
    int deviceList[MAX_NUM_DEVICES];

```

...

```

const int NRHS = 2;
const int N    = 8;

const int IA   = 1;
const int JA   = 1;
const int T_A  = 256; /* tile size of A */
const int lda  = N;

const int IB   = 1;
const int JB   = 1;
const int T_B  = 10; /* tile size of B */
const int ldb  = N;

double *A = NULL; /* A is N-by-N */
double *B = NULL; /* B is N-by-NRHS */
double *Xref = NULL; /* X is N-by-NRHS */
double *Xans = NULL; /* X is N-by-NRHS */
int info = 0;

cudaLibMgMatrixDesc_t descrA;
cudaLibMgGrid_t gridA;
cusolverMgGridMapping_t mapping = CUDALIBMG_GRID_MAPPING_COL_MAJOR;

double **array_d_A = NULL;

int64_t lwork_potrf = 0 ;
int64_t lwork_potri = 0 ;
int64_t lwork = 0 ; /* workspace: number of elements per device */
double **array_d_work = NULL;

printf("test 1D Laplacian of order %d\n", N);

printf("step 1: create Mg handle and select devices \n");
status = cusolverMgCreate(&handle);
assert(CUSOLVER_STATUS_SUCCESS == status);

cudaStat = cudaGetDeviceCount( &nbGpus );
assert( cudaSuccess == cudaStat );

nbGpus = (nbGpus < MAX_NUM_DEVICES)? nbGpus : MAX_NUM_DEVICES;
printf("\tthere are %d GPUs \n", nbGpus);
for(int j = 0 ; j < nbGpus ; j++){
    deviceList[j] = j;
    cudaDeviceProp prop;
    cudaGetDeviceProperties(&prop, j);
    printf("\tdevice %d, %s, cc %d.%d \n",j, prop.name, prop.major,
prop.minor);
}

status = cusolverMgDeviceSelect(
    handle,
    nbGpus,
    deviceList);
assert(CUSOLVER_STATUS_SUCCESS == status);

```

...

```

printf("step 2: Enable peer access.\n");
assert( 0 == enablePeerAccess( nbGpus, deviceList ) );

printf("step 3: allocate host memory A \n");
A = (double *)malloc (sizeof(double)*lda*N);
B = (double *)malloc (sizeof(double)*ldb*NRHS);
Xref = (double *)malloc (sizeof(double)*ldb*NRHS);
Xans = (double *)malloc (sizeof(double)*ldb*NRHS);
assert( NULL != A );
assert( NULL != B );
assert( NULL != Xref );
assert( NULL != Xans );

/* permute 1D Laplacian to enable pivoting */
printf("step 4: prepare 1D Laplacian for A and Xref = ones(N,NRHS) \n");
gen_ld_laplacian(
    N,
    &A[ IDX2F( IA, JA, lda ) ],
    lda
);

#ifdef SHOW_FORMAT
    print_matrix( N, N, A, lda, "A");
#endif

/* X = ones(N,1) */
for(int row = 1 ; row <= N ; row++){
    for(int col = 1 ; col <= NRHS ; col++){
        Xref[IDX2F(row, col, ldb)] = 1.0;
    }
}

#ifdef SHOW_FORMAT
    print_matrix( N, NRHS, Xref, ldb, "Reference solution (X)");
#endif

/* Set B := A * X */
printf("step 5: create rhs for reference solution on host B = A*Xref \n");
gen_ref_B (
    N,
    NRHS,
    A, /* input */
    lda,
    Xref, /* input */
    ldb, /* same leading dimension as B */
    B, /* output */
    ldb);

#ifdef SHOW_FORMAT
    print_matrix( N, NRHS, B, ldb, "Generated rhs (B)");
#endif

```

...

```

printf("step 6: create matrix descriptors for A and B \n");
status = cusolverMgCreateDeviceGrid(&gridA, 1, nbGpus, deviceList,
mapping );
assert(CUSOLVER_STATUS_SUCCESS == status);
/* (global) A is N-by-N */
status = cusolverMgCreateMatrixDesc(
    &descrA,
    N, /* number of rows of (global) A */
    N, /* number of columns of (global) A */
    N, /* number of rows in a tile */
    T_A, /* number of columns in a tile */
    CUDA_R_64F,
    gridA );
assert(CUSOLVER_STATUS_SUCCESS == status);

printf("step 7: allocate distributed matrices A and B \n");
array_d_A = (double**)malloc(sizeof(double*)*nbGpus);
assert(NULL != array_d_A);

/* A := 0 */
createMat<double>(
    nbGpus,
    deviceList,
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A
);

printf("step 8: prepare data on devices \n");
/* distribute A to array_d_A */
memcpyH2D<double>(
    nbGpus,
    deviceList,
    N,
    N,
/* input */
    A,
    lda,
/* output */
    N, /* number of columns of global A */
    T_A, /* number of columns per column tile */
    lda, /* leading dimension of local A */
    array_d_A, /* host pointer array of dimension nbGpus */
    IA,
    JA
);

printf("step 9: allocate workspace space \n");
status = cusolverMgPotrf_bufferSize(
    handle,
    CUBLAS_FILL_MODE_LOWER,
    N,
    (void**)array_d_A,
    IA, /* base-1 */
    JA, /* base-1 */
    descrA,
    CUDA_R_64F,
    &lwork_potrf);
assert(CUSOLVER_STATUS_SUCCESS == status);

```

...

```

    status = cusolverMgPotri_bufferSize(
        handle,
        CUBLAS_FILL_MODE_LOWER,
        N,
        (void**)array_d_A,
        IA,
        JA,
        descrA,
        CUDA_R_64F,
        &lwork_potri);
    assert(CUSOLVER_STATUS_SUCCESS == status);

    lwork = (lwork_potrf > lwork_potri)? lwork_potrf : lwork_potri;
    printf("\tallocate device workspace, lwork = %lld \n", (long long)lwork);

    array_d_work = (double**)malloc(sizeof(double*)*nbGpus);
    assert( NULL != array_d_work);
    /* array_d_work[j] points to device workspace of device j */
    workspaceAlloc(
        nbGpus,
        deviceList,
        sizeof(double)*lwork, /* number of bytes per device */
        (void**)array_d_work
    );
    cudaStat = cudaDeviceSynchronize(); /* sync all devices */
    assert(cudaSuccess == cudaStat);

    printf("step 10: solve A*X = B by POTRF and POTRI \n");
    status = cusolverMgPotrf(
        handle,
        CUBLAS_FILL_MODE_LOWER,
        N,
        (void**)array_d_A,
        IA,
        JA,
        descrA,
        CUDA_R_64F,
        (void**)array_d_work,
        lwork,
        &info /* host */
    );
    assert(CUSOLVER_STATUS_SUCCESS == status);
    cudaStat = cudaDeviceSynchronize(); /* sync all devices */
    assert(cudaSuccess == cudaStat);
    assert(0 == info); /* check if A is singular */

```

...

```

    status = cusolverMgPotri(
        handle,
        CUBLAS_FILL_MODE_LOWER,
        N,
        (void**)array_d_A,
        IA,
        JA,
        descrA,
        CUDA_R_64F,
        (void**)array_d_work,
        lwork,
        &info /* host */
    );
    assert(CUSOLVER_STATUS_SUCCESS == status);
    cudaStat = cudaDeviceSynchronize(); /* sync all devices */
    assert(cudaSuccess == cudaStat);
    assert(0 == info); /* check if parameters are valid */

    printf("step 11: Gather inv(A) from devices to host \n");
    memcpyD2H<double>(
        nbGpus,
        deviceList,
        N,
        N,
/* input */
        N, /* number of columns of global B */
        T_A, /* number of columns per column tile */
        ldb, /* leading dimension of local B */
        array_d_A,
        IA,
        JA,
/* output */
        A, /* N-by-1 */
        ldb
    );

#ifdef SHOW_FORMAT
    /* A is N-by-N */
    print_matrix(N, N, A, lda, "Computed solution inv(A)");
#endif

    printf("step 12: solve linear system B := inv(A) * B \n");
    solve_system_with_invA(
        N,
        NRHS,
        A,
        lda,
        B,
        ldb,
        Xans,
        ldb);

```

...

```

#ifdef SHOW_FORMAT
    /* A is N-by-N */
    print_matrix(N, NRHS, Xans, ldb, "Computed solution Xans");
#endif

    printf("step 13: measure residual error |Xref - Xans| \n");
    double max_err = 0.0;
    for(int col = 1; col <= NRHS ; col++){
        printf("errors for X[:,%d] \n", col);
        double err = 0.0; /* absolute error per column */
        for(int row = 1; row <= N ; row++){
            double Xref_ij = Xref[ IDX2F(row, col, ldb) ];
            double Xans_ij = Xans[ IDX2F(row, col, ldb) ];
            double err = fabs(Xref_ij - Xans_ij);
            max_err = ( err > max_err ) ? err : max_err;
        }
        double Xref_nrm_inf = vec_nrm_inf(N, &Xref[ IDX2F( 1, col, ldb) ]);
        double Xans_nrm_inf = vec_nrm_inf(N, &Xans[ IDX2F( 1, col, ldb) ]);
        double A_nrm_inf = 4.0;
        double rel_err = max_err / (A_nrm_inf * Xans_nrm_inf + Xref_nrm_inf);
        printf("\t|b - A*x|_inf = %E\n", max_err);
        printf("\t|Xref|_inf = %E\n", Xref_nrm_inf);
        printf("\t|Xans|_inf = %E\n", Xans_nrm_inf);
        printf("\t|A|_inf = %E\n", A_nrm_inf);
        /* relative error is around machine zero */
        /* the user can use |b - A*x| / (N*|A|*|x|+|b|) as well */
        printf("\t|b - A*x| / (|A|*|x|+|b|) = %E\n\n", rel_err);
    }

    printf("step 14: free resources \n");
    destroyMat(
        nbGpus,
        deviceList,
        N, /* number of columns of global A */
        T_A, /* number of columns per column tile */
        (void**)array_d_A );

    workspaceFree( nbGpus, deviceList, (void**)array_d_work );

    if (NULL != A) free(A);
    if (NULL != B) free(B);
    if (NULL != Xref) free(Xref);
    if (NULL != Xans) free(Xans);

    if (NULL != array_d_A ) free(array_d_A);
    if (NULL != array_d_work) free(array_d_work);

    return 0;
}

```

# Appendix J.

## ACKNOWLEDGEMENTS

NVIDIA would like to thank the following individuals and institutions for their contributions:

- ▶ CPU LAPACK routines from netlib, CLAPACK-3.2.1 (<http://www.netlib.org/clapack/>)

The following is license of CLAPACK-3.2.1.

Copyright (c) 1992-2008 The University of Tennessee. All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

- Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.
- Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer listed in this license in the documentation and/or other materials provided with the distribution.
- Neither the name of the copyright holders nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT OWNER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

- ▶ METIS-5.1.0 (<http://glaros.dtc.umn.edu/gkhome/metis/metis/overview>)

The following is license of METIS (Apache 2.0 license).

Copyright 1995-2013, Regents of the University of Minnesota

Licensed under the Apache License, Version 2.0 (the "License"); you may not use this file except in compliance with the License. You may obtain a copy of the License at

<http://www.apache.org/licenses/LICENSE-2.0>

Unless required by applicable law or agreed to in writing, software distributed under the License is distributed on an "AS IS" BASIS, WITHOUT WARRANTIES OR CONDITIONS OF ANY KIND, either express or implied. See the License for the specific language governing permissions and limitations under the License.

- ▶ QD (A C++/fortran-90 double-double and quad-double package) (<http://crd-legacy.lbl.gov/~dhbailey/mpdist/>)

The following is license of QD (modified BSD license).

Copyright (c) 2003-2009, The Regents of the University of California, through Lawrence Berkeley National Laboratory (subject to receipt of any required approvals from U.S. Dept. of Energy) All rights reserved.

1. Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

(1) Redistributions of source code must retain the copyright notice, this list of conditions and the following disclaimer.

(2) Redistributions in binary form must reproduce the copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.

(3) Neither the name of the University of California, Lawrence Berkeley National Laboratory, U.S. Dept. of Energy nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

2. THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT OWNER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

3. You are under no obligation whatsoever to provide any bug fixes, patches, or upgrades to the features, functionality or performance of the source code ("Enhancements") to anyone; however, if you choose to make your Enhancements

available either publicly, or directly to Lawrence Berkeley National Laboratory, without imposing a separate written license agreement for such Enhancements, then you hereby grant the following license: a non-exclusive, royalty-free perpetual license to install, use, modify, prepare derivative works, incorporate into other computer software, distribute, and sublicense such enhancements or derivative works thereof, in binary and source code form.

# Appendix K.

## BIBLIOGRAPHY

- [1] Timothy A. Davis, Direct Methods for sparse Linear Systems, siam 2006.
- [2] E. Chuthill and J. McKee, reducing the bandwidth of sparse symmetric matrices, ACM '69 Proceedings of the 1969 24th national conference, Pages 157-172.
- [3] Alan George, Joseph W. H. Liu, An Implementation of a Pseudoperipheral Node Finder, ACM Transactions on Mathematical Software (TOMS) Volume 5 Issue 3, Sept. 1979 Pages 284-295.
- [4] J. R. Gilbert and T. Peierls, Sparse partial pivoting in time proportional to arithmetic operations, SIAM J. Sci. Statist. Comput., 9 (1988), pp. 862-874.
- [5] Alan George and Esmond Ng, An Implementation of Gaussian Elimination with Partial Pivoting for Sparse Systems, SIAM J. Sci. and Stat. Comput., 6(2), 390-409.
- [6] Alan George and Esmond Ng, Symbolic Factorization for Sparse Gaussian Elimination with Partial Pivoting, SIAM J. Sci. and Stat. Comput., 8(6), 877-898.
- [7] John R. Gilbert, Xiaoye S. Li, Esmond G. Ng, Barry W. Peyton, Computing Row and Column Counts for Sparse QR and LU Factorization, BIT 2001, Vol. 41, No. 4, pp. 693-711.
- [8] Patrick R. Amestoy, Timothy A. Davis, Iain S. Duff, An Approximate Minimum Degree Ordering Algorithm, SIAM J. Matrix Analysis Applic. Vol 17, no 4, pp. 886-905, Dec. 1996.
- [9] Alan George, Joseph W. Liu, A Fast Implementation of the Minimum Degree Algorithm Using Quotient Graphs, ACM Transactions on Mathematical Software, Vol 6, No. 3, September 1980, page 337-358.
- [10] Alan George, Joseph W. Liu, Computer Solution of Large Sparse Positive Definite Systems, Englewood Cliffs, New Jersey: Prentice-Hall, 1981.
- [11] Iain S. Duff, ALGORITHM 575 Permutations for a Zero-Free Diagonal, ACM Transactions on Mathematical Software, Vol 7, No 3, September 1981, Page 387-390
- [12] Iain S. Duff and Jacko Koster, On algorithms for permuting large entries to the diagonal of a sparse matrix, SIAM Journal on Matrix Analysis and Applications, 2001, Vol. 22, No. 4 : pp. 973-996

- [13] "A Fast and Highly Quality Multilevel Scheme for Partitioning Irregular Graphs".  
George Karypis and Vipin Kumar. SIAM Journal on Scientific Computing, Vol. 20, No. 1,  
pp. 359-392, 1999.

## **Notice**

ALL NVIDIA DESIGN SPECIFICATIONS, REFERENCE BOARDS, FILES, DRAWINGS, DIAGNOSTICS, LISTS, AND OTHER DOCUMENTS (TOGETHER AND SEPARATELY, "MATERIALS") ARE BEING PROVIDED "AS IS." NVIDIA MAKES NO WARRANTIES, EXPRESSED, IMPLIED, STATUTORY, OR OTHERWISE WITH RESPECT TO THE MATERIALS, AND EXPRESSLY DISCLAIMS ALL IMPLIED WARRANTIES OF NONINFRINGEMENT, MERCHANTABILITY, AND FITNESS FOR A PARTICULAR PURPOSE.

Information furnished is believed to be accurate and reliable. However, NVIDIA Corporation assumes no responsibility for the consequences of use of such information or for any infringement of patents or other rights of third parties that may result from its use. No license is granted by implication of otherwise under any patent rights of NVIDIA Corporation. Specifications mentioned in this publication are subject to change without notice. This publication supersedes and replaces all other information previously supplied. NVIDIA Corporation products are not authorized as critical components in life support devices or systems without express written approval of NVIDIA Corporation.

## **Trademarks**

NVIDIA and the NVIDIA logo are trademarks or registered trademarks of NVIDIA Corporation in the U.S. and other countries. Other company and product names may be trademarks of the respective companies with which they are associated.

## **Copyright**

© 2014-2020 NVIDIA Corporation. All rights reserved.