# TABLE OF CONTENTS

Chapter 1. Introduction ................................................................. 1  
1.1. Naming Conventions ................................................................. 1  
1.2. Asynchronous Execution ............................................................ 2  
1.3. Static Library support ................................................................. 2  

Chapter 2. Using the cuSPARSE API ................................................ 4  
2.1. Thread Safety ............................................................................ 4  
2.2. Scalar Parameters ....................................................................... 4  
2.3. Parallelism with Streams ............................................................. 4  

Chapter 3. cuSPARSE Indexing and Data Formats ................................. 6  
3.1. Index Base Format ................................................................. 6  
3.2. Vector Formats ........................................................................ 6  
3.2.1. Dense Format ................................................................. 6  
3.2.2. Sparse Format ................................................................. 6  
3.3. Matrix Formats ........................................................................ 7  
3.3.1. Dense Format ................................................................. 7  
3.3.2. Coordinate Format (COO) .................................................... 8  
3.3.3. Compressed Sparse Row Format (CSR) ................................. 8  
3.3.4. Compressed Sparse Column Format (CSC) ............................... 9  
3.3.5. Ellpack-Itpack Format (ELL) ................................................ 10  
3.3.6. Hybrid Format (HYB) ........................................................ 11  
3.3.7. Block Compressed Sparse Row Format (BSR) ......................... 11  
3.3.8. Extended BSR Format (BSRX) ............................................. 13  

Chapter 4. cuSPARSE Types Reference .......................................... 15  
4.1. Data types .............................................................................. 15  
4.2. cusparseAction_t ................................................................ ...... 15  
4.3. cusparseDirection_t ................................................................. 15  
4.4. cusparseHandle_t ...................................................................... 16  
4.5. cusparseHybMat_t .................................................................... 16  
4.5.1. cusparseHybPartition_t ...................................................... 16  
4.6. cusparseMatDescr_t ................................................................. 16  
4.6.1. cusparseDiagType_t ........................................................... 17  
4.6.2. cusparseFillMode_t ........................................................... 17  
4.6.3. cusparseIndexBase_t .......................................................... 17  
4.6.4. cusparseMatrixType_t .......................................................... 17  
4.7. cusparseOperation_t ............................................................... 18  
4.8. cusparsePointerMode_t ............................................................. 18  
4.9. cusparseAlgMode_t ................................................................. 18  
4.10. cusparseSolvePolicy_t ............................................................ 19  
4.11. cusparseSolveAnalysisInfo_t .................................................. 19  
4.12. cusparseSolveAnalysisInfo_t .................................................. 19
4.13. csrsv2Info_t............................................................................................. 19
4.14. csrsm2Info_t............................................................................................ 19
4.15. csric02Info_t............................................................................................ 19
4.16. csrilu02Info_t.......................................................................................... 20
4.17. bsrsv2Info_t........................................................................................... 20
4.18. bsrsm2Info_t............................................................................................ 20
4.19. bsrjc02Info_t............................................................................................ 20
4.20. bsrlu02Info_t........................................................................................... 20
4.21. csrgemm2Info_t........................................................................................ 20
4.22. cusparseStatus_t....................................................................................... 20

Chapter 5. cuSPARSE Helper Function Reference....................................................... 22
5.1. cusparseCreate().........................................................................................22
5.2. cusparseCreateSolveAnalysisInfo().................................................................... 22
5.3. cusparseCreateHybMat()................................................................................23
5.4. cusparseCreateMatDescr()..............................................................................23
5.5. cusparseCreateSolveAnalysisInfo().................................................................... 23
5.6. cusparseDestroy()........................................................................................24
5.7. cusparseDestroySolveAnalysisInfo()................................................................... 24
5.8. cusparseDestroyHybMat()...............................................................................24
5.9. cusparseDestroyMatDescr().............................................................................24
5.10. cusparseDestroySolveAnalysisInfo()..................................................................25
5.11. cusparseGetLevelInfo()................................................................................25
5.12. cusparseGetMatDiagType()............................................................................26
5.13. cusparseGetMatFillMode()............................................................................26
5.14. cusparseGetMatIndexBase()..........................................................................26
5.15. cusparseGetMatType().................................................................................27
5.16. cusparseGetPointerMode()............................................................................27
5.17. cusparseGetVersion()..................................................................................27
5.18. cusparseSetMatDiagType()............................................................................28
5.19. cusparseSetMatFillMode().............................................................................28
5.20. cusparseSetMatIndexBase()..........................................................................29
5.21. cusparseSetMatType().................................................................................29
5.22. cusparseSetPointerMode()............................................................................29
5.23. cusparseSetStream()...................................................................................30
5.24. cusparseGetStream()..................................................................................30
5.25. cusparseCreateCsrsv2Info()...........................................................................31
5.26. cusparseDestroyCsrsv2Info()..........................................................................31
5.27. cusparseCreateCsrsm2Info()..........................................................................31
5.28. cusparseDestroyCsrsm2Info().........................................................................32
5.29. cusparseCreateCsrjc02Info()..........................................................................32
5.30. cusparseDestroyCsrjc02Info().........................................................................32
5.31. cusparseCreateCsrilu02Info().........................................................................33
5.32. cusparseDestroyCsrilu02Info()........................................................................33
Chapter 6. cuSPARSE Level 1 Function Reference

6.1. cusparse<t>axpyi()................................................................. 38
6.2. cusparse<t>doti()................................................................ 39
6.3. cusparse<t>dotci().............................................................. 41
6.4. cusparse<t>gthr().............................................................. 42
6.5. cusparse<t>gthrz()............................................................. 43
6.6. cusparse<t>roti()............................................................... 44
6.7. cusparse<t>sctr()............................................................... 45

Chapter 7. cuSPARSE Level 2 Function Reference

7.1. cusparse<t>bsrmv()............................................................ 48
7.2. cusparse<t>bsrxmlv()........................................................ 51
7.3. cusparse<t>csrmv()............................................................ 55
7.4. cusparseCsrmvEx()........................................................... 57
7.5. cusparseCsrmvEx_bufferSize()........................................... 58
7.6. cusparse<t>csrmv_mp()...................................................... 59
7.7. cusparse<t>gemvix().......................................................... 61
7.8. cusparse<t>gemvi_bufferSize().......................................... 64
7.9. cusparse<t>bsrsv2_bufferSize()......................................... 66
7.10. cusparse<t>bsrsv2_analysis()............................................ 69
7.11. cusparse<t>bsrsv2_solve()............................................... 72
7.12. cusparseXbsrsv2_zeroPivot()........................................... 76
7.13. cusparse<t>csrsv_analysis()............................................ 77
7.14. cusparseCsrsv_analysisEx()............................................ 79
7.15. cusparse<t>csrsv_solve()................................................ 80
7.16. cusparseCsrsv_solveEx()................................................ 82
7.17. cusparse<t>csrsv2_bufferSize()........................................ 83
7.18. cusparse<t>csrsv2_analysis().......................................... 85
7.19. cusparse<t>csrsv2_solve()............................................... 88
7.20. cusparseXcsrsv2_zeroPivot().......................................... 92
7.21. cusparse<t>hybmv()......................................................... 93
7.22. cusparse<t>hybsv_analysis()........................................... 94
7.23. cusparse<t>hybsv_solve() .......................................................................................... 96
7.24. cusparseCsr2cscEx2() .............................................................................................. 97
7.25. cusparseCsr2cscEx2_bufferSize() ........................................................................... 99

Chapter 8. cuSPARSE Level 3 Function Reference ...................................................... 100
8.1. cusparse<t>csrmm() ................................................................................................. 101
8.2. cusparse<t>csrmm2() ............................................................................................... 104
8.3. cusparse<t>csrsrn_analysis() .................................................................................. 108
8.4. cusparse<t>csrsm_solve() ........................................................................................ 111
8.5. cusparse<t>csrsm2_bufferSizeExt() ....................................................................... 114
8.6. cusparse<t>csrsm2_analysis() ................................................................................ 117
8.7. cusparse<t>cssrm2_solve() ..................................................................................... 120
8.8. cusparseXcsrsm2_zeroPivot() ................................................................................. 123
8.9. cusparse<t>bsrmm() .............................................................................................. 124
8.10. cusparse<t>bsrsm2_bufferSize() ......................................................................... 128
8.11. cusparse<t>bsrsm2_analysis() ............................................................................. 131
8.12. cusparse<t>bsrsm2_solve() .................................................................................. 135
8.13. cusparseXbsrsm2_zeroPivot() .............................................................................. 138
8.14. cusparse<t>gemm() ............................................................................................... 139

Chapter 9. cuSPARSE Extra Function Reference ...................................................... 141
9.1. cusparse<t>csrgeam() ............................................................................................ 142
9.2. cusparse<t>csrgeam2() .......................................................................................... 146
9.3. cusparse<t>csrgeam() ............................................................................................ 150
9.4. cusparse<t>csrgeam2() .......................................................................................... 154

Chapter 10. cuSPARSE Preconditioners Reference .................................................. 159
10.1. Incomplete Cholesky Factorization: level 0 .......................................................... 159
10.1.1. cusparse<t>csr0() ............................................................................................... 160
10.1.2. cusparse<t>csr02_bufferSize() ......................................................................... 162
10.1.3. cusparse<t>csr02_analysis() .......................................................................... 164
10.1.4. cusparse<t>csr02() .......................................................................................... 167
10.1.5. cusparseXcsr02_zeroPivot() .......................................................................... 171
10.1.6. cusparse<t>bsr02_bufferSize() ....................................................................... 172
10.1.7. cusparse<t>bsr02_analysis() .......................................................................... 175
10.1.8. cusparse<t>bsr02() ........................................................................................ 178
10.1.9. cusparseXbsr02_zeroPivot() ......................................................................... 182
10.2. Incomplete LU Factorization: level 0 .................................................................... 182
10.2.1. cusparse<t>csrilu0() ....................................................................................... 183
10.2.2. cusparseCsrilu0Ex() ...................................................................................... 185
10.2.3. cusparse<t>csrilu02_numericBoost() ............................................................. 186
10.2.4. cusparse<t>csrilu02_bufferSize() ................................................................... 187
10.2.5. cusparse<t>csrilu02_analysis() ..................................................................... 189
10.2.6. cusparse<t>csrilu02() .................................................................................... 192
10.2.7. cusparseXcsrilu02_zeroPivot() ..................................................................... 196
10.2.8. cusparse<t>bsrilu02_numericBoost() ............................................................. 197
10.2.9. cusparse<t>bsrilu02_bufferSize() ............................................................. 199
10.2.10. cusparse<t>bsrilu02_analysis() .............................................................. 202
10.2.11. cusparse<t>bsrilu02() ....................................................................... 205
10.2.12. cusparseXbsrilu02_zeroPivot() ............................................................ 209
10.3. tridiagonal solve ...................................................................................... 209
  10.3.1. cusparse<t>gtsv() ........................................................................... 210
  10.3.2. cusparse<t>gtsv_nopivot() ................................................................ 212
  10.3.3. cusparse<t>gtsv2_buffSizeExt() ........................................................... 214
  10.3.4. cusparse<t>gtsv2() ........................................................................... 216
  10.3.5. cusparse<t>gtsv2_nopivot_bufferSizeExt() .......................................... 218
  10.3.6. cusparse<t>gtsv2_nopivot() ............................................................... 220
10.4. batched tridiagonal solve ........................................................................ 221
  10.4.1. cusparse<t>gtsvStridedBatch() ............................................................ 222
  10.4.2. cusparse<t>gtsv2StridedBatch_buffSizeExt() ........................................ 224
  10.4.3. cusparse<t>gtsv2StridedBatch() .......................................................... 226
  10.4.4. cusparse<t>gtsv2StridedBatch_nopivot() ............................................. 228
10.5. batched pentadiagonal solve .................................................................... 231
  10.5.1. cusparse<t>gpsvInterleavedBatch() ....................................................... 232

Chapter 11. cuSPARSE Reorderings Reference ................................................. 235
  11.1. cusparse<t>csrcolor() ............................................................................ 235

Chapter 12. cuSPARSE Format Conversion Reference .................................... 238
  12.1. cusparse<t>bsr2csr() .............................................................................. 239
  12.2. cusparse<t>gebsr2gebsc_buffSize() .......................................................... 242
  12.3. cusparse<t>gebsr2gebsr() .................................................................... 244
  12.4. cusparse<t>gebsr2gebsr_buffSize() .......................................................... 247
  12.5. cusparse<t>gebsr2gebsr() .................................................................... 249
  12.6. cusparse<t>gebsr2csr() ......................................................................... 253
  12.7. cusparse<t>csr2gebsr_buffSize() .............................................................. 256
  12.8. cusparse<t>csr2gebsr() ......................................................................... 258
  12.9. cusparse<t>coo2csr() ............................................................................. 261
  12.10. cusparse<t>csc2dense() ....................................................................... 263
  12.11. cusparse<t>csc2hyb() .......................................................................... 264
  12.12. cusparse<t>csr2bsr() ........................................................................... 266
  12.13. cusparse<t>csr2coo() .......................................................................... 268
  12.14. cusparse<t>csr2csc() .......................................................................... 270
  12.15. cusparseCsr2cscEx() ......................................................................... 271
  12.16. cusparse<t>csr2dense() ...................................................................... 272
  12.17. cusparse<t>csr2csr_compress() ............................................................. 274
  12.18. cusparse<t>csr2hyb() ......................................................................... 278
  12.19. cusparse<t>dense2csc() ...................................................................... 280
  12.20. cusparse<t>dense2csr() ...................................................................... 281
  12.21. cusparse<t>dense2hyb() .................................................................... 283
  12.22. cusparse<t>hyb2csc() ......................................................................... 284
12.23. cusparse<t>hyb2csr()................................................................. 286
12.24. cusparse<t>hyb2dense().............................................................. 287
12.25. cusparse<t>nnz().................................................................... 288
12.26. cusparseCreateIdentityPermutation()............................................. 289
12.27. cusparseXcoosort()..................................................................... 290
12.28. cusparseXcsrsort()..................................................................... 292
12.29. cusparseXcscsort()..................................................................... 294
12.30. cusparseXcsru2csr()................................................................... 297
12.31. cusparseXpruneDense2csr().......................................................... 301
12.32. cusparseXpruneCsr2csr()............................................................ 304
12.33. cusparseXpruneDense2csrPercentage().......................................... 307
12.34. cusparseXpruneCsr2csrByPercentage().......................................... 310
12.35. cusparse<t>nnz_compress()........................................................ 313

Chapter 13. cuSPARSE Generic API Reference.................................................. 315

13.1. Generic Types Reference....................................................................... 315
13.1.1. cudaDataType_t........................................................................ 315
13.1.2. cusparseFormat_t.................................................................... 316
13.1.3. cusparseOrder_t...................................................................... 316
13.1.4. cusparseSpMMAlg_t.................................................................. 316
13.1.5. cusparseIndexType_t................................................................. 316

13.2. Generic Sparse API helper functions..................................................... 317
13.2.1. cusparseCreateCoo().................................................................. 317
13.2.2. cusparseDestroySpMat()............................................................. 318
13.2.3. cusparseSpMatGetNumBatches().................................................. 318
13.2.4. cusparseCooGet()..................................................................... 318
13.2.5. cusparseCooGet()..................................................................... 319
13.2.6. cusparseSpMatGetIndexBase().................................................... 319
13.2.7. cusparseSpMatSetNumBatches()................................................... 319

13.3. Generic Dense API helper functions..................................................... 320
13.3.1. cusparseCreateDnMat()............................................................... 320
13.3.2. cusparseDestroyDnMat()............................................................. 320
13.3.3. cusparseDnMatGet().................................................................. 321
13.3.4. cusparseDnMatGetStridedBatch()................................................ 321
13.3.5. cusparseDnMatSetNumBatches()................................................... 321

13.4. Generic SpMM API functions.............................................................. 322
13.4.1. cusparseSpMM()...................................................................... 322
13.4.2. cusparseSpMM_bufferSize()....................................................... 323

Chapter 14. Appendix A: cuSPARSE Library C++ Example............................... 324

Chapter 15. Appendix B: cuSPARSE Fortran Bindings...................................... 326
15.1. Example B, Fortran Application.......................................................... 328

Chapter 16. Appendix C: Examples of sparse matrix vector multiplication........ 329
16.1. power method............................................................................. 329

Chapter 17. Appendix D: Examples of sorting................................................. 335
17.1. COO sort

Chapter 18. Appendix E: Examples of prune
  18.1. prune dense to sparse
  18.2. prune sparse to sparse
  18.3. prune dense to sparse by percentage
  18.4. prune sparse to sparse by percentage

Chapter 19. Appendix F: Examples of gtsv
  19.1. batched tridiagonal solver

Chapter 20. Appendix G: Examples of gpsv
  20.1. batched penta-diagonal solver

Chapter 21. Appendix H: Examples of csrsm2
  21.1. forward triangular solver

Chapter 22. Appendix I: Acknowledgements

Chapter 23. Bibliography
Chapter 1.
INTRODUCTION

The cuSPARSE library contains a set of basic linear algebra subroutines used for handling sparse matrices. It is implemented on top of the NVIDIA® CUDA™ runtime (which is part of the CUDA Toolkit) and is designed to be called from C and C++. The library routines can be classified into four categories:

- Level 1: operations between a vector in sparse format and a vector in dense format
- Level 2: operations between a matrix in sparse format and a vector in dense format
- Level 3: operations between a matrix in sparse format and a set of vectors in dense format (which can also usually be viewed as a dense tall matrix)
- Conversion: operations that allow conversion between different matrix formats, and compression of csr matrices.

The cuSPARSE library allows developers to access the computational resources of the NVIDIA graphics processing unit (GPU), although it does not auto-parallelize across multiple GPUs. The cuSPARSE API assumes that input and output data reside in GPU (device) memory, unless it is explicitly indicated otherwise by the string DevHostPtr in a function parameter’s name (for example, the parameter *resultDevHostPtr in the function cusparse<t>doti()).

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().

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

1.1. Naming Conventions

The cuSPARSE library functions are available for data types float, double, cuComplex, and cuDoubleComplex. The sparse Level 1, Level 2, and Level 3 functions follow this naming convention:
where \(<t>\) can be \(S\), \(D\), \(C\), \(Z\), or \(X\), corresponding to the data types \(\text{float}\), \(\text{double}\), \(\text{cuComplex}\), \(\text{cuDoubleComplex}\), and the generic type, respectively.

The \(<\text{matrix data format}>\) can be \(\text{dense}\), \(\text{coo}\), \(\text{csr}\), \(\text{csc}\), or \(\text{hyb}\), corresponding to the dense, coordinate, compressed sparse row, compressed sparse column, and hybrid storage formats, respectively.

Finally, the \(<\text{operation}>\) can be \(\text{axpyi}\), \(\text{doti}\), \(\text{dotci}\), \(\text{gthr}\), \(\text{gthrz}\), \(\text{roti}\), or \(\text{sctr}\), corresponding to the Level 1 functions; it also can be \(\text{mv}\) or \(\text{sv}\), corresponding to the Level 2 functions, as well as \(\text{mm}\) or \(\text{sm}\), corresponding to the Level 3 functions.

All of the functions have the return type \(\text{cusparseStatus_t}\) and are explained in more detail in the chapters that follow.

### 1.2. Asynchronous Execution

The cuSPARSE library functions are executed asynchronously with respect to the host and may return control to the application on the host before the result is ready. Developers can use the \(\text{cudaDeviceSynchronize()}\) function to ensure that the execution of a particular cuSPARSE library routine has completed.

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

### 1.3. Static Library support

Starting with release 6.5, the cuSPARSE Library is also delivered in a static form as \text{libcusparse_static.a} on Linux and Mac OSes. The static cuSPARSE library and all others static maths libraries depend on a common thread abstraction layer library called \text{libculibos.a} on Linux and Mac and \text{culibos.lib} on Windows.

For example, on Linux, to compile a small application using cuSPARSE against the dynamic library, the following command can be used:

```
 nvcc myCusparseApp.c -lcusparse -o myCusparseApp
```

Whereas to compile against the static cuSPARSE library, the following command has to be used:

```
 nvcc myCusparseApp.c -lcusparse_static -lculibos -o myCusparseApp
```
It is also possible to use the native Host C++ compiler. Depending on the Host Operating system, some additional libraries like `pthread` or `dl` might be needed on the linking line. The following command on Linux is suggested:

```
g++ myCusparseApp.c -lcusparse_static -lculibos -lcudart_static -lpthread -ldl -I <cuda-toolkit-path>/include -L <cuda-toolkit-path>/lib64 -o myCusparseApp
```

Note that in the latter case, the library `cuda` is not needed. The CUDA Runtime will try to open explicitly the `cuda` library if needed. In the case of a system which does not have the CUDA driver installed, this allows the application to gracefully manage this issue and potentially run if a CPU-only path is available.
Chapter 2.
USING THE CUSPARSE API

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

2.1. Thread Safety
The library is thread safe and its functions can be called from multiple host threads. However, simultaneous read/writes of the same objects (or of the same handle) are not safe. Hence the handle must be private per thread, i.e., only one handle per thread is safe.

2.2. Scalar Parameters
In the cuSPARSE API, the scalar parameters α and β can be passed by reference on the host or the device.

The few functions that return a scalar result, such as \texttt{doti()} and \texttt{nnz()}, return the resulting value by reference on the host or the device. Even though these functions return immediately, similarly to those that return matrix and vector results, the scalar result is not ready until execution of the routine on the GPU completes. This requires proper synchronization be used when reading the result from the host.

This feature allows the cuSPARSE library functions to execute completely asynchronously using streams, even when α and β are generated by a previous kernel. This situation arises, for example, when the library is used to implement iterative methods for the solution of linear systems and eigenvalue problems [3].

2.3. Parallelism with Streams
If the application performs several small independent computations, or if it makes data transfers in parallel with the computation, 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 create CUDA streams using the function `cudaStreamCreate()` and set the stream to be used by each individual cuSPARSE library routine by calling `cusparseSetStream()` just before calling the actual cuSPARSE routine. Then, computations performed in separate streams would 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.

When streams are used, we recommend using the new cuSPARSE API with scalar parameters and results passed by reference in the device memory to achieve maximum computational overlap.

Although a developer can create many streams, in practice it is not possible to have more than 16 concurrent kernels executing at the same time.
The cuSPARSE library supports dense and sparse vector, and dense and sparse matrix formats.

3.1. Index Base Format

The library supports zero- and one-based indexing. The index base is selected through the `cusparseIndexBase_t` type, which is passed as a standalone parameter or as a field in the matrix descriptor `cusparseMatDescr_t` type.

3.2. Vector Formats

This section describes dense and sparse vector formats.

3.2.1. Dense Format

Dense vectors are represented with a single data array that is stored linearly in memory, such as the following 7×1 dense vector.

\[
\begin{bmatrix}
1.0 & 0.0 & 0.0 & 2.0 & 3.0 & 0.0 & 4.0
\end{bmatrix}
\]

(This vector is referenced again in the next section.)

3.2.2. Sparse Format

Sparse vectors are represented with two arrays.

- The data array has the nonzero values from the equivalent array in dense format.
- The integer index array has the positions of the corresponding nonzero values in the equivalent array in dense format.

For example, the dense vector in section 3.2.1 can be stored as a sparse vector with one-based indexing.
It can also be stored as a sparse vector with zero-based indexing.

\[
\begin{bmatrix}
1.0 & 2.0 & 3.0 & 4.0 \\
0 & 3 & 4 & 6
\end{bmatrix}
\]

In each example, the top row is the data array and the bottom row is the index array, and it is assumed that the indices are provided in increasing order and that each index appears only once.

### 3.3. Matrix Formats

Dense and several sparse formats for matrices are discussed in this section.

#### 3.3.1. Dense Format

The dense matrix \( X \) is assumed to be stored in column-major format in memory and is represented by the following parameters.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>( m ) (integer)</td>
<td>The number of rows in the matrix.</td>
</tr>
<tr>
<td>( n ) (integer)</td>
<td>The number of columns in the matrix.</td>
</tr>
<tr>
<td>( \text{ldX} ) (integer)</td>
<td>The leading dimension of ( X ), which must be greater than or equal to ( m ). If ( \text{ldX} ) is greater than ( m ), then ( X ) represents a sub-matrix of a larger matrix stored in memory.</td>
</tr>
<tr>
<td>( X ) (pointer)</td>
<td>Points to the data array containing the matrix elements. It is assumed that enough storage is allocated for ( X ) to hold all of the matrix elements and that cuSPARSE library functions may access values outside of the sub-matrix, but will never overwrite them.</td>
</tr>
</tbody>
</table>

For example, \( m \times n \) dense matrix \( X \) with leading dimension \( \text{ldX} \) can be stored with one-based indexing as shown.

\[
\begin{bmatrix}
X_{1,1} & X_{1,2} & \ldots & X_{1,n} \\
X_{2,1} & X_{2,2} & \ldots & X_{2,n} \\
\vdots & \vdots & \ddots & \vdots \\
X_{m,1} & X_{m,2} & \ldots & X_{mn} \\
\vdots & \vdots & \ddots & \vdots \\
X_{\text{ldX},1} & X_{\text{ldX},2} & \ldots & X_{\text{ldX},n}
\end{bmatrix}
\]

Its elements are arranged linearly in memory in the order below.

\[
\begin{bmatrix}
X_{1,1} & X_{2,1} & \ldots & X_{m,1} & X_{\text{ldX},1} & \ldots & X_{1,n} & X_{2,n} & \ldots & X_{mn} & X_{\text{ldX},n}
\end{bmatrix}
\]

This format and notation are similar to those used in the NVIDIA CUDA cuBLAS library.
3.3.2. Coordinate Format (COO)

The m×n sparse matrix A is represented in COO format by the following parameters.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnz</td>
<td>(integer) The number of nonzero elements in the matrix.</td>
</tr>
<tr>
<td>cooValA</td>
<td>(pointer) Points to the data array of length nnz that holds all nonzero values of A in row-major format.</td>
</tr>
<tr>
<td>cooRowIndA</td>
<td>(pointer) Points to the integer array of length nnz that contains the row indices of the corresponding elements in array cooValA.</td>
</tr>
<tr>
<td>cooColIndA</td>
<td>(pointer) Points to the integer array of length nnz that contains the column indices of the corresponding elements in array cooValA.</td>
</tr>
</tbody>
</table>

A sparse matrix in COO format is assumed to be stored in row-major format: the index arrays are first sorted by row indices and then within the same row by compressed column indices. It is assumed that each pair of row and column indices appears only once.

For example, consider the following 4 × 5 matrix A.

\[
\begin{bmatrix}
1.0 & 4.0 & 0.0 & 0.0 & 0.0 \\
0.0 & 2.0 & 3.0 & 0.0 & 0.0 \\
5.0 & 0.0 & 0.0 & 7.0 & 8.0 \\
10.0 & 0.0 & 9.0 & 0.0 & 6.0
\end{bmatrix}
\]

It is stored in COO format with zero-based indexing this way.

\[
\begin{align*}
\text{cooValA} &= [1.0 \ 4.0 \ 2.0 \ 3.0 \ 5.0 \ 7.0 \ 8.0 \ 9.0 \ 6.0] \\
\text{cooRowIndA} &= [0 \ 0 \ 1 \ 1 \ 2 \ 2 \ 2 \ 3 \ 3] \\
\text{cooColIndA} &= [0 \ 1 \ 1 \ 2 \ 3 \ 4 \ 2 \ 4]
\end{align*}
\]

In the COO format with one-based indexing, it is stored as shown.

\[
\begin{align*}
\text{cooValA} &= [1.0 \ 4.0 \ 2.0 \ 3.0 \ 5.0 \ 7.0 \ 8.0 \ 9.0 \ 6.0] \\
\text{cooRowIndA} &= [1 \ 1 \ 2 \ 2 \ 3 \ 3 \ 4 \ 4] \\
\text{cooColIndA} &= [1 \ 2 \ 2 \ 3 \ 1 \ 4 \ 5 \ 3]
\end{align*}
\]

3.3.3. Compressed Sparse Row Format (CSR)

The only way the CSR differs from the COO format is that the array containing the row indices is compressed in CSR format. The m×n sparse matrix A is represented in CSR format by the following parameters.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnz</td>
<td>(integer) The number of nonzero elements in the matrix.</td>
</tr>
<tr>
<td>csrValA</td>
<td>(pointer) Points to the data array of length nnz that holds all nonzero values of A in row-major format.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>(pointer) Points to the integer array of length m+1 that holds indices into the arrays csrColIndA and csrValA. The first m entries of this array contain the indices of the first nonzero element in the i-th row for i=0,...,m, while the last entry contains nnz+csrRowPtrA(0). In general, csrRowPtrA(0) is 0 or 1 for zero- and one-based indexing, respectively.</td>
</tr>
</tbody>
</table>

www.nvidia.com

cuSPARSE Library

DU-06709-001_v10.1 | 8
Sparse matrices in CSR format are assumed to be stored in row-major CSR format, in other words, the index arrays are first sorted by row indices and then within the same row by column indices. It is assumed that each pair of row and column indices appears only once.

Consider again the $4 \times 5$ matrix $A$.

$$
egin{bmatrix}
1.0 & 4.0 & 0.0 & 0.0 & 0.0 \\
0.0 & 2.0 & 3.0 & 0.0 & 0.0 \\
5.0 & 0.0 & 0.0 & 7.0 & 8.0 \\
0.0 & 0.0 & 9.0 & 0.0 & 6.0
\end{bmatrix}
$$

It is stored in CSR format with zero-based indexing as shown.

$\text{csrValA} = [1.0, 4.0, 2.0, 3.0, 5.0, 7.0, 8.0, 9.0, 6.0]$
$\text{csrRowPtrA} = [0, 2, 4, 7, 9]$
$\text{csrColIndA} = [0, 1, 1, 2, 0, 3, 4, 2, 4]$

This is how it is stored in CSR format with one-based indexing.

$\text{csrValA} = [1.0, 4.0, 2.0, 3.0, 5.0, 7.0, 8.0, 9.0, 6.0]$
$\text{csrRowPtrA} = [1, 3, 5, 8, 10]$
$\text{csrColIndA} = [1, 2, 2, 3, 1, 4, 5, 3, 5]$

### 3.3.4. Compressed Sparse Column Format (CSC)

The CSC format is different from the COO format in two ways: the matrix is stored in column-major format, and the array containing the column indices is compressed in CSC format. The $m \times n$ matrix $A$ is represented in CSC format by the following parameters:

<table>
<thead>
<tr>
<th>nnz</th>
<th>(integer)</th>
<th>The number of nonzero elements in the matrix.</th>
</tr>
</thead>
<tbody>
<tr>
<td>cscValA</td>
<td>(pointer)</td>
<td>Points to the data array of length $\text{nnz}$ that holds all nonzero values of $A$ in column-major format.</td>
</tr>
<tr>
<td>cscRowIndA</td>
<td>(pointer)</td>
<td>Points to the integer array of length $\text{nnz}$ that contains the row indices of the corresponding elements in array $\text{cscValA}$.</td>
</tr>
<tr>
<td>cscColPtrA</td>
<td>(pointer)</td>
<td>Points to the integer array of length $n+1$ that holds indices into the arrays $\text{cscRowIndA}$ and $\text{cscValA}$. The first $n$ entries of this array contain the indices of the first nonzero element in the $i$th row for $i=1, \ldots, n$, while the last entry contains $\text{nnz} + \text{cscColPtrA}(0)$. In general, $\text{cscColPtrA}(0)$ is 0 or 1 for zero- and one-based indexing, respectively.</td>
</tr>
</tbody>
</table>

The matrix $A$ in CSR format has exactly the same memory layout as its transpose in CSC format (and vice versa).

For example, consider once again the $4 \times 5$ matrix $A$. 
It is stored in CSC format with zero-based indexing this way.

\[
\text{cscValA} = \begin{bmatrix} 1.0 & 5.0 & 4.0 & 2.0 & 3.0 & 9.0 & 7.0 & 8.0 & 6.0 \end{bmatrix} \\
\text{cscRowIndA} = [0 \ 2 \ 0 \ 1 \ 1 \ 3 \ 2 \ 2 \ 3] \\
\text{cscColPtrA} = [0 \ 2 \ 4 \ 6 \ 7 \ 9]
\]

In CSC format with one-based indexing, this is how it is stored.

\[
\text{cscValA} = \begin{bmatrix} 1.0 & 5.0 & 4.0 & 2.0 & 3.0 & 9.0 & 7.0 & 8.0 & 6.0 \end{bmatrix} \\
\text{cscRowIndA} = [1 \ 3 \ 1 \ 2 \ 4 \ 3 \ 3 \ 4] \\
\text{cscColPtrA} = [1 \ 3 \ 5 \ 7 \ 8 \ 10]
\]

Each pair of row and column indices appears only once.

### 3.3.5. Ellpack-Itpack Format (ELL)

An \( m \times n \) sparse matrix \( \mathbf{A} \) with at most \( k \) nonzero elements per row is stored in the Ellpack-Itpack (ELL) format \([2]\) using two dense arrays of dimension \( m \times k \). The first data array contains the values of the nonzero elements in the matrix, while the second integer array contains the corresponding column indices.

For example, consider the \( 4 \times 5 \) matrix \( \mathbf{A} \).

\[
\begin{bmatrix}
1.0 & 4.0 & 0.0 & 0.0 & 0.0 \\
0.0 & 2.0 & 3.0 & 0.0 & 0.0 \\
5.0 & 0.0 & 0.0 & 7.0 & 8.0 \\
0.0 & 0.0 & 9.0 & 0.0 & 6.0
\end{bmatrix}
\]

This is how it is stored in ELL format with zero-based indexing.

\[
\begin{align*}
\text{data} & = \begin{bmatrix} 1.0 & 4.0 & 0.0 \\ 2.0 & 3.0 & 0.0 \\ 5.0 & 7.0 & 8.0 \\ 9.0 & 6.0 & 0.0 \end{bmatrix} \\
\text{indices} & = \begin{bmatrix} 0 & 1 & -1 \\ 1 & 2 & -1 \\ 0 & 3 & 4 \\ 2 & 4 & -1 \end{bmatrix}
\end{align*}
\]

It is stored this way in ELL format with one-based indexing.

\[
\begin{align*}
\text{data} & = \begin{bmatrix} 1.0 & 4.0 & 0.0 \\ 2.0 & 3.0 & 0.0 \\ 5.0 & 7.0 & 8.0 \\ 9.0 & 6.0 & 0.0 \end{bmatrix} \\
\text{indices} & = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 3 & -1 \\ 1 & 4 & 5 \\ 3 & 5 & -1 \end{bmatrix}
\end{align*}
\]
Sparse matrices in ELL format are assumed to be stored in column-major format in memory. Also, rows with less than \( k \) nonzero elements are padded in the data and indices arrays with zero and \(-1\), respectively.

The ELL format is not supported directly, but it is used to store the regular part of the matrix in the HYB format that is described in the next section.

### 3.3.6. Hybrid Format (HYB)

The HYB sparse storage format is composed of a regular part, usually stored in ELL format, and an irregular part, usually stored in COO format [1]. The ELL and COO parts are always stored using zero-based indexing. HYB is implemented as an opaque data format that requires the use of a conversion operation to store a matrix in it. The conversion operation partitions the general matrix into the regular and irregular parts automatically or according to developer-specified criteria.

For more information, please refer to the description of `cusparseHybPartition_t` type, as well as the description of the conversion routines `dense2hyb`, `csc2hyb` and `csr2hyb`.

### 3.3.7. Block Compressed Sparse Row Format (BSR)

The only difference between the CSR and BSR formats is the format of the storage element. The former stores primitive data types (`single`, `double`, `cuComplex`, and `cuDoubleComplex`) whereas the latter stores a two-dimensional square block of primitive data types. The dimension of the square block is `blockDim`. The \( m \times n \) sparse matrix \( A \) is equivalent to a block sparse matrix \( A_b \) with \( mb = \frac{m + blockDim - 1}{blockDim} \) block rows and \( nb = \frac{n + blockDim - 1}{blockDim} \) block columns. If \( m \) or \( n \) is not multiple of `blockDim`, then zeros are filled into \( A_b \).

\( A \) is represented in BSR format by the following parameters.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Type</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>blockDim</td>
<td>(integer)</td>
<td>Block dimension of matrix ( A ).</td>
</tr>
<tr>
<td>mb</td>
<td>(integer)</td>
<td>The number of block rows of ( A ).</td>
</tr>
<tr>
<td>nb</td>
<td>(integer)</td>
<td>The number of block columns of ( A ).</td>
</tr>
<tr>
<td>nnzb</td>
<td>(integer)</td>
<td>The number of nonzero blocks in the matrix.</td>
</tr>
<tr>
<td>bsrValA</td>
<td>(pointer)</td>
<td>Points to the data array of length ( nnzb \times blockDim^2 ) that holds all elements of nonzero blocks of ( A ). The block elements are stored in either column-major order or row-major order.</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>(pointer)</td>
<td>Points to the integer array of length ( mb+1 ) that holds indices into the arrays <code>bsrColIndA</code> and <code>bsrValA</code>. The first ( mb ) entries of this array contain the indices of the first nonzero block in the ( i )th block row for ( i=1,\ldots,mb ), while the last entry contains ( nnzb+bsrRowPtrA(0) ). In general, <code>bsrRowPtrA(0)</code> is 0 or 1 for zero- and one-based indexing, respectively.</td>
</tr>
<tr>
<td>bsrColIndA</td>
<td>(pointer)</td>
<td>Points to the integer array of length ( nnzb ) that contains the column indices of the corresponding blocks in array <code>bsrValA</code>.</td>
</tr>
</tbody>
</table>
As with CSR format, (row, column) indices of BSR are stored in row-major order. The index arrays are first sorted by row indices and then within the same row by column indices.

For example, consider again the $4 \times 5$ matrix $A$.

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

If $\text{blockDim}$ is equal to 2, then $mb$ is 2, $nb$ is 3, and matrix $A$ is split into $2 \times 3$ block matrix $A_B$. The dimension of $A_B$ is $4 \times 6$, slightly bigger than matrix $A$, so zeros are filled in the last column of $A_B$. The element-wise view of $A_B$ is this.

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

Based on zero-based indexing, the block-wise view of $A_B$ can be represented as follows.

$$
A_B = \begin{bmatrix}
A_{00} & A_{01} & A_{02} \\
A_{10} & A_{11} & A_{12}
\end{bmatrix}
$$

The basic element of BSR is a nonzero $A_{ij}$ block, one that contains at least one nonzero element of $A$. Five of six blocks are nonzero in $A_B$.

$$
A_{00} = \begin{bmatrix} 1 & 4 \\ 2 & \end{bmatrix}, A_{01} = \begin{bmatrix} 0 & 0 \\ 3 & 0 \end{bmatrix}, A_{10} = \begin{bmatrix} 5 & 0 \\ 0 & 0 \end{bmatrix}, A_{11} = \begin{bmatrix} 0 & 7 \\ 9 & 0 \end{bmatrix}, A_{12} = \begin{bmatrix} 8 & 0 \\ \end{bmatrix}
$$

BSR format only stores the information of nonzero blocks, including block indices $(i, j)$ and values $A_{ij}$. Also row indices are compressed in CSR format.

- $\text{bsrValA} = [A_{00}, A_{01}, A_{10}, A_{11}, A_{12}]$
- $\text{bsrRowPtrA} = [0, 2, 5]$
- $\text{bsrColIndA} = [0, 1, 0, 1, 2]$

There are two ways to arrange the data element of block $A_{ij}$: row-major order and column-major order. Under column-major order, the physical storage of $\text{bsrValA}$ is this.

$bssValA=[1 \ 0 \ 4 \ 2 \ 0 \ 3 \ 0 \ 0 \ | \ 5 \ 0 \ 0 \ 0 \ | \ 0 \ 9 \ 7 \ 0 \ | \ 8 \ 6 \ 0 \ 0 ]$

Under row-major order, the physical storage of $\text{bsrValA}$ is this.

$bssValA=[1 \ 4 \ 0 \ 2 \ 0 \ 0 \ 3 \ 0 \ | \ 5 \ 0 \ 0 \ 0 \ | \ 0 \ 7 \ 9 \ 0 \ | \ 8 \ 0 \ 6 \ 0 ]$

Similarly, in BSR format with one-based indexing and column-major order, $A$ can be represented by the following.

$$
A_B = \begin{bmatrix}
A_{11} & A_{12} & A_{13} \\
A_{21} & A_{22} & A_{23}
\end{bmatrix}
$$

$bssValA=[1 \ 0 \ 4 \ 2 \ 0 \ 3 \ 0 \ 0 \ | \ 5 \ 0 \ 0 \ 0 \ | \ 0 \ 9 \ 7 \ 0 \ | \ 8 \ 6 \ 0 \ 0 ]$
The general BSR format has two parameters, rowBlockDim and colBlockDim. rowBlockDim is number of rows within a block and colBlockDim is number of columns within a block. If rowBlockDim=colBlockDim, general BSR format is the same as BSR format. If rowBlockDim=colBlockDim=1, general BSR format is the same as CSR format. The conversion routine gebsr2gebsr is used to do conversion among CSR, BSR and general BSR.

In the cuSPARSE Library, the storage format of blocks in BSR format can be column-major or row-major, independently of the base index. However, if the developer uses BSR format from the Math Kernel Library (MKL) and wants to directly interface with the cuSPARSE Library, then cusparseDirection_t CUSPARSE_DIRECTION_COLUMN should be used if the base index is one; otherwise, cusparseDirection_t CUSPARSE_DIRECTION_ROW should be used.

### 3.3.8. Extended BSR Format (BSRX)

BSRX is the same as the BSR format, but the array bsrRowPtrA is separated into two parts. The first nonzero block of each row is still specified by the array bsrRowPtrA, which is the same as in BSR, but the position next to the last nonzero block of each row is specified by the array bsrEndPtrA. Briefly, BSRX format is simply like a 4-vector variant of BSR format.

Matrix $A$ is represented in BSRX format by the following parameters.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>blockDim</td>
<td>Block dimension of matrix $A$.</td>
</tr>
<tr>
<td>mb</td>
<td>(integer) The number of block rows of $A$.</td>
</tr>
<tr>
<td>nb</td>
<td>(integer) The number of block columns of $A$.</td>
</tr>
<tr>
<td>nnzb</td>
<td>(integer) number of nonzero blocks in the matrix $A$.</td>
</tr>
<tr>
<td>bsrValA</td>
<td>(pointer) Points to the data array of length $nnzb \times blockDim^{2}$ that holds all the elements of the nonzero blocks of $A$. The block elements are stored in either column-major order or row-major order.</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>(pointer) Points to the integer array of length $mb$ that holds indices into the arrays bsrColIndA and bsrValA; bsrRowPtrA(i) is the position of the first nonzero block of the i-th block row in bsrColIndA and bsrValA.</td>
</tr>
<tr>
<td>bsrEndPtrA</td>
<td>(pointer) Points to the integer array of length $mb$ that holds indices into the arrays bsrColIndA and bsrValA; bsrRowPtrA(i) is the position next to the last nonzero block of the i-th block row in bsrColIndA and bsrValA.</td>
</tr>
<tr>
<td>bsrColIndA</td>
<td>(pointer) Points to the integer array of length $nnzb$ that contains the column indices of the corresponding blocks in array bsrValA.</td>
</tr>
</tbody>
</table>

A simple conversion between BSR and BSRX can be done as follows. Suppose the developer has a $2 \times 3$ block sparse matrix $A_b$ represented as shown.

$$
A_b = \begin{bmatrix}
A_{00} & A_{01} & A_{02} \\
A_{10} & A_{11} & A_{12} 
\end{bmatrix}
$$
Assume it has this BSR format.

\[ \begin{align*}
\text{bsrValA of BSR} &= [A_{00} \ A_{01} \ A_{10} \ A_{11} \ A_{12}] \\
\text{bsrRowPtrA of BSR} &= [0 \ 2 \ 5] \\
\text{bsrColIndA of BSR} &= [0 \ 1 \ 0 \ 1 \ 2]
\end{align*} \]

The `bsrRowPtrA` of the BSRX format is simply the first two elements of the `bsrRowPtrA` BSR format. The `bsrEndPtrA` of BSRX format is the last two elements of the `bsrRowPtrA` of BSR format.

\[ \begin{align*}
\text{bsrRowPtrA of BSRX} &= [0 \ 2] \\
\text{bsrEndPtrA of BSRX} &= [2 \ 5]
\end{align*} \]

The advantage of the BSRX format is that the developer can specify a submatrix in the original BSR format by modifying `bsrRowPtrA` and `bsrEndPtrA` while keeping `bsrColIndA` and `bsrValA` unchanged.

For example, to create another block matrix \( \hat{A} = \begin{bmatrix} O & O & O \\ O & A_{11} & O \end{bmatrix} \) that is slightly different from \( A \), the developer can keep `bsrColIndA` and `bsrValA`, but reconstruct \( \hat{A} \) by properly setting of `bsrRowPtrA` and `bsrEndPtrA`. The following 4-vector characterizes \( \hat{A} \).

\[ \begin{align*}
\text{bsrValA of } \hat{A} &= [A_{00} \ A_{01} \ A_{10} \ A_{11} \ A_{12}] \\
\text{bsrColIndA of } \hat{A} &= [0 \ 1 \ 0 \ 1 \ 2] \\
\text{bsrRowPtrA of } \hat{A} &= [0 \ 3] \\
\text{bsrEndPtrA of } \hat{A} &= [0 \ 4]
\end{align*} \]
Chapter 4.
CUSPARSE TYPES REFERENCE

4.1. Data 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.

4.2. cusparseAction_t

This type indicates whether the operation is performed only on indices or on data and indices.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_ACTION_SYMBOLIC</td>
<td>the operation is performed only on indices.</td>
</tr>
<tr>
<td>CUSPARSE_ACTION_NUMERIC</td>
<td>the operation is performed on data and indices.</td>
</tr>
</tbody>
</table>

4.3. cusparseDirection_t

This type indicates whether the elements of a dense matrix should be parsed by rows or by columns (assuming column-major storage in memory of the dense matrix) in function cusparse[SD|DC|Z]nnz. Besides storage format of blocks in BSR format is also controlled by this type.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_DIRECTION_ROW</td>
<td>the matrix should be parsed by rows.</td>
</tr>
<tr>
<td>CUSPARSE_DIRECTION_COLUMN</td>
<td>the matrix should be parsed by columns.</td>
</tr>
</tbody>
</table>
### 4.4. cusparseHandle_t

This is a pointer type to an opaque cuSPARSE context, which the user must initialize by calling prior to calling `cusparseCreate()` any other library function. The handle created and returned by `cusparseCreate()` must be passed to every cuSPARSE function.

### 4.5. cusparseHybMat_t

This is a pointer type to an opaque structure holding the matrix in HYB format, which is created by `cusparseCreateHybMat` and destroyed by `cusparseDestroyHybMat`.

#### 4.5.1. cusparseHybPartition_t

This type indicates how to perform the partitioning of the matrix into regular (ELL) and irregular (COO) parts of the HYB format.

The partitioning is performed during the conversion of the matrix from a dense or sparse format into the HYB format and is governed by the following rules. When `CUSPARSE_HYB_PARTITION_AUTO` is selected, the cuSPARSE library automatically decides how much data to put into the regular and irregular parts of the HYB format. When `CUSPARSE_HYB_PARTITION_USER` is selected, the width of the regular part of the HYB format should be specified by the caller. When `CUSPARSE_HYB_PARTITION_MAX` is selected, the width of the regular part of the HYB format equals to the maximum number of non-zero elements per row, in other words, the entire matrix is stored in the regular part of the HYB format.

The default is to let the library automatically decide how to split the data.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_HYB_PARTITION_AUTO</code></td>
<td>the automatic partitioning is selected (default).</td>
</tr>
<tr>
<td><code>CUSPARSE_HYB_PARTITION_USER</code></td>
<td>the user specified threshold is used.</td>
</tr>
<tr>
<td><code>CUSPARSE_HYB_PARTITION_MAX</code></td>
<td>the data is stored in ELL format.</td>
</tr>
</tbody>
</table>

### 4.6. cusparseMatDescr_t

This structure is used to describe the shape and properties of a matrix.

```c
typedef struct {
    cusparseMatrixType_t MatrixType;
    cusparseFillMode_t FillMode;
    cusparseDiagType_t DiagType;
    cusparseIndexBase_t IndexBase;
} cusparseMatDescr_t;
```
4.6.1. cusparseDiagType_t
This type indicates if the matrix diagonal entries are unity. The diagonal elements are always assumed to be present, but if CUSPARSE_DIAG_TYPE_UNIT is passed to an API routine, then the routine assumes that all diagonal entries are unity and will not read or modify those entries. Note that in this case the routine assumes the diagonal entries are equal to one, regardless of what those entries are actually set to in memory.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_DIAG_TYPE_NON_UNIT</td>
<td>the matrix diagonal has non-unit elements.</td>
</tr>
<tr>
<td>CUSPARSE_DIAG_TYPE_UNIT</td>
<td>the matrix diagonal has unit elements.</td>
</tr>
</tbody>
</table>

4.6.2. cusparseFillMode_t
This type indicates if the lower or upper part of a matrix is stored in sparse storage.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_FILL_MODE_LOWER</td>
<td>the lower triangular part is stored.</td>
</tr>
<tr>
<td>CUSPARSE_FILL_MODE_UPPER</td>
<td>the upper triangular part is stored.</td>
</tr>
</tbody>
</table>

4.6.3. cusparseIndexBase_t
This type indicates if the base of the matrix indices is zero or one.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_INDEX_BASE_ZERO</td>
<td>the base index is zero.</td>
</tr>
<tr>
<td>CUSPARSE_INDEX_BASE_ONE</td>
<td>the base index is one.</td>
</tr>
</tbody>
</table>

4.6.4. cusparseMatrixType_t
This type indicates the type of matrix stored in sparse storage. Notice that for symmetric, Hermitian and triangular matrices only their lower or upper part is assumed to be stored.

The whole idea of matrix type and fill mode is to keep minimum storage for symmetric/Hermitian matrix, and also to take advantage of symmetric property on SpMV (Sparse Matrix Vector multiplication). To compute \( y = A^*x \) when \( A \) is symmetric and only lower triangular part is stored, two steps are needed. First step is to compute \( y = (L+D)^*x \) and second step is to compute \( y = L^T * x + y \). Given the fact that the transpose operation \( y = L^T * x \) is 10x slower than non-transpose version \( y = L * x \), the symmetric property does not show up any performance gain. It is better for the user to extend the symmetric matrix to a general matrix and apply \( y = A^*x \) with matrix type CUSPARSE_MATRIX_TYPE_GENERAL.

In general, SpMV, preconditioners (incomplete Cholesky or incomplete LU) and triangular solver are combined together in iterative solvers, for example PCG and
GMRES. If the user always uses general matrix (instead of symmetric matrix), there is no need to support other than general matrix in preconditioners. Therefore the new routines, [bsr|csr]sv2 (triangular solver), [bsr|csr]ilu02 (incomplete LU) and [bsr|csr]ic02 (incomplete Cholesky), only support matrix type CUSPARSE_MATRIX_TYPE_GENERAL.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_MATRIX_TYPE_GENERAL</td>
<td>the matrix is general.</td>
</tr>
<tr>
<td>CUSPARSE_MATRIX_TYPE_SYMMETRIC</td>
<td>the matrix is symmetric.</td>
</tr>
<tr>
<td>CUSPARSE_MATRIX_TYPE_HERMITIAN</td>
<td>the matrix is Hermitian.</td>
</tr>
<tr>
<td>CUSPARSE_MATRIX_TYPE_TRIANGULAR</td>
<td>the matrix is triangular.</td>
</tr>
</tbody>
</table>

4.7. cusparseOperation_t

This type indicates which operations need to be performed with the sparse matrix.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_OPERATION_NON_TRANSPOSE</td>
<td>the non-transpose operation is selected.</td>
</tr>
<tr>
<td>CUSPARSE_OPERATION_TRANSPOSE</td>
<td>the transpose operation is selected.</td>
</tr>
<tr>
<td>CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE</td>
<td>the conjugate transpose operation is selected.</td>
</tr>
</tbody>
</table>

4.8. cusparsePointerMode_t

This type indicates whether the scalar values are passed by reference on the host or device. It is important to point out that if several scalar values are passed by reference in the function call, all of them will conform to the same single pointer mode. The pointer mode can be set and retrieved using cusparseSetPointerMode() and cusparseGetPointerMode() routines, respectively.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_POINTER_MODE_HOST</td>
<td>the scalars are passed by reference on the host.</td>
</tr>
<tr>
<td>CUSPARSE_POINTER_MODE_DEVICE</td>
<td>the scalars are passed by reference on the device.</td>
</tr>
</tbody>
</table>

4.9. cusparseAlgMode_t

This is type for algorithm parameter to cusparseCsrmvEx() and cusparseCsrmvEx_bufferSize() functions. Note that previously defined values CUSPARSE_ALGO (for naive algorithm) and CUSPARSE_ALG1 (for merge path algorithm) are deprecated and replaced by named aliases specified below.

www.nvidia.com
cuSPARSE Library
<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_ALG_NAIVE</td>
<td>Use default naive algorithm.</td>
</tr>
<tr>
<td>CUSPARSE_ALG_MERGE_PATH</td>
<td>Use load-balancing algorithm that suits better for irregular nonzero-patterns.</td>
</tr>
</tbody>
</table>

### 4.10. cusparseSolvePolicy_t

This type indicates whether level information is generated and used in `csrsv2`, `csric02`, `csrilu02`, `bsrsv2`, `bsric02` and `bsrilu02`.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_SOLVE_POLICY_NO_LEVEL</td>
<td>no level information is generated and used.</td>
</tr>
<tr>
<td>CUSPARSE_SOLVE_POLICY_USE_LEVEL</td>
<td>generate and use level information.</td>
</tr>
</tbody>
</table>

### 4.11. cusparseSolveAnalysisInfo_t

This is a pointer type to an opaque structure holding the information collected in the analysis phase of the solution of the sparse triangular linear system. It is expected to be passed unchanged to the solution phase of the sparse triangular linear system.

### 4.12. cusparseSolveAnalysisInfo_t

This is a pointer type to an opaque structure holding the information collected in the analysis phase of the solution of the sparse triangular linear system. It is expected to be passed unchanged to the solution phase of the sparse triangular linear system.

### 4.13. csrsv2Info_t

This is a pointer type to an opaque structure holding the information used in `csrsv2_bufferSize()`, `csrsv2_analysis()`, and `csrsv2_solve()`.

### 4.14. csrsm2Info_t

This is a pointer type to an opaque structure holding the information used in `csrsm2_bufferSize()`, `csrsm2_analysis()`, and `csrsm2_solve()`.

### 4.15. csric02Info_t

This is a pointer type to an opaque structure holding the information used in `csric02_bufferSize()`, `csric02_analysis()`, and `csric02()`.
4.16. csrcilu02Info_t
This is a pointer type to an opaque structure holding the information used in
csrcilu02_bufferSize(), csrcilu02_analysis(), and csrcilu02().

4.17. bsrsv2Info_t
This is a pointer type to an opaque structure holding the information used in
bsrsv2_bufferSize(), bsrsv2_analysis(), and bsrsv2_solve().

4.18. bsrsm2Info_t
This is a pointer type to an opaque structure holding the information used in
bsrsm2_bufferSize(), bsrsm2_analysis(), and bsrsm2_solve().

4.19. bsric02Info_t
This is a pointer type to an opaque structure holding the information used in
bsric02_bufferSize(), bsric02_analysis(), and bsric02().

4.20. bsrilu02Info_t
This is a pointer type to an opaque structure holding the information used in
bsrilu02_bufferSize(), bsrilu02_analysis(), and bsrilu02().

4.21. csrgemm2Info_t
This is a pointer type to an opaque structure holding the information used in
csrgemm2_bufferSizeExt(), and csrgemm2().

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

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>The operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INIT</td>
<td>The cuSPARSE 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 cuSPARSE routine, or an error in the hardware setup.</td>
</tr>
</tbody>
</table>
CuSPARSE Types Reference

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
<th>Correction</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>Resource allocation failed inside the cuSPARSE library. This is usually caused by a cudaMalloc() failure.</td>
<td>To correct: prior to the function call, deallocate previously allocated memory as much as possible.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>An unsupported value or parameter was passed to the function (a negative vector size, for example).</td>
<td>To correct: ensure that all the parameters being passed have valid values.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>The function requires a feature absent from the device architecture; usually caused by the lack of support for atomic operations or double precision.</td>
<td>To correct: compile and run the application on a device with appropriate compute capability, which is 1.1 for 32-bit atomic operations and 1.3 for double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MAPPING_ERROR</td>
<td>An access to GPU memory space failed, which is usually caused by a failure to bind a texture.</td>
<td>To correct: prior to the function call, unbind any previously bound textures.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>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.</td>
<td>To correct: check that the hardware, an appropriate version of the driver, and the cuSPARSE library are correctly installed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL</td>
<td>An internal cuSPARSE operation failed. This error is usually caused by a cudaMemcpyAsync() failure.</td>
<td>To correct: check that the hardware, an appropriate version of the driver, and the cuSPARSE 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.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>The matrix type is not supported by this function. This is usually caused by passing an invalid matrix descriptor to the function.</td>
<td>To correct: check that the fields in cusparseMatDescr_t descrA were set correctly.</td>
</tr>
</tbody>
</table>
The cuSPARSE helper functions are described in this section.

### 5.1. cusparseCreate()

```c
 cusparseStatus_t
 cusparseCreate(cusparseHandle_t *handle)
```

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

**Output**

| handle | the pointer to the handle to the cuSPARSE context. |

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the initialization succeeded.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the CUDA Runtime initialization failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device compute capability (CC) is less than 1.1. The CC of at least 1.1 is required.</td>
</tr>
</tbody>
</table>

### 5.2. cusparseCreateSolveAnalysisInfo()

```c
 cusparseStatus_t
 cusparseCreateSolveAnalysisInfo(cusparseSolveAnalysisInfo_t *info)
```

This function creates and initializes the solve and analysis structure to default values.

**Input**

| info | the pointer to the solve and analysis structure. |

**Status Returned**
<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the structure was initialized successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.3. cusparseCreateHybMat()

cusparseStatus_t
cusparseCreateHybMat(cusparseHybMat_t *hybA)

This function creates and initializes the hybA opaque data structure.

Input

| hybA | the pointer to the hybrid format storage structure. |

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the structure was initialized successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.4. cusparseCreateMatDescr()

cusparseStatus_t
cusparseCreateMatDescr(cusparseMatDescr_t *descrA)

This function initializes the matrix descriptor. It sets the fields MatrixType and IndexBase to the default values CUSPARSE_MATRIX_TYPE_GENERAL and CUSPARSE_INDEX_BASE_ZERO, respectively, while leaving other fields uninitialized.

Input

| descrA | the pointer to the matrix descriptor. |

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the descriptor was initialized successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.5. cusparseCreateSolveAnalysisInfo()

cusparseStatus_t
cusparseCreateSolveAnalysisInfo(cusparseSolveAnalysisInfo_t *info)

This function creates and initializes the solve and analysis structure to default values.

Input

| info | the pointer to the solve and analysis structure. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the structure was initialized successfully. |
5.6. cusparseDestroy()

cusparseStatus_t
cusparseDestroy(cusparseHandle_t handle)

This function releases CPU-side resources used by the cuSPARSE library. The release of GPU-side resources may be deferred until the application shuts down.

Input

| handle | the handle to the cuSPARSE context. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the shutdown succeeded. |
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized. |

5.7. cusparseDestroySolveAnalysisInfo()

cusparseStatus_t
cusparseDestroySolveAnalysisInfo(cusparseSolveAnalysisInfo_t info)

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

Input

| info | the solve and analysis structure. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully |

5.8. cusparseDestroyHybMat()

cusparseStatus_t
cusparseDestroyHybMat(cusparseHybMat_t hybA)

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

Input

| hybA | the hybrid format storage structure. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully. |
5.9. cusparseDestroyMatDescr()

```c
cusparseStatus_t
 cusparseDestroyMatDescr(cusparseMatDescr_t descrA)
```

This function releases the memory allocated for the matrix descriptor.

**Input**

| descrA | the matrix descriptor. |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully. |

5.10. cusparseDestroySolveAnalysisInfo()

```c
cusparseStatus_t
 cusparseDestroySolveAnalysisInfo(cusparseSolveAnalysisInfo_t info)
```

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

**Input**

| info | the solve and analysis structure. |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully. |

5.11. cusparseGetLevelInfo()

```c
cusparseStatus_t
 cusparseGetLevelInfo(cusparseHandle_t handle,
                      cusparseSolveAnalysisInfo_t info,
                      int *nlevels,
                      int **levelPtr,
                      int **levelInd)
```

This function returns the number of levels and the assignment of rows into the levels computed by either the csrsv_analysis, csrsm_analysis or hybsv_analysis routines.

**Input**

| handle | handle to the cuSPARSE library context. |
| info | the pointer to the solve and analysis structure. |

**Output**

<p>| nlevels | number of levels. |</p>
<table>
<thead>
<tr>
<th>levelPtr</th>
<th>integer array of nlevels+1 elements that contains the start of every level and the end of the last level plus one.</th>
</tr>
</thead>
<tbody>
<tr>
<td>levelInd</td>
<td>integer array of m (number of rows in the matrix) elements that contains the row indices belonging to every level.</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the structure was initialized successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library or the solve analysis structure was not initialized.</td>
</tr>
</tbody>
</table>

### 5.12. cusparseGetMatDiagType()

```c
 cusparseDiagType_t
 cusparseGetMatDiagType(const cusparseMatDescr_t descrA)
```

This function returns the `DiagType` field of the matrix descriptor `descrA`.

**Input**

- `descrA` the matrix descriptor.

**Returned**

- One of the enumerated diagType types.

### 5.13. cusparseGetMatFillMode()

```c
 cusparseFillMode_t
 cusparseGetMatFillMode(const cusparseMatDescr_t descrA)
```

This function returns the `FillMode` field of the matrix descriptor `descrA`.

**Input**

- `descrA` the matrix descriptor.

**Returned**

- One of the enumerated fillMode types.

### 5.14. cusparseGetMatIndexBase()

```c
 cusparseIndexBase_t
 cusparseGetMatIndexBase(const cusparseMatDescr_t descrA)
```

This function returns the `IndexBase` field of the matrix descriptor `descrA`.

**Input**

- `descrA` the matrix descriptor.
5.15. cusparseGetMatType()

```c
 cusparseMatrixType_t
 cusparseGetMatType(const cusparseMatDescr_t descrA)
```

This function returns the `MatrixType` field of the matrix descriptor `descrA`.

**Input**
- `descrA` the matrix descriptor.

**Returned**
- One of the enumerated matrix types.

5.16. cusparseGetPointerMode()

```c
 cusparseStatus_t
 cusparseGetPointerMode(cusparseHandlet handle, cusparsePointerMode_t *mode)
```

This function obtains the pointer mode used by the cuSPARSE library. Please see the section on the `cusparsePointerMode_t` type for more details.

**Input**
- `handle` the handle to the cuSPARSE context.

**Output**
- `mode` One of the enumerated pointer mode types.

**Status Returned**
- `CUSPARSE_STATUS_SUCCESS` the pointer mode was returned successfully.
- `CUSPARSE_STATUS_NOT_INITIALIZED` the library was not initialized.

5.17. cusparseGetVersion()

```c
 cusparseStatus_t
 cusparseGetVersion(cusparseHandle_t handle, int *version)
```

This function returns the version number of the cuSPARSE library.

**Input**
- `handle` the handle to the cuSPARSE context.

**Output**
<table>
<thead>
<tr>
<th>version</th>
<th>the version number of the library.</th>
</tr>
</thead>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the version was returned successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
</tbody>
</table>

### 5.18. cusparseSetMatDiagType()

```c
#include <cusparse.h>

cusparseStatus_t cusparseSetMatDiagType(cusparseMatDescr_t descrA, cusparseDiagType_t diagType)
```

This function sets the **DiagType** field of the matrix descriptor `descrA`.

**Input**

<table>
<thead>
<tr>
<th>diagType</th>
<th>One of the enumerated diagType types.</th>
</tr>
</thead>
</table>

**Output**

<table>
<thead>
<tr>
<th>descrA</th>
<th>the matrix descriptor.</th>
</tr>
</thead>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the field <strong>DiagType</strong> was set successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>An invalid <strong>diagType</strong> parameter was passed.</td>
</tr>
</tbody>
</table>

### 5.19. cusparseSetMatFillMode()

```c
#include <cusparse.h>

cusparseStatus_t cusparseSetMatFillMode(cusparseMatDescr_t descrA, cusparseFillMode_t fillMode)
```

This function sets the **FillMode** field of the matrix descriptor `descrA`.

**Input**

<table>
<thead>
<tr>
<th>fillMode</th>
<th>One of the enumerated fillMode types.</th>
</tr>
</thead>
</table>

**Output**

<table>
<thead>
<tr>
<th>descrA</th>
<th>the matrix descriptor.</th>
</tr>
</thead>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the <strong>FillMode</strong> field was set successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>An invalid <strong>fillMode</strong> parameter was passed.</td>
</tr>
</tbody>
</table>
5.20. `cusparseSetMatIndexBase()`

```c
cusparseStatus_t
cusparseSetMatIndexBase(cusparseMatDescr_t descrA, 
cusparseIndexBase_t base)
```

This function sets the `IndexBase` field of the matrix descriptor `descrA`.

**Input**

| base               | One of the enumerated indexBase types. |

**Output**

| descrA              | the matrix descriptor. |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the <code>IndexBase</code> field was set successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>An invalid <code>base</code> parameter was passed.</td>
</tr>
</tbody>
</table>

5.21. `cusparseSetMatType()`

```c
cusparseStatus_t
cusparseSetMatType(cusparseMatDescr_t descrA, cusparseMatrixType_t type)
```

This function sets the `MatrixType` field of the matrix descriptor `descrA`.

**Input**

| type               | One of the enumerated matrix types. |

**Output**

| descrA              | the matrix descriptor. |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the <code>MatrixType</code> field was set successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>An invalid <code>type</code> parameter was passed.</td>
</tr>
</tbody>
</table>

5.22. `cusparseSetPointerMode()`

```c
cusparseStatus_t
cusparseSetPointerMode(cusparseHandle_t handle, 
cusparsePointerMode_t mode)
```

This function sets the pointer mode used by the cuSparse library. The default is for the values to be passed by reference on the host. Please see the section on the `cublasPointerMode_t` type for more details.

**Input**
handle | the handle to the cuSPARSE context.
mode | One of the enumerated pointer mode types.

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the pointer mode was set successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the library was not initialized.</td>
</tr>
</tbody>
</table>

### 5.23. cusparseSetStream()

cusparseStatus_t
cusparseSetStream(cusparseHandle_t handle, cudaStream_t streamId)

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

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>the handle to the cuSPARSE context.</td>
</tr>
<tr>
<td>streamId</td>
<td>the stream to be used by the library.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the stream was set successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the library was not initialized.</td>
</tr>
</tbody>
</table>

### 5.24. cusparseGetStream()

cusparseStatus_t
cusparseGetStream(cusparseHandle_t handle, cudaStream_t *streamId)

This function gets the cuSPARSE library stream, which is being used to execute all calls to the cuSPARSE library functions. If the cuSPARSE library stream is not set, all kernels use the default NULL stream.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>the handle to the cuSPARSE context.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>streamId</td>
<td>the stream to be used by the library.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the stream was returned successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the library was not initialized.</td>
</tr>
</tbody>
</table>
5.25. cusparseCreateCsrsv2Info()

cusparseStatus_t
 cusparseCreateCsrsv2Info(csrsv2Info_t *info);

This function creates and initializes the solve and analysis structure of csrsv2 to default values.

**Input**

| info | the pointer to the solve and analysis structure of csrsv2. |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the structure was initialized successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.26. cusparseDestroyCsrsv2Info()

cusparseStatus_t
 cusparseDestroyCsrsv2Info(csrsv2Info_t info);

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

**Input**

| info | the solve (csrsv2_solve) and analysis (csrsv2_analysis) structure. |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully. |

5.27. cusparseCreateCsrsm2Info()

cusparseStatus_t
 cusparseCreateCsrsm2Info(csrsm2Info_t *info);

This function creates and initializes the solve and analysis structure of csrsm2 to default values.

**Input**

| info | the pointer to the solve and analysis structure of csrsm2. |

**Status Returned**
### 5.28. `cusparseDestroyCsrsm2Info()`

```c
cusparseStatus_t
cusparseDestroyCsrsm2Info(csrsm2Info_t info);
```

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

**Input**

| info | the solve (csrsm2_solve) and analysis (csrsm2_analysis) structure. |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully. |

### 5.29. `cusparseCreateCsric02Info()`

```c
cusparseStatus_t
cusparseCreateCsric02Info(csric02Info_t *info);
```

This function creates and initializes the solve and analysis structure of incomplete Cholesky to default values.

**Input**

| info | the pointer to the solve and analysis structure of incomplete Cholesky. |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the structure was initialized successfully. |
| CUSPARSE_STATUS_ALLOC_FAILED | the resources could not be allocated. |

### 5.30. `cusparseDestroyCsric02Info()`

```c
cusparseStatus_t
cusparseDestroyCsric02Info(csric02Info_t info);
```

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

**Input**

| info | the solve (csric02_solve) and analysis (csric02_analysis) structure. |
Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the resources were released successfully.</td>
</tr>
</tbody>
</table>

5.31. cusparseCreateCsrilu02Info()

cusparseStatus_t cusparseCreateCsrilu02Info(csrilu02Info_t *info);

This function creates and initializes the solve and analysis structure of incomplete LU to default values.

Input

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>the pointer to the solve and analysis structure of incomplete LU.</td>
</tr>
</tbody>
</table>

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the structure was initialized successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.32. cusparseDestroyCsrilu02Info()

cusparseStatus_t cusparseDestroyCsrilu02Info(csrilu02Info_t info);

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

Input

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>the solve (csrilu02_solve) and analysis (csrilu02_analysis) structure.</td>
</tr>
</tbody>
</table>

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the resources were released successfully.</td>
</tr>
</tbody>
</table>

5.33. cusparseCreateBsrsv2Info()

cusparseStatus_t cusparseCreateBsrsv2Info(bsrsrv2Info_t *info);

This function creates and initializes the solve and analysis structure of bsrsv2 to default values.

Input

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>the pointer to the solve and analysis structure of bsrsv2.</td>
</tr>
</tbody>
</table>
Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the structure was initialized successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.34. cusparseDestroyBsrsv2Info()

cusparseStatus_t
cusparseDestroyBsrsv2Info(bsrsv2Info_t info);

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

Input

| info | the solve (bsrsv2_solve) and analysis (bsrsv2_analysis) structure. |

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the resources were released successfully.</td>
</tr>
</tbody>
</table>

5.35. cusparseCreateBsrsm2Info()

cusparseStatus_t
cusparseCreateBsrsm2Info(bsrsm2Info_t *info);

This function creates and initializes the solve and analysis structure of bsrsm2 to default values.

Input

| info | the pointer to the solve and analysis structure of bsrsm2. |

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the structure was initialized successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.36. cusparseDestroyBsrsm2Info()

cusparseStatus_t
cusparseDestroyBsrsm2Info(bsrsm2Info_t info);

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

Input

| info | the solve (bsrsm2_solve) and analysis (bsrsm2_analysis) structure. |
5.37. `cusparseCreateBsric02Info()`

```c
 cusparseStatus_t
cusparseCreateBsric02Info(bsric02Info_t *info);
```

This function creates and initializes the solve and analysis structure of block incomplete Cholesky to *default* values.

**Input**

| info                 | the pointer to the solve and analysis structure of block incomplete Cholesky. |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the structure was initialized successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
</tbody>
</table>

5.38. `cusparseDestroyBsric02Info()`

```c
 cusparseStatus_t
cusparseDestroyBsric02Info(bsric02Info_t info);
```

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

**Input**

| info                 | the solve (bsric02_solve) and analysis (bsric02_analysis) structure. |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully. |

5.39. `cusparseCreateBsrilu02Info()`

```c
 cusparseStatus_t
cusparseCreateBsrilu02Info(bsrilu02Info_t *info);
```

This function creates and initializes the solve and analysis structure of block incomplete LU to *default* values.

**Input**

| info                 | the pointer to the solve and analysis structure of block incomplete LU. |

5.40. **cusparseDestroyBsrilu02Info()**

```c
 cusparseStatus_t
cusparseDestroyBsrilu02Info(bsrilu02Info_t info);
```

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

**Input**

| info                          | the solve (bsrilu02_solve) and analysis (bsrilu02_analysis) structure. |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS       | the resources were released successfully. |
5.43. cusparseCreatePruneInfo()

cusparseStatus_t
cusparseCreatePruneInfo(pruneInfo_t *info);

This function creates and initializes structure of prune to default values.

Input

| info          | the pointer to the structure of prune. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the structure was initialized successfully. |
| CUSPARSE_STATUS_ALLOC_FAILED | the resources could not be allocated. |

5.44. cusparseDestroyPruneInfo()

cusparseStatus_t
cusparseDestroyPruneInfo(pruneInfo_t info);

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

Input

| info          | the structure of prune. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the resources were released successfully. |
This chapter describes sparse linear algebra functions that perform operations between dense and sparse vectors.

6.1. cusparse<t>axpyi()

This function multiplies the vector \( x \) in sparse format by the constant \( \alpha \) and adds the result to the vector \( y \) in dense format. This operation can be written as

\[
y = y + \alpha \times x
\]

In other words,

\[
y[xInd[i]-idxBase] = y[xInd[i]-idxBase] + \alpha \times Val[i]
\]

This function requires no extra storage. It is executed asynchronously with respect to the host, and it may return control to the application on the host before the result is ready.

Input
handle  handle to the cuSPARSE library context.
nnz   number of elements in vector \( \mathbf{x} \).
alpha <type> scalar used for multiplication.
xVal <type> vector with \( \text{nnz} \) nonzero values of vector \( \mathbf{x} \).
xInd integer vector with \( \text{nnz} \) indices of the nonzero values of vector \( \mathbf{x} \).
y  <type> vector in dense format.
idxBase CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.

Output

\( \mathbf{y} \)  <type> updated vector in dense format (that is unchanged if \( \text{nnz} = 0 \)).

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the idxBase is neither CUSPARSE_INDEX_BASE_ZERO nor CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
</tbody>
</table>

6.2. \texttt{cusparse<t>doti()}

```c
 cusparseStatus_t
 cusparseSdoti(cusparseHandle_t handle, int nnz,
              const float *xVal,
              const int *xInd, const float *y,
              float *resultDevHostPtr,
              cusparseIndexBase_t idxBase)
cusparseStatus_t
 cusparseDdoti(cusparseHandle_t handle, int nnz,
              const double *xVal,
              const int *xInd, const double *y,
              double *resultDevHostPtr,
              cusparseIndexBase_t idxBase)
cusparseStatus_t
 cusparseCdoti(cusparseHandle_t handle, int nnz,
              const cuComplex *xVal,
              const int *xInd, const cuComplex *y,
              cuComplex *resultDevHostPtr,
              cusparseIndexBase_t idxBase)
cusparseStatus_t
 cusparseZdoti(cusparseHandle_t handle, int nnz, const
              cuDoubleComplex *xVal,
              const int *xInd, const cuDoubleComplex *y,
              cuDoubleComplex *resultDevHostPtr,
              cusparseIndexBase_t idxBase)
```
This function returns the dot product of a vector \( \mathbf{x} \) in sparse format and vector \( \mathbf{y} \) in dense format. This operation can be written as

\[
result = \mathbf{y}^T \mathbf{x}
\]

In other words,

\[
\text{for } i=0 \text{ to } \text{nnz}-1 \\
\quad \text{resultDevHostPtr} += \mathbf{xVal}[i] \times \mathbf{y}[\mathbf{xInd}[i-\text{idxBase}]]
\]

This function requires some temporary extra storage that is allocated internally. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnz</td>
<td>number of elements in vector ( \mathbf{x} ).</td>
</tr>
<tr>
<td>( \mathbf{xVal} )</td>
<td>(&lt;\text{type}&gt;) vector with ( \text{nnz} ) nonzero values of vector ( \mathbf{x} ).</td>
</tr>
<tr>
<td>( \mathbf{xInd} )</td>
<td>integer vector with ( \text{nnz} ) indices of the nonzero values of vector ( \mathbf{x} ).</td>
</tr>
<tr>
<td>( \mathbf{y} )</td>
<td>(&lt;\text{type}&gt;) vector in dense format.</td>
</tr>
<tr>
<td>resultDevHostPtr</td>
<td>pointer to the location of the result in the device or host memory.</td>
</tr>
<tr>
<td>idxBase</td>
<td>( \text{CUSPARSE_INDEX_BASE_ZERO} ) or ( \text{CUSPARSE_INDEX_BASE_ONE} ).</td>
</tr>
</tbody>
</table>

**Output**

| resultDevHostPtr        | scalar result in the device or host memory (that is zero if \( \text{nnz} == 0 \)). |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the reduction buffer could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the ( \text{idxBase} ) is neither ( \text{CUSPARSE_INDEX_BASE_ZERO} ) nor ( \text{CUSPARSE_INDEX_BASE_ONE} ).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
6.3. cusparse<t>dotci()

cusparseStatus_t
cusparseCdotci(cusparseHandle_t handle, int nnz,
const cuComplex *xVal,
const int *xInd, const cuComplex *y,
cuComplex *resultDevHostPtr, cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseZdotci(cusparseHandle_t handle, int nnz,
const cuDoubleComplex *xVal,
const int *xInd, const cuDoubleComplex *y,
cuDoubleComplex *resultDevHostPtr, cusparseIndexBase_t idxBase)

This function returns the dot product of a complex conjugate of vector \( \mathbf{x} \) in sparse format and vector \( \mathbf{y} \) in dense format. This operation can be written as

\[
result = \mathbf{x}^H \mathbf{y}
\]

In other words,

\[
\text{for } i=0 \text{ to } \text{nnz}-1 \\
\text{resultDevHostPtr } += \overline{xVal[i]} \cdot y[xInd[i-idxBASE]]
\]

This function requires some temporary extra storage that is allocated internally. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of elements in vector ( \mathbf{x} ).</td>
</tr>
<tr>
<td>xVal</td>
<td>(&lt;\text{type}&gt;) vector with ( \text{nnz} ) nonzero values of vector ( \mathbf{x} ).</td>
</tr>
<tr>
<td>xInd</td>
<td>integer vector with ( \text{nnz} ) indices of the nonzero values of vector ( \mathbf{x} ).</td>
</tr>
<tr>
<td>y</td>
<td>(&lt;\text{type}&gt;) vector in dense format.</td>
</tr>
<tr>
<td>resultDevHostPtr</td>
<td>pointer to the location of the result in the device or host memory.</td>
</tr>
<tr>
<td>idxBase</td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>resultDevHostPtr</td>
<td>scalar result in the device or host memory (that is zero if ( \text{nnz} = 0 )).</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the reduction buffer could not be allocated.</td>
</tr>
</tbody>
</table>
CUSPARSE_STATUS_INVALID_VALUE | the idxBase is neither CUSPARSE_INDEX_BASE_ZERO nor CUSPARSE_INDEX_BASE_ONE.

CUSPARSE_STATUS_ARCH_MISMATCH | the device does not support double precision.

CUSPARSE_STATUS_EXECUTION_FAILED | the function failed to launch on the GPU.

CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed.

### 6.4. cusparse<t>gthr()

```c
 cusparseStatus_t
cusparseSgthr(cusparseHandle_t handle, int nnz,
              const float       *y,
              float           *xVal, const int *xInd,
              cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseDgthr(cusparseHandle_t handle, int nnz,
              const double     *y,
              double          *xVal, const int *xInd,
              cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseCgthr(cusparseHandle_t handle, int nnz,
              const cuComplex   *y,
              cuComplex       *xVal, const int *xInd,
              cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseZgthr(cusparseHandle_t handle, int nnz,
              const cuDoubleComplex *y,
              cuDoubleComplex   *xVal, const int *xInd,
              cusparseIndexBase_t idxBase)
```

This function gathers the elements of the vector \( y \) listed in the index array \( x\text{Ind} \) into the data array \( x\text{Val} \).

This function requires no extra storage. It is executed asynchronously with respect to the host and it may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnz</td>
<td>number of elements in vector ( x ).</td>
</tr>
<tr>
<td>( y )</td>
<td>&lt;type&gt; vector in dense format (of ( \text{size\geq max(xInd)-idxBase+1} )).</td>
</tr>
<tr>
<td>( x\text{Ind} )</td>
<td>integer vector with ( nnz ) indices of the nonzero values of vector ( x ).</td>
</tr>
<tr>
<td>idxBase</td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

**Output**

| \( x\text{Val} \) | \<type> vector with \( nnz \) nonzero values that were gathered from vector \( y \) (that is unchanged if \( nnz == 0 \)). |
Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the idxBase is neither CUSPARSE_INDEX_BASE_ZERO nor CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
</tbody>
</table>

6.5. cusparse<t>gthrz()

cusparseStatus_t
cusparseSgthrz(cusparseHandle_t handle, int nnz, float *y, float *xVal, const int *xInd, cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseDgthrz(cusparseHandle_t handle, int nnz, double *y, double *xVal, const int *xInd, cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseCgthrz(cusparseHandle_t handle, int nnz, cuComplex *y, cuComplex *xVal, const int *xInd, cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseZgthrz(cusparseHandle_t handle, int nnz, cuDoubleComplex *y, cuDoubleComplex *xVal, const int *xInd, cusparseIndexBase_t idxBase)

This function gathers the elements of the vector y listed in the index array xInd into the data array xVal. Also, it zeros out the gathered elements in the vector y.

This function requires no extra storage. It is executed asynchronously with respect to the host, and it may return control to the application on the host before the result is ready.

Input

<table>
<thead>
<tr>
<th>Input</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of elements in vector x.</td>
</tr>
<tr>
<td>y</td>
<td>&lt;type&gt; vector in dense format (of size≥max(xInd)-idxBase+1).</td>
</tr>
<tr>
<td>xInd</td>
<td>integer vector with nnz indices of the nonzero values of vector x.</td>
</tr>
<tr>
<td>idxBase</td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

Output

<table>
<thead>
<tr>
<th>Output</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>xVal</td>
<td>&lt;type&gt; vector with nnz nonzero values that were gathered from vector y (that is unchanged if nnz == 0).</td>
</tr>
</tbody>
</table>
6.6. cusparse\langle t\rangle roti()

cusparseStatus_t
cusparseSroti(cusparseHandle_t handle, int nnz, float *xVal,
const int *xInd,
float *y, const float *c, const float *s,
cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseDroti(cusparseHandle_t handle, int nnz, double *xVal,
const int *xInd,
double *y, const double *c, const double *s,
cusparseIndexBase_t idxBase)

This function applies the Givens rotation matrix

\[
G = \begin{pmatrix} c & s \\ -s & c \end{pmatrix}
\]

to sparse \(x\) and dense \(y\) vectors. In other words,

\[
\begin{align*}
\text{for } i = 0 \text{ to } \text{nnz}-1 \\
\quad y[\text{xInd}[i]-\text{idxBase}] &= c \cdot y[\text{xInd}[i]-\text{idxBase}] - s \cdot \text{xVal}[i] \\
\quad \text{x}[i] &= c \cdot \text{xVal}[i] + s \cdot y[\text{xInd}[i]-\text{idxBase}]
\end{align*}
\]

### Input

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnz</td>
<td>number of elements in vector (x).</td>
</tr>
<tr>
<td>xVal</td>
<td>\langle type\rangle vector with \text{nnz} nonzero values of vector (x).</td>
</tr>
<tr>
<td>xInd</td>
<td>integer vector with \text{nnz} indices of the nonzero values of vector (x).</td>
</tr>
<tr>
<td>y</td>
<td>\langle type\rangle vector in dense format.</td>
</tr>
<tr>
<td>c</td>
<td>cosine element of the rotation matrix.</td>
</tr>
<tr>
<td>s</td>
<td>sine element of the rotation matrix.</td>
</tr>
<tr>
<td>idxBase</td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>
Output

<table>
<thead>
<tr>
<th>xVal</th>
<th>&lt;type&gt; updated vector in sparse format (that is unchanged if nnz == 0).</th>
</tr>
</thead>
<tbody>
<tr>
<td>y</td>
<td>&lt;type&gt; updated vector in dense format (that is unchanged if nnz == 0).</td>
</tr>
</tbody>
</table>

Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the idxBase is neither CUSPARSE_INDEX_BASE_ZERO nor CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
</tbody>
</table>

6.7. cusparse<t>sctr()

cusparseStatus_t
cusparseSsctr(cusparseHandle_t handle, int nnz,
    const float *xVal,
    const int *xInd, float *y,
    cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseDsctr(cusparseHandle_t handle, int nnz,
    const double *xVal,
    const int *xInd, double *y,
    cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseCsctr(cusparseHandle_t handle, int nnz,
    const cuComplex *xVal,
    const int *xInd, cuComplex *y,
    cusparseIndexBase_t idxBase)
cusparseStatus_t
cusparseZsctr(cusparseHandle_t handle, int nnz,
    const cuDoubleComplex *xVal,
    const int *xInd, cuDoubleComplex *y,
    cusparseIndexBase_t idxBase)

This function scatters the elements of the vector \( x \) in sparse format into the vector \( y \) in dense format. It modifies only the elements of \( y \) whose indices are listed in the array \( x\text{Ind} \).

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

Input

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnz</td>
<td>number of elements in vector ( x ).</td>
</tr>
<tr>
<td>xVal</td>
<td>&lt;type&gt; vector with ( nnz ) nonzero values of vector ( x ).</td>
</tr>
<tr>
<td>xInd</td>
<td>integer vector with ( \text{nnz} ) indices of the nonzero values of vector ( x ).</td>
</tr>
<tr>
<td>---</td>
<td>---</td>
</tr>
<tr>
<td>y</td>
<td>(&lt;\text{type}&gt;) dense vector (of size ( \geq \max(x\text{Ind}) - \text{idxBase} + 1 )).</td>
</tr>
<tr>
<td>idxBase</td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

**Output**

| y | \(<\text{type}>\) vector with \( \text{nnz} \) nonzero values that were scattered from vector \( x \) (that is unchanged if \( \text{nnz} = 0 \)). |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the ( \text{idxBase} ) is neither CUSPARSE_INDEX_BASE_ZERO nor CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
</tbody>
</table>
This chapter describes the sparse linear algebra functions that perform operations between sparse matrices and dense vectors.

In particular, the solution of sparse triangular linear systems is implemented in two phases. First, during the analysis phase, the sparse triangular matrix is analyzed to determine the dependencies between its elements by calling the appropriate `csrsv_analysis()` function. The analysis is specific to the sparsity pattern of the given matrix and to the selected `cusparseOperation_t` type. The information from the analysis phase is stored in the parameter of type `cusparseSolveAnalysisInfo_t` that has been initialized previously with a call to `cusparseCreateSolveAnalysisInfo()`.

Second, during the solve phase, the given sparse triangular linear system is solved using the information stored in the `cusparseSolveAnalysisInfo_t` parameter by calling the appropriate `csrsv_solve()` function. The solve phase may be performed multiple times with different right-hand sides, while the analysis phase needs to be performed only once. This is especially useful when a sparse triangular linear system must be solved for a set of different right-hand sides one at a time, while its coefficient matrix remains the same.

Finally, once all the solves have completed, the opaque data structure pointed to by the `cusparseSolveAnalysisInfo_t` parameter can be released by calling `cusparseDestroySolveAnalysisInfo()`. For more information please refer to [3].
7.1. `cusparse<t>bsrmv()`

This function performs the matrix-vector operation

\[ y = \alpha \cdot \text{op}(A) \cdot x + \beta \cdot y \]

where \( A \) is an \((mb \times \text{blockDim}) \times (nb \times \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays `bsrVal`, `bsrRowPtr`, and `bsrColInd`; `x` and `y` are vectors; \( \alpha \) and \( \beta \) are scalars; and

\[
\text{op}(A) = \begin{cases} 
  A & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
  A^T & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_TRANSPOSE} \\
  A^H & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} 
\end{cases}
\]

Several comments on `bsrmv()`:

- Only `CUSPARSE_OPERATION_NON_TRANSPOSE` is supported, that is
  \[ y = \alpha \cdot A \cdot x + \beta \cdot y \]

- Only `CUSPARSE_MATRIX_TYPE_GENERAL` is supported.

- The size of vector `x` should be \((nb \times \text{blockDim})\) at least, and the size of vector `y` should be \((mb \times \text{blockDim})\) at least; otherwise, the kernel may return `CUSPARSE_STATUS_EXECUTION_FAILED` because of an out-of-bounds array.
For example, suppose the user has a CSR format and wants to try `bsrmv()`, the following code demonstrates how to use `csr2bsr()` conversion and `bsrmv()` multiplication in single precision.

```c
// Suppose that A is m x n sparse matrix represented by CSR format,
// hx is a host vector of size n, and hy is also a host vector of size m.
// m and n are not multiple of blockDim.
// step 1: transform CSR to BSR with column-major order
int base, nnz;
int nnzb;
cusparseDirection_t dirA = CUSPARSE_DIRECTION_COLUMN;
int mb = (m + blockDim-1)/blockDim;
int nb = (n + blockDim-1)/blockDim;
cudaMalloc((void**)&bsrRowPtrC, sizeof(int) *(mb+1));
cusparseXcsr2bsrNnz(handle, dirA, m, n,
   descrA, csrRowPtrA, csrColIndA, blockDim,
   descrC, bsrRowPtrC, &nnzb);

cudaMalloc((void**)&bsrValC, sizeof(float)*(blockDim*blockDim)*nnzb);
cusparseScsr2bsr(handle, dirA, m, n,
   descrA, csrValA, csrRowPtrA, csrColIndA, blockDim,
   descrC, bsrValC, bsrRowPtrC, bsrColIndC);

// step 2: allocate vector x and vector y large enough for bsrmv
cudaMalloc((void**)&x, sizeof(float)*(nb*blockDim));
cudaMalloc((void**)&y, sizeof(float)*(mb*blockDim));
cudaMalloc((void**)&brix, sizeof(float)*m, cudaMallocHostToDevice);
cudaMemcpy(brix, bx, sizeof(float)*m, cudaMallocHostToDevice);

// step 3: perform bsrmv
cusparseSbsrmv(handle, dirA, transA, mb, nb, nnzb, &alpha,
   descrC, bsrValC, bsrRowPtrC, bsrColIndC, blockDim, x, &beta, y);
```

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>handle</strong></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><strong>dir</strong></td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td><strong>trans</strong></td>
<td>the operation op(A). Only CUSPARSE_OPERATION_NON_TRANSPOSE is supported.</td>
</tr>
<tr>
<td><strong>mb</strong></td>
<td>number of block rows of matrix A.</td>
</tr>
<tr>
<td><strong>nb</strong></td>
<td>number of block columns of matrix A.</td>
</tr>
<tr>
<td><strong>nnzb</strong></td>
<td>number of nonzero blocks of matrix A.</td>
</tr>
<tr>
<td><strong>alpha</strong></td>
<td>&lt;type&gt; scalar used for multiplication.</td>
</tr>
<tr>
<td><strong>descr</strong></td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td><strong>bsrVal</strong></td>
<td>&lt;type&gt; array of nnz (= csrRowPtrA(mb) - csrRowPtrA(0)) nonzero blocks of matrix A.</td>
</tr>
<tr>
<td><strong>bsrRowPtr</strong></td>
<td>integer array of mb + 1 elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
</tbody>
</table>
### bsrColInd
- **integer array of** \( \text{nnz} = \text{csrRowPtrA}(\text{mb}) - \text{csrRowPtrA}(0) \) **column indices of the nonzero blocks of matrix** \( A \).

### blockDim
- **block dimension of sparse matrix** \( A \), **larger than zero**.

### x
- **\(<\text{type}>\) vector of** \( nb \times \text{blockDim} \) **elements**.

### beta
- **\(<\text{type}>\) scalar used for multiplication**. If **beta** is zero, **y** **does not have to be a valid input**.

### y
- **\(<\text{type}>\) vector of** \( mb \times \text{blockDim} \) **elements**.

### Output
- **y** **\(<\text{type}>\) updated vector.**

### Status Returned
<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (( m,n,\text{nnz}&lt;0 ), ( \text{trans} \neq \text{CUSPARSE_OPERATION_NON_TRANSPOSE} ), ( \text{blockDim}&lt;1 ), ( \text{dir} ) is not row-major or column-major, or ( \text{IndexBase} ) of descr is not base-0 or base-1 ).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.2. cusparse<t>bsrsmv()

cusparseStatus_t
cusparseSbsrsmv(cusparseHandle_t handle,
cusparseDirection_t dir,
cusparseOperation_t trans,
int sizeOfMask,
int mb,
int nb,
int nnzb,
const float *alpha,
const cusparseMatDescr_t descr,
const float *bsrVal,
const int *bsrMaskPtr,
const int *bsrRowPtr,
const int *bsrEndPtr,
const int *bsrColInd,
int blockDim,
const float *x,
const float *beta,
float *y)

cusparseStatus_t
cusparseDbsrsmv(cusparseHandle_t handle,
cusparseDirection_t dir,
cusparseOperation_t trans,
int sizeOfMask,
int mb,
int nb,
int nnzb,
const double *alpha,
const cusparseMatDescr_t descr,
const double *bsrVal,
const int *bsrMaskPtr,
const int *bsrRowPtr,
const int *bsrEndPtr,
const int *bsrColInd,
int blockDim,
const double *x,
const double *beta,
double *y)

cusparseStatus_t
cusparseCbsrsmv(cusparseHandle_t handle,
cusparseDirection_t dir,
cusparseOperation_t trans,
int sizeOfMask,
int mb,
int nb,
int nnzb,
const cuComplex *alpha,
const cusparseMatDescr_t descr,
const cuComplex *bsrVal,
const int *bsrMaskPtr,
const int *bsrRowPtr,
const int *bsrEndPtr,
const int *bsrColInd,
int blockDim,
const cuComplex *x,
const cuComplex *beta,
cuComplex *y)

cusparseStatus_t
cusparseZbsrsmv(cusparseHandle_t handle,
cusparseDirection_t dir,
cusparseOperation_t trans,
int sizeOfMask,
int mb,
This function performs a \texttt{bsrmv} and a mask operation
\[
y(\text{mask}) = (\alpha \ast \text{op}(A) \ast x + \beta \ast y)(\text{mask})
\]
where \(A\) is an \((mb \times \text{blockDim}) \times (nb \times \text{blockDim})\) sparse matrix that is defined in BSRX storage format by the four arrays \texttt{bsrVal}, \texttt{bsrRowPtr}, \texttt{bsrEndPtr}, and \texttt{bsrColInd}; \(x\) and \(y\) are vectors; \(\alpha\) and \(\beta\) are scalars; and
\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE}
\end{cases}
\]
The mask operation is defined by array \texttt{bsrMaskPtr} which contains updated block row indices of \(y\). If row \(i\) is not specified in \texttt{bsrMaskPtr}, then \texttt{bsrmv()} does not touch row block \(i\) of \(A\) and \(y\).

For example, consider the \(2 \times 3\) block matrix \(A\):
\[
A = \begin{bmatrix} 
A_{11} & A_{12} & O \\
A_{21} & A_{22} & A_{23} 
\end{bmatrix}
\]
and its one-based BSR format (three vector form) is
\[
\begin{align*}
\text{bsrVal} &= [A_{11} \ A_{12} \ A_{21} \ A_{22} \ A_{23}] \\
\text{bsrRowPtr} &= [1 \ 3 \ 6] \\
\text{bsrColInd} &= [1 \ 2 \ 1 \ 2 \ 3]
\end{align*}
\]
Suppose we want to do the following \texttt{bsrmv} operation on a matrix \(\overline{A}\) which is slightly different from \(A\).
\[
\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} := \alpha \ast \overline{A} \ast \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \beta \ast y_2
\]
We don’t need to create another BSR format for the new matrix \(\overline{A}\), all that we should do is to keep \texttt{bsrVal} and \texttt{bsrColInd} unchanged, but modify \texttt{bsrRowPtr} and add an additional array \texttt{bsrEndPtr} which points to the last nonzero elements per row of \(\overline{A}\) plus \(1\).

For example, the following \texttt{bsrRowPtr} and \texttt{bsrEndPtr} can represent matrix \(\overline{A}\):
\[
\begin{align*}
\text{bsrRowPtr} &= [1 \ 4] \\
\text{bsrEndPtr} &= [1 \ 5]
\end{align*}
\]
Further we can use a mask operator (specified by array \texttt{bsrMaskPtr}) to update particular block row indices of \(y\) only because \(y_1\) is never changed. In this case, \texttt{bsrMaskPtr} = [2] and \textbf{sizeOfMask} = 1.

The mask operator is equivalent to the following operation:
\[
\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} := \alpha \ast \begin{bmatrix} ? \\ ? \end{bmatrix} \ast \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \beta \ast \begin{bmatrix} ? \end{bmatrix}
\]
If a block row is not present in the `bsrMaskPtr`, then no calculation is performed on that row, and the corresponding value in \( y \) is unmodified. The question mark "?" is used to indicate row blocks not in `bsrMaskPtr`.

In this case, first row block is not present in `bsrMaskPtr`, so `bsrRowPtr[0]` and `bsrEndPtr[0]` are not touched also.

\[
\begin{align*}
\text{bsrRowPtr} &= [? 4] \\
\text{bsrEndPtr} &= [? 5]
\end{align*}
\]

A couple of comments on `bsrsmv()`:

- Only `CUSPARSE_OPERATION_NON_TRANSPOSE` and `CUSPARSE_MATRIX_TYPE_GENERAL` are supported.
- Parameters `bsrMaskPtr`, `bsrRowPtr`, `bsrEndPtr` and `bsrColInd` are consistent with base index, either one-based or zero-based. The above example is one-based.

**Input**

<table>
<thead>
<tr>
<th><code>handle</code></th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>dir</code></td>
<td>storage format of blocks, either <code>CUSPARSE_DIRECTION_ROW</code> or <code>CUSPARSE_DIRECTION_COLUMN</code>.</td>
</tr>
<tr>
<td><code>trans</code></td>
<td>the operation ( \text{op}(A) ). Only <code>CUSPARSE_OPERATION_NON_TRANSPOSE</code> is supported.</td>
</tr>
<tr>
<td><code>sizeOfMask</code></td>
<td>number of updated block rows of ( y ).</td>
</tr>
<tr>
<td><code>mb</code></td>
<td>number of block rows of matrix ( A ).</td>
</tr>
<tr>
<td><code>nb</code></td>
<td>number of block columns of matrix ( A ).</td>
</tr>
<tr>
<td><code>nnzb</code></td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><code>alpha</code></td>
<td><code>&lt;type&gt;</code> scalar used for multiplication.</td>
</tr>
<tr>
<td><code>descr</code></td>
<td>the descriptor of matrix ( A ). 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>.</td>
</tr>
<tr>
<td><code>bsrVal</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnz</code> nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><code>bsrMaskPtr</code></td>
<td>integer array of <code>sizeOfMask</code> elements that contains the indices corresponding to updated block rows.</td>
</tr>
<tr>
<td><code>bsrRowPtr</code></td>
<td>integer array of <code>mb</code> elements that contains the start of every block row.</td>
</tr>
<tr>
<td><code>bsrEndPtr</code></td>
<td>integer array of <code>mb</code> elements that contains the end of every block row plus one.</td>
</tr>
<tr>
<td><code>bsrColInd</code></td>
<td>integer array of <code>nnzb</code> column indices of the nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><code>blockDim</code></td>
<td>block dimension of sparse matrix ( A ), larger than zero.</td>
</tr>
<tr>
<td><strong>x</strong></td>
<td>&lt;type&gt; vector of ( nb \times blockDim ) elements.</td>
</tr>
<tr>
<td>------------</td>
<td>-----------------------------------------------------------</td>
</tr>
<tr>
<td><strong>beta</strong></td>
<td>&lt;type&gt; scalar used for multiplication. If ( \text{beta} ) is zero, ( y ) does not have to be a valid input.</td>
</tr>
<tr>
<td><strong>y</strong></td>
<td>&lt;type&gt; vector of ( mb \times blockDim ) elements.</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ((m,n,nnz&lt;0, \text{trans} \neq \text{CUSPARSE_OPERATION_NON_TRANSPOSE}, \text{blockDim} &lt; 1, \text{dir} \text{ is not row-major or column-major, or IndexBase of descr is not base-0 or base-1 })).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
This function performs the matrix-vector operation

\[ y = \alpha \cdot \text{op}(A) \cdot x + \beta \cdot y \]

\( A \) is an \( m \times n \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA}, \text{csrRowPtrA}, \) and \( \text{csrColIndA} \); \( x \) and \( y \) are vectors; \( \alpha \) and \( \beta \) are scalars; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE}
\end{cases}
\]

When using the (conjugate) transpose of a general matrix or a Hermitian/symmetric matrix, this routine may produce slightly different results during different runs with the same input parameters. For these matrix types it uses atomic operations to compute the final result, consequently many threads may be adding floating point numbers to the same memory location without any specific ordering, which may produce slightly different results for each run.

If exactly the same output is required for any input when multiplying by the transpose of a general matrix, the following procedure can be used:
1. Convert the matrix from CSR to CSC format using one of the `csr2csc()` functions. Notice that by interchanging the rows and columns of the result you are implicitly transposing the matrix.

2. Call the `csrmv()` function with the `cusparseOperation_t` parameter set to `CUSPARSE_OPERATION_NON_TRANSPOSE` and with the interchanged rows and columns of the matrix stored in CSC format. This (implicitly) multiplies the vector by the transpose of the matrix in the original CSR format.

This function requires no extra storage for the general matrices when operation `CUSPARSE_OPERATION_NON_TRANSPOSE` is selected. It requires some extra storage for Hermitian/symmetric matrices and for the general matrices when an operation different than `CUSPARSE_OPERATION_NON_TRANSPOSE` is selected. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>trans</td>
<td>the operation op(A).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix A.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of matrix A.</td>
</tr>
<tr>
<td>alpha</td>
<td>&lt;type&gt; scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is</td>
</tr>
<tr>
<td></td>
<td><code>CUSPARSE_MATRIX_TYPE_GENERAL</code>, <code>CUSPARSE_MATRIX_TYPE_SYMMETRIC</code>, and</td>
</tr>
<tr>
<td></td>
<td><code>CUSPARSE_MATRIX_TYPE_HERMITIAN</code>. Also, the supported index bases are</td>
</tr>
<tr>
<td></td>
<td><code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code>.</td>
</tr>
<tr>
<td>csrValA</td>
<td>&lt;type&gt; array of nnz (= csrRowPtrA(m) - csrRowPtrA(0)) nonzero elements of</td>
</tr>
<tr>
<td></td>
<td>matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of m+1 elements that contains the start of every row and the</td>
</tr>
<tr>
<td></td>
<td>end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of nnz (= csrRowPtrA(m) - csrRowPtrA(0)) column indices of</td>
</tr>
<tr>
<td></td>
<td>the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>x</td>
<td>&lt;type&gt; vector of n elements if op(A) = A, and m elements if op(A) = AT or</td>
</tr>
<tr>
<td></td>
<td>op(A) = AH.</td>
</tr>
<tr>
<td>beta</td>
<td>&lt;type&gt; scalar used for multiplication. If beta is zero, y does not have to</td>
</tr>
<tr>
<td></td>
<td>be a valid input.</td>
</tr>
<tr>
<td>y</td>
<td>&lt;type&gt; vector of m elements if op(A) = A, and n elements if op(A) = AT or</td>
</tr>
<tr>
<td></td>
<td>op(A) = AH.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>y</td>
<td>&lt;type&gt; updated vector.</td>
</tr>
</tbody>
</table>
Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n, nnz&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision (compute capability (c.c.) &gt;= 1.3 required), symmetric/Hermitian matrix (c.c. &gt;= 1.2 required), or transpose operation (c.c. &gt;= 1.1 required).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

7.4. cusparseCsrmvEx()

cusparseStatus_t
cusparseCsrmvEx(cusparseHandle_t handle,  
cusparseAlgMode_t alg,  
cusparseOperation_t transA,  
int m, int n, int nnz,  
const void *alpha, cudaDataType alphatype,  
const cusparseMatDescr_t descrA,  
const void *csrValA, cudaDataType csrValAtype,  
const int *csrRowPtrA,  
const int *csrColIndA,  
const void *x, cudaDataType xtype,  
const void *beta, cudaDataType betatype,  
void *y, cudaDataType ytype,  
cudaDataType executiontype,  
void* buffer);

This function is an extended version of cusparse<t>csrmv() which performs the matrix-vector multiply operation. For detailed description of the functionality, see cusparse<t>csrmv(). Also see cusparseAlgMode_t for alg parameter description

For alg CUSPARSE_ALG_NAIVE: for half-precision execution type, the minimum GPU architecture is SM_53. Also, for both half-precision IO and execution, only CUSPARSE_MATRIX_TYPEGENERAL and CUSPARSE_OPERATION_NON_TRANSPOSE are supported.

For alg CUSPARSE_ALG_MERGE_PATH: half-precision is not supported, only CUSPARSE_MATRIX_TYPEGENERAL, CUSPARSE_OPERATION_NON_TRANSPOSE and CUSPARSE_INDEX_BASE_ZERO are supported. Input, output and execution types should be the same.

All pointers should be aligned with 128 bytes.

Input specifically required by cusparseCsrmvEx

<table>
<thead>
<tr>
<th>alg</th>
<th>Algorithm implementation for csrmv, see cusparseAlgMode_t for possible values.</th>
</tr>
</thead>
</table>

DU-06709-001_v10.1 | 57
### 7.5. cusparseCsrmvEx_bufferSize()

```c
 cusparseCsrmvEx_bufferSize(cusparseHandle_t handle,
                           cusparseAlgMode_t alg,
                           cusparseOperation_t transA,
                           int m, int n, int nnz,
                           const void *alpha,
                           cudaDataType alphatype,
                           const cusparseMatDescr_t descrA,
                           const void *csrValA,
                           cudaDataType csrValAtype,
                           const int *csrRowPtrA,
                           const int *csrColIndA,
                           const void *x,
                           cudaDataType xtype,
                           const void *beta,
                           cudaDataType betatype,
                           void *y, cudaDataType ytype,
                           cusparseOperation_t executiontype,
                           size_t *bufferSizeInBytes);
```

This function returns the size of the workspace needed by `cusparseCsrmvEx`. User needs to allocate a buffer of this size and give that buffer to `cusparseCsrmvEx` as an argument. All the arguments are similar to `cusparseCsrmvEx` except the following output argument.

**Output specifically required by cusparseCsrmvEx_bufferSize**

<table>
<thead>
<tr>
<th>BufferSizeInBytes</th>
<th>Pointer to a size_t variable, which will be assigned with the size of workspace needed by cusparseCsrmv. Returned number is always more than zero, so cudaMalloc() with provided number of bytes needed would not return an error due to number of requested bytes being zero.</th>
</tr>
</thead>
</table>

---

<table>
<thead>
<tr>
<th>alphatype</th>
<th>Data type of alpha.</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrValAtype</td>
<td>Data type of csrValA.</td>
</tr>
<tr>
<td>xtype</td>
<td>Data type of x.</td>
</tr>
<tr>
<td>betatype</td>
<td>Data type of beta.</td>
</tr>
<tr>
<td>ytype</td>
<td>Data type of y.</td>
</tr>
<tr>
<td>executiontype</td>
<td>Data type used for computation.</td>
</tr>
<tr>
<td>buffer</td>
<td>Pointer to workspace buffer. Not currently used, however must be allocated and be aligned with word boundaries. Recommended to always use cusparseCsrmvEx_bufferSize to obtain the right size for this buffer.</td>
</tr>
</tbody>
</table>
7.6. cusparse<t>csrsv_m()  

This function performs a load-balanced matrix-vector operation

\[ y = \alpha \cdot \text{op}(A) \cdot x + \beta \cdot y \]

\[ A \] is an \( m \times n \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA}, \text{csrRowPtrA}, \) and \( \text{csrColIndA} \); \( x \) and \( y \) are vectors; \( \alpha \) and \( \beta \) are scalars; and

\[ \text{op}(A) = \begin{cases} A & \text{if trans} == \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\ A^T & \text{if trans} == \text{CUSPARSE_OPERATION_TRANSPOSE} \\ A^H & \text{if trans} == \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} \end{cases} \]

Note: This function is deprecated in favor of \text{cusparseCsrsvEx()} with \text{CUSPARSE_ALG_MERGE_PATH} parameter which provides same functionality with better performance.

This routine was introduced specifically to address some of the loss of performance in the regular csrsv() code due to irregular sparsity patterns. The core kernels are based
on the "MergePath" approach created by Duanne Merrill. By using this approach, we are able to provide performance independent of a sparsity pattern across data types.

Remark: \texttt{csrmv\_mp} only supports matrix type \texttt{CUSPARSE\_MATRIX\_TYPE\_GENERAL} and \texttt{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} operation.

Appendix C provides a simple example of \texttt{csrmv\_mp}.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the \texttt{cuSPARSE} library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>trans</td>
<td>the operation ( \text{op}(A) ). only support \texttt{CUSPARSE_OPERATION_NON_TRANSPOSE}.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix ( A ).</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>alpha</td>
<td>\langle \text{type} \rangle scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}. Also, the supported index bases are \texttt{CUSPARSE_INDEX_BASE_ZERO} and \texttt{CUSPARSE_INDEX_BASE_ONE}.</td>
</tr>
<tr>
<td>csrValA</td>
<td>\langle \text{type} \rangle array of ( \text{nnz} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0) ) nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of ( m+1 ) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of ( \text{nnz} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0) ) column indices of the nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>x</td>
<td>\langle \text{type} \rangle vector of ( n ) elements if ( \text{op}(A) = A ), and ( m ) elements if ( \text{op}(A) = A^T ) or ( \text{op}(A) = A^H ).</td>
</tr>
<tr>
<td>beta</td>
<td>\langle \text{type} \rangle scalar used for multiplication. If \text{beta} is zero, ( y ) does not have to be a valid input.</td>
</tr>
<tr>
<td>y</td>
<td>\langle \text{type} \rangle vector of ( m ) elements if ( \text{op}(A) = A ), and ( n ) elements if ( \text{op}(A) = A^T ) or ( \text{op}(A) = A^H ).</td>
</tr>
</tbody>
</table>

**Output**

| y | \langle \text{type} \rangle updated vector. |

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>\texttt{CUSPARSE_STATUS_SUCCESS}</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_NOT_INITIALIZED}</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_ALLOC_FAILED}</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_INVALID_VALUE}</td>
<td>invalid parameters were passed ( (m, n, \text{nnz}&lt;0) ).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision (compute capability (c.c.) &gt;= 1.3 required), symmetric/Hermitian matrix (c.c. &gt;= 1.2 required), or transpose operation (c.c. &gt;= 1.1 required).</td>
</tr>
<tr>
<td>-----------------------------</td>
<td>---------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

7.7. cusparse<t>gemvi()

cusparseStatus_t
cusparseSgemvi(cusparseHandle_t handle, cusparseOperation_t transA,
    int m, int n, Const float *alpha,
    const float *A,
    int lda, int nnz,
    const float *x,
    const int *xInd,
    const float *beta,
    float *y,
    cusparseIndexBase_t idxBase,
    void *pBuffer)

cusparseStatus_t
cusparseDgemvi(cusparseHandle_t handle, cusparseOperation_t transA,
    int m, int n, Const double *alpha,
    const double *A,
    int lda, int nnz,
    const double *x,
    const int *xInd,
    const double *beta,
    double *y,
    cusparseIndexBase_t idxBase,
    void *pBuffer)

cusparseStatus_t
cusparseCgemvi(cusparseHandle_t handle, cusparseOperation_t transA,
    int m, int n, Const cuComplex *alpha,
    const cuComplex *A,
    int lda, int nnz,
    const cuComplex *x,
    const int *xInd,
    const cuComplex *beta,
    cuComplex *y,
    cusparseIndexBase_t idxBase,
    void *pBuffer)

cusparseStatus_t
cusparseZgemvi(cusparseHandle_t handle, cusparseOperation_t transA,
    int m, int n, Const cuDoubleComplex *alpha,
    const cuDoubleComplex *A,
    int lda, int nnz,
    const cuDoubleComplex *x,
    const int *xInd,
    const cuDoubleComplex *beta,
    cuDoubleComplex *y,
    cusparseIndexBase_t idxBase,
    void *pBuffer)

This function performs the matrix-vector operation
\[ y = \alpha \cdot \text{op}(A) \cdot x + \beta \cdot y \]

\( A \) is an \( m \times n \) dense matrix and a sparse vector \( x \) that is defined in a sparse storage format by the two arrays \( xVal, xInd \) of length \( \text{nnz} \), and \( y \) is a dense vector; \( \alpha \) and \( \beta \) are scalars; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE} 
\end{cases}
\]

To simplify the implementation, we have not (yet) optimized the transpose multiple case. We recommend the following for users interested in this case.

1. Convert the matrix from CSR to CSC format using one of the \( \text{csr2csc()} \) functions. Notice that by interchanging the rows and columns of the result you are implicitly transposing the matrix.

2. Call the \( \text{gemvi()} \) function with the \( \text{cusparseOperation\_t} \) parameter set to \( \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \) and with the interchanged rows and columns of the matrix stored in CSC format. This (implicitly) multiplies the vector by the transpose of the matrix in the original CSR format.

This function requires no extra storage for the general matrices when operation \( \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \) is selected. It requires some extra storage for Hermitian/symmetric matrices and for the general matrices when an operation different than \( \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \) is selected. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>trans</td>
<td>the operation ( \text{op}(A) ).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix ( A ).</td>
</tr>
<tr>
<td>alpha</td>
<td>(&lt;\text{type}&gt; ) scalar used for multiplication.</td>
</tr>
<tr>
<td>A</td>
<td>the pointer to dense matrix ( A ).</td>
</tr>
<tr>
<td>lda</td>
<td>size of the leading dimension of ( A ).</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of vector ( x ).</td>
</tr>
<tr>
<td>x</td>
<td>(&lt;\text{type}&gt; ) sparse vector of ( \text{nnz} ) elements of size ( n ) if ( \text{op}(A) = A ), and size ( m ) if ( \text{op}(A) = A^T ) or ( \text{op}(A) = A^H ).</td>
</tr>
<tr>
<td>xInd</td>
<td>Indices of non-zero values in ( x ).</td>
</tr>
<tr>
<td>beta</td>
<td>(&lt;\text{type}&gt; ) scalar used for multiplication. If ( \text{beta} ) is zero, ( y ) does not have to be a valid input.</td>
</tr>
<tr>
<td>y</td>
<td>(&lt;\text{type}&gt; ) dense vector of ( m ) elements if ( \text{op}(A) = A ), and ( n ) elements if ( \text{op}(A) = A^T ) or ( \text{op}(A) = A^H ).</td>
</tr>
<tr>
<td>idxBase</td>
<td>0 or 1, for 0 based or 1 based indexing, respectively</td>
</tr>
<tr>
<td>pBuffer</td>
<td>working space buffer, of size given by Xgemvi_getBufferSize()</td>
</tr>
<tr>
<td>---------</td>
<td>-------------------------------------------------------------</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>y</th>
<th>&lt;type&gt; updated dense vector.</th>
</tr>
</thead>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n, nnz&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision (compute capability (c.c.) &gt;= 1.3 required), symmetric/Hermitian matrix (c.c. &gt;= 1.2 required), or transpose operation (c.c. &gt;= 1.1 required).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.8. `cusparse<t>gemvi_bufferSize()`

```c
 cusparseStatus_t
cusparseSgemvi_bufferSize(cusparseHandle_t handle,
    cusparseOperation_t transA,
    int m,
    int n,
    int nnz,
    int *pBufferSize)

cusparseStatus_t
cusparseDgemvi_bufferSize(cusparseHandle_t handle,
    cusparseOperation_t transA,
    int m,
    int n,
    int nnz,
    int *pBufferSize)

cusparseStatus_t
cusparseCgemvi_bufferSize(cusparseHandle_t handle,
    cusparseOperation_t transA,
    int m,
    int n,
    int nnz,
    int *pBufferSize)

cusparseStatus_t
cusparseZgemvi_bufferSize(cusparseHandle_t handle,
    cusparseOperation_t transA,
    int m,
    int n,
    int nnz,
    int *pBufferSize)
```

This function returns size of buffer used in `gemvi()`

**A** is an \((m) \times (n)\) dense matrix.

\[
\text{op}(A) = \begin{cases} 
  A & \text{if trans == CUSPARSE_OPERATION_NON_TRANSPOSE} \\
  A^T & \text{if trans == CUSPARSE_OPERATION_TRANSPOSE} \\
  A^H & \text{if trans == CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE}
\end{cases}
\]

**Input**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation <code>op(A)</code>.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix Y.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero entries of vector (x) multiplying A.</td>
</tr>
</tbody>
</table>

**Output**

| BufferSize | number of elements needed the buffer used in `gemvi()`. |
Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n, nnz&lt;=0)</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.9. `cusparse<t>bsrsv2_bufferSize()`

cusparseStatus_t
cusparseSbsrsv2_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
float *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrsv2Info_t info,
int *pBufferSizeInBytes);

cusparseStatus_t
cusparseDbsrsv2_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
double *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrsv2Info_t info,
int *pBufferSizeInBytes);

cusparseStatus_t
cusparseCbsrsv2_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
cuComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrsv2Info_t info,
int *pBufferSizeInBytes);

cusparseStatus_t
cusparseZbsrsv2_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
cuDoubleComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrsv2Info_t info,
int *pBufferSizeInBytes);
This function returns size of the buffer used in \texttt{bsrsv2}, a new sparse triangular linear system \( \text{op}(A) \cdot y = \alpha \cdot x \).

\( A \) is an \((mb \cdot \text{blockDim}) \times (mb \cdot \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \texttt{bsrValA}, \texttt{bsrRowPtrA}, and \texttt{bsrColIndA}; \( x \) and \( y \) are the right-hand-side and the solution vectors; \( \alpha \) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if trans} = \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if trans} = \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if trans} = \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE} 
\end{cases}
\]

Although there are six combinations in terms of parameter \texttt{trans} and the upper (lower) triangular part of \( A \), \texttt{bsrsv2\_bufferSize()} returns the maximum size buffer among these combinations. The buffer size depends on the dimensions \( mb, \text{blockDim}, \) and the number of nonzero blocks of the matrix \( \text{nnzb} \). If the user changes the matrix, it is necessary to call \texttt{bsrsv2\_bufferSize()} again to have the correct buffer size; otherwise a segmentation fault may occur.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>dirA</td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation ( \text{op}(A) ).</td>
</tr>
<tr>
<td>mb</td>
<td>number of block rows of matrix ( A ).</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types are CUSPARSE_DIAG_TYPE_UNIT and CUSPARSE_DIAG_TYPE_NON_UNIT.</td>
</tr>
<tr>
<td>bsrValA</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnzb} = (\text{bsrRowPtrA}(mb) - \text{bsrRowPtrA}(0)) ) nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>integer array of ( \text{mb} + 1 ) elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td>bsrColIndA</td>
<td>integer array of ( \text{nnzb} = (\text{bsrRowPtrA}(mb) - \text{bsrRowPtrA}(0)) ) column indices of the nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>blockDim</td>
<td>block dimension of sparse matrix ( A ); must be larger than zero.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>info</th>
<th>record of internal states based on different algorithms.</th>
</tr>
</thead>
<tbody>
<tr>
<td>pBufferSizeInBytes</td>
<td>number of bytes of the buffer used in the bsrsv2_analysis() and bsrsv2_solve().</td>
</tr>
</tbody>
</table>

**Status Returned**
<table>
<thead>
<tr>
<th>cuSPARSE_Status</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (mb, nnzb&lt;=0), base index is not 0 or 1.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.10. cusparse<t>bsrsv2_analysis()

cusparseStatus_t
 cusparseSbsrsv2_analysis(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const cusparseMatDescr_t descrA,
  const float *bsrValA,
  const int *bsrRowPtrA,
  const int *bsrColIndA,
  int blockDim,
  bsrsv2Info_t info,
  cusparseSolvePolicy_t policy,
  void *pBuffer);

cusparseStatus_t
 cusparseDbsrsv2_analysis(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const cusparseMatDescr_t descrA,
  const double *bsrValA,
  const int *bsrRowPtrA,
  const int *bsrColIndA,
  int blockDim,
  bsrsv2Info_t info,
  cusparseSolvePolicy_t policy,
  void *pBuffer);

cusparseStatus_t
 cusparseCbsrsv2_analysis(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const cusparseMatDescr_t descrA,
  const cuComplex *bsrValA,
  const int *bsrRowPtrA,
  const int *bsrColIndA,
  int blockDim,
  bsrsv2Info_t info,
  cusparseSolvePolicy_t policy,
  void *pBuffer);

cusparseStatus_t
 cusparseZbsrsv2_analysis(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const cusparseMatDescr_t descrA,
  const cuDoubleComplex *bsrValA,
  const int *bsrRowPtrA,
  const int *bsrColIndA,
  int blockDim,
  bsrsv2Info_t info,
  cusparseSolvePolicy_t policy,
  void *pBuffer);
This function performs the analysis phase of \texttt{bsrsv2}, a new sparse triangular linear system $\text{op}(A) * y = \alpha x$.

$A$ is an $(mb \times \text{blockDim}) \times (mb \times \text{blockDim})$ sparse matrix that is defined in BSR storage format by the three arrays \texttt{bsrValA}, \texttt{bsrRowPtrA}, and \texttt{bsrColIndA}; $x$ and $y$ are the right-hand side and the solution vectors; $\alpha$ is a scalar; and

$$\text{op}(A) = \begin{cases} A & \text{if trans} = = \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\ A^T & \text{if trans} = = \text{CUSPARSE_OPERATION_TRANSPOSE} \\ A^H & \text{if trans} = = \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} \end{cases}$$

The block of BSR format is of size $\text{blockDim} \times \text{blockDim}$, stored as column-major or row-major as determined by parameter \texttt{dirA}, which is either \texttt{CUSPARSE_DIRECTION_COLUMN} or \texttt{CUSPARSE_DIRECTION_ROW}. The matrix type must be \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, and the fill mode and diagonal type are ignored.

It is expected that this function will be executed only once for a given matrix and a particular operation type.

This function requires a buffer size returned by \texttt{bsrsv2_bufferSize()}. The address of \texttt{pBuffer} must be multiple of 128 bytes. If it is not, \texttt{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Function \texttt{bsrsv2_analysis()} reports a structural zero and computes level information, which stored in the opaque structure \texttt{info}. The level information can extract more parallelism for a triangular solver. However \texttt{bsrsv2_solve()} can be done without level information. To disable level information, the user needs to specify the policy of the triangular solver as \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}.

Function \texttt{bsrsv2_analysis()} always reports the first structural zero, even when parameter \texttt{policy} is \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}. No structural zero is reported if \texttt{CUSPARSE_DIAG_TYPE_UNIT} is specified, even if block $A(j,j)$ is missing for some $j$. The user needs to call \texttt{cusparseXbsrsv2_zeroPivot()} to know where the structural zero is.

It is the user's choice whether to call \texttt{bsrsv2_solve()} if \texttt{bsrsv2_analysis()} reports a structural zero. In this case, the user can still call \texttt{bsrsv2_solve()}, which will return a numerical zero at the same position as a structural zero. However the result $x$ is meaningless.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>dirA</td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation $\text{op}(A)$.</td>
</tr>
<tr>
<td>mb</td>
<td>number of block rows of matrix $A$.</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of nonzero blocks of matrix $A$.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types</td>
</tr>
</tbody>
</table>
**bsrValA**

<type> array of \( nnzb = bsrRowPtrA(mb) - bsrRowPtrA(0) \) nonzero blocks of matrix \( A \).

**bsrRowPtrA**

integer array of \( mb + 1 \) elements that contains the start of every block row and the end of the last block row plus one.

**bsrColIndA**

integer array of \( nnzb = bsrRowPtrA(mb) - bsrRowPtrA(0) \) column indices of the nonzero blocks of matrix \( A \).

**blockDim**

block dimension of sparse matrix \( A \), larger than zero.

**info**

structure initialized using `cusparseCreateBsrsv2Info()`.

**policy**

the supported policies are `CUSPARSE_SOLVE_POLICY_NO_LEVEL` and `CUSPARSE_SOLVE_POLICY_USE_LEVEL`.

**pBuffer**

buffer allocated by the user, the size is return by `bsrsv2_bufferSize()`.

**Output**

**info**

structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged).

**Status Returned**

- **CUSPARSE_STATUS_SUCCESS**
  the operation completed successfully.

- **CUSPARSE_STATUS_NOT_INITIALIZED**
  the library was not initialized.

- **CUSPARSE_STATUS_ALLOC_FAILED**
  the resources could not be allocated.

- **CUSPARSE_STATUS_INVALID_VALUE**
  invalid parameters were passed (\( mb, nnzb <= 0 \)).

- **CUSPARSE_STATUS_ARCH_MISMATCH**
  the device only supports compute capability 2.0 and above.

- **CUSPARSE_STATUS_EXECUTION_FAILED**
  the function failed to launch on the GPU.

- **CUSPARSE_STATUS_INTERNAL_ERROR**
  an internal operation failed.

- **CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED**
  the matrix type is not supported.
7.11. cusparse<t>bsrsv2_solve()

cusparseStatus_t
cusparseSbsrsv2_solve(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const float *alpha,
  const cusparseMatDescr_t descrA,
  const float *bsrValA,
  const int *bsrRowPtrA,
  const int *bsrColIndA,
  int blockDim,
  bsrsv2Info_t info,
  const float *x,
  float *y,
  cusparseSolvePolicy_t policy,
  void *pBuffer);

cusparseStatus_t
cusparseDbsrsv2_solve(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const double *alpha,
  const cusparseMatDescr_t descrA,
  const double *bsrValA,
  const int *bsrRowPtrA,
  const int *bsrColIndA,
  int blockDim,
  bsrsv2Info_t info,
  const double *x,
  double *y,
  cusparseSolvePolicy_t policy,
  void *pBuffer);

cusparseStatus_t
cusparseCbsrsv2_solve(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const cuComplex *alpha,
  const cusparseMatDescr_t descrA,
  const cuComplex *bsrValA,
  const int *bsrRowPtrA,
  const int *bsrColIndA,
  int blockDim,
  bsrsv2Info_t info,
  const cuComplex *x,
  cuComplex *y,
  cusparseSolvePolicy_t policy,
  void *pBuffer);

cusparseStatus_t
cusparseZbsrsv2_solve(cusparseHandle_t handle,
  cusparseDirection_t dirA,
  cusparseOperation_t transA,
  int mb,
  int nnzb,
  const cuDoubleComplex *alpha,
  const cusparseMatDescr_t descrA,
  const cuDoubleComplex *bsrValA,
This function performs the solve phase of \texttt{bsrsv2}, a new sparse triangular linear system 
\[ \text{op}(A) y = \alpha x. \]
\( A \) is an \((mb\times\text{blockDim})\times(mb\times\text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \texttt{bsrValA}, \texttt{bsrRowPtrA}, and \texttt{bsrColIndA}; \( x \) and \( y \) are the right-hand-side and the solution vectors; \( \alpha \) is a scalar; and 
\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE} 
\end{cases}
\]
The block in BSR format is of size \texttt{blockDim*blockDim}, stored as column-major or row-major as determined by parameter \texttt{dirA}, which is either \texttt{CUSPARSE\_DIRECTION\_COLUMN} or \texttt{CUSPARSE\_DIRECTION\_ROW}. The matrix type must be \texttt{CUSPARSE\_MATRIX\_TYPE\_GENERAL}, and the fill mode and diagonal type are ignored.
Function \texttt{bsrsv02\_solve()} can support an arbitrary \texttt{blockDim}.

This function may be executed multiple times for a given matrix and a particular operation type.

This function requires a buffer size returned by \texttt{bsrsv2\_bufferSize()}. The address of \texttt{pBuffer} must be multiple of 128 bytes. If it is not, \texttt{CUSPARSE\_STATUS\_INVALID\_VALUE} is returned.

Although \texttt{bsrsv2\_solve()} can be done without level information, the user still needs to be aware of consistency. If \texttt{bsrsv2\_analysis()} is called with policy \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL}, \texttt{bsrsv2\_solve()} can be run with or without levels. On the other hand, if \texttt{bsrsv2\_analysis()} is called with \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}, \texttt{bsrsv2\_solve()} can only accept \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}; otherwise, \texttt{CUSPARSE\_STATUS\_INVALID\_VALUE} is returned.

The level information may not improve the performance, but may spend extra time doing analysis. For example, a tridiagonal matrix has no parallelism. In this case, \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} performs better than \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL}. If the user has an iterative solver, the best approach is to do \texttt{bsrsv2\_analysis()} with \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} once. Then do \texttt{bsrsv2\_solve()} with \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} in the first run, and with \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} in the second run, and pick the fastest one to perform the remaining iterations.

Function \texttt{bsrsv02\_solve()} has the same behavior as \texttt{csrsv02\_solve()}. That is, \texttt{bsr2csr(bsrsv02}(A)) = \texttt{csrsv02(bsr2csr(A))}. The numerical zero of \texttt{csrsv02\_solve()} means there exists some zero \( A(j,j) \). The numerical zero of \texttt{bsrsv02\_solve()} means there exists some block \( A(j,j) \) that is not invertible.

Function \texttt{bsrsv2\_solve()} reports the first numerical zero, including a structural zero. No numerical zero is reported if \texttt{CUSPARSE\_DIAG\_TYPE\_UNIT} is specified, even if \( A(j,j) \) is not invertible for some \( j \). The user needs to call \texttt{cusparseXbsrsv2\_zeroPivot()} to know where the numerical zero is.
For example, suppose $L$ is a lower triangular matrix with unit diagonal, then the following code solves $L \cdot y = x$ by level information.

```c
// Suppose that $L$ is an $m \times m$ sparse matrix represented by BSR format,
// The number of block rows/columns is $mb$, and
// the number of nonzero blocks is $nnzb$.
// $L$ is lower triangular with unit diagonal.
// Assumption:
// - dimension of matrix $L$ is $m(=mb*blockDim)$,
// - matrix $L$ has $nnz(=nnzb*blockDim*blockDim)$ nonzero elements,
// - handle is already created by cusparseCreate(),
// - $(d_bsrRowPtr, d_bsrColInd, d_bsrVal)$ is BSR of $L$ on device memory,
// - $d_y$ is right hand side vector on device memory.
// - $d_x$ and $d_y$ are of size $m$.

cusparseMatDescr_t descr = 0;
bsrsv2Info_t info = 0;
int pBufferSize;
void *pBuffer = 0;
int structural_zero;
int numerical_zero;
const double alpha = 1.1;
const cusparseSolvePolicy_t policy = CUSPARSE_SOLVE_POLICY_USE_LEVEL;
const cusparseOperation_t trans = CUSPARSE_OPERATION_NON_TRANSPOSE;
const cusparseDirection_t dir = CUSPARSE_DIRECTION_COLUMN;

// step 1: create a descriptor which contains
// - matrix $L$ is base-1
// - matrix $L$ is lower triangular
// - matrix $L$ has unit diagonal, specified by parameter CUSPARSE_DIAG_TYPE_UNIT
//   ($L$ may not have all diagonal elements.)
cusparseCreateMatDescr(&descr);
cusparseSetMatIndexBase(descr, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatFillMode(descr, CUSPARSE_FILL_MODE_LOWER);
cusparseSetMatDiagType(descr, CUSPARSE_DIAG_TYPE_UNIT);

// step 2: create an empty info structure
cusparseCreateBsrsv2Info(&info);

// step 3: query how much memory used in bsrsv2, and allocate the buffer
cusparseBsrsv2_bufferSize(handle, dir, trans, mb, nnzb, descr,
    d_bsrVal, d_bsrRowPtr, d_bsrColInd, blockDim, &pBufferSize);

// pBuffer returned by cudaMalloc is automatically aligned to 128 bytes.
cudaMalloc((void**)&pBuffer, pBufferSize);

// step 4: perform analysis
cusparseBsrsv2_analysis(handle, dir, trans, mb, nnzb, descr,
    d_bsrVal, d_bsrRowPtr, d_bsrColInd, blockDim, info, policy, pBuffer);

// L has unit diagonal, so no structural zero is reported.
status = cusparseXbsrsv2_zeroPivot(handle, info, &structural_zero);
if (CUSPARSE_STATUS_ZERO_PIVOT == status){
    printf("L(\%d,\%d) is missing\n", structural_zero, structural_zero);
}

// step 5: solve $L \cdot y = x$
cusparseBsrsv2_solve(handle, dir, trans, mb, nnzb, &alpha, descr,
    d_bsrVal, d_bsrRowPtr, d_bsrColInd, blockDim, info,
    d_x, d_y, policy, pBuffer);

// L has unit diagonal, so no numerical zero is reported.
status = cusparseXbsrsv2_zeroPivot(handle, info, &numerical_zero);
if (CUSPARSE_STATUS_ZERO_PIVOT == status){
    printf("L(\%d,\%d) is zero\n", numerical_zero, numerical_zero);
}

// step 6: free resources
cudaFree(pBuffer);
cusparseDestroyBsrsv2Info(info);
cusparseDestroyMatDescr(descr);
cusparseDestroy(handle);
```
Input

| handle      | handle to the cuSPARSE library context. |
| dirA        | storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN. |
| transA      | the operation op(A). |
| mb          | number of block rows and block columns of matrix A. |
| alpha       | <type> scalar used for multiplication. |
| descrA      | the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types are CUSPARSE_DIAG_TYPE_UNIT and CUSPARSE_DIAG_TYPE_NON_UNIT. |
| bsrValA     | <type> array of nnzb (= bsrRowPtrA(mb) - bsrRowPtrA(0) ) nonzero blocks of matrix A. |
| bsrRowPtrA  | integer array of mb + 1 elements that contains the start of every block row and the end of the last block row plus one. |
| bsrColIndA  | integer array of nnzb (= bsrRowPtrA(mb) - bsrRowPtrA(0) ) column indices of the nonzero blocks of matrix A. |
| blockDim   | block dimension of sparse matrix A, larger than zero. |
| info        | structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged). |
| x           | <type> right-hand-side vector of size m. |
| policy      | the supported policies are CUSPARSE_SOLVE_POLICY_NO_LEVEL and CUSPARSE_SOLVE_POLICY_USE_LEVEL. |
| pBuffer     | buffer allocated by the user, the size is returned by bsrsv2_bufferSize(). |

Output

| y           | <type> solution vector of size m. |

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (mb,nnzb&lt;=0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MAPPING_ERROR</td>
<td>the texture binding failed.</td>
</tr>
</tbody>
</table>
CUSPARSE_STATUS_EXECUTION_FAILED | the function failed to launch on the GPU.
CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed.
CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported.

7.12. cusparseXbsrsv2_zeroPivot()

cusparseStatus_t

cusparseXbsrsv2_zeroPivot(cusparseHandle_t handle,
                           bsrsv2Info_t info,
                           int *position);

If the returned error code is CUSPARSE_STATUS_ZERO_PIVOT, position=j means
A(j,j) is either structural zero or numerical zero (singular block). Otherwise
position=-1.

The position can be 0-based or 1-based, the same as the matrix.

Function cusparseXbsrsv2_zeroPivot() is a blocking call. It calls
cudaDeviceSynchronize() to make sure all previous kernels are done.

The position can be in the host memory or device memory. The user can set the proper
mode with cusparseSetPointerMode().

Input

| handle | handle to the cuSPARSE library context. |
| info   | info contains a structural zero or numerical zero if the user already called bsrsv2_analysis() or bsrsv2_solve(). |

Output

| position | if no structural or numerical zero, position is -1; otherwise if A(j,j) is missing or U(j,j) is zero, position=j. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully. |
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized. |
| CUSPARSE_STATUS_INVALID_VALUE | info is not valid. |
| CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed. |
7.13. cusparse<t>csrsv_analysis()

This function performs the analysis phase of the solution of a sparse triangular linear system

\[ \text{op}(A) \times y = \alpha \times x \]

where \( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA, csrRowPtrA, and csrColIndA} \); \( x \) and \( y \) are the right-hand-side and the solution vectors; \( \alpha \) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if trans} == \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
A^T & \text{if trans} == \text{CUSPARSE_OPERATION_TRANSPOSE} \\
A^H & \text{if trans} == \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE}
\end{cases}
\]
The routine `csrsv_analysis` supports analysis phase of `csrsv_solve`, `csric0` and `csrilu0`. The user has to be careful of which routine is called after `csrsv_analysis`. The matrix descriptor must be the same for `csrsv_analysis` and its subsequent call to `csrsv_solve`, `csric0` and `csrilu0`.

For `csrsv_solve`, the matrix type must be `CUSPARSE_MATRIX_TYPE_TRIANGULAR` or `CUSPARSE_MATRIX_TYPE_GENERAL`.

For `csrilu0`, the matrix type must be `CUSPARSE_MATRIX_TYPE_GENERAL`.

For `csric0`, the matrix type must be `CUSPARSE_MATRIX_TYPE_SYMMETRIC` or `CUSPARSE_MATRIX_TYPE_HERMITIAN`.

It is expected that this function will be executed only once for a given matrix and a particular operation type.

This function requires a significant amount of extra storage that is proportional to the matrix size. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

### Input

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>trans</td>
<td>the operation op(A)</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix types are <code>CUSPARSE_MATRIX_TYPE_TRIANGULAR</code> and <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> for <code>csrsv_solve</code>, <code>CUSPARSE_MATRIX_TYPE_SYMMETRIC</code> and <code>CUSPARSE_MATRIX_TYPE_HERMITIAN</code> for <code>csric0</code>, <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> for <code>csrilu0</code>, while the supported diagonal types are <code>CUSPARSE_DIAG_TYPE_UNIT</code> and <code>CUSPARSE_DIAG_TYPE_NON_UNIT</code>.</td>
</tr>
<tr>
<td>csrValA</td>
<td>&lt;type&gt; array of nnz (= csrRowPtrA(m) - csrRowPtrA(0) ) nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of m + 1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of nnz (= csrRowPtrA(m) - csrRowPtrA(0) ) column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>info</td>
<td>structure initialized using <code>cusparseCreateSolveAnalysisInfo</code>.</td>
</tr>
</tbody>
</table>

### Output

| info | structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged). |
Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, nnz&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

7.14. cusparseCsrsv_analysisEx()

cusparseStatus_t cusparseCsrsv_analysisEx(cusparseHandle_t handle,
cusparseOperation_t transA,
int m, int nnz,
const cusparseMatDescr_t descrA,
const void *csrSortedValA,
cudaDataType csrSortedValAtype,
const int *csrSortedRowPtrA,
const int *csrSortedColIndA,
cusparseSolveAnalysisInfo_t info,
cudaDataType executiontype);

This function is an extended version of cusparse<t>csrsv_analysis(). For detailed description of the functionality, see cusparse<t>csrsv_analysis().

This function does not support half-precision execution type, but it supports half-precision IO with single precision execution.

Input specifically required by cusparseCsrsv_analysisEx

<table>
<thead>
<tr>
<th>csrSortedValAtype</th>
<th>Data type of csrSortedValA.</th>
</tr>
</thead>
<tbody>
<tr>
<td>executiontype</td>
<td>Data type used for computation.</td>
</tr>
</tbody>
</table>
7.15. `cusparse<t>csrsv_solve()`

```c
int *csrSortedColIndA,
```

This function performs the solve phase of the solution of a sparse triangular linear system

\[ \text{op}(A) \times x = \alpha \times f \]

where \( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrSortedValA}, \text{csrSortedRowPtrA}, \) and \( \text{csrSortedColIndA} \); \( f \) and \( x \) are the right-hand-side and the solution vectors; \( \alpha \) is a scalar; and
This function may be executed multiple times for a given matrix and a particular operation type.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>trans</td>
<td>the operation $\text{op}(A)$</td>
</tr>
<tr>
<td>m</td>
<td>number of rows and columns of matrix $A$.</td>
</tr>
<tr>
<td>alpha</td>
<td>&lt;type&gt; scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix types are CUSPARSE_MATRIX_TYPE_TRIANGULAR and CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types are CUSPARSE_DIAG_TYPE_UNIT and CUSPARSE_DIAG_TYPE_NON_UNIT.</td>
</tr>
<tr>
<td>csrSortedValA</td>
<td>&lt;type&gt; array of $\text{nnz} = \text{csrSortedRowPtrA}(m) - \text{csrSortedRowPtrA}(0)$ nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>csrSortedRowPtrA</td>
<td>integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrSortedColIndA</td>
<td>integer array of $\text{nnz} = \text{csrSortedRowPtrA}(m) - \text{csrSortedRowPtrA}(0)$ column indices of the nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>info</td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
<tr>
<td>f</td>
<td>&lt;type&gt; right-hand-side vector of size $m$.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>x</td>
<td>&lt;type&gt; solution vector of size $m$.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ($m&lt;0$).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MAPPING_ERROR</td>
<td>the texture binding failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
7.16. `cusparseCsrsv_solveEx()`

```c
cusparseStatus_t cusparseCsrsv_solveEx(cusparseHandle_t handle, 
cusparseOperation_t transA, 
int m, 
const void *alpha, cudaDataType
alphatype,
const cusparseMatDescr_t descrA, 
const void *csrSortedValA, 
cudaDataType csrSortedValAtype,
const int *csrSortedRowPtrA, 
const int *csrSortedColIndA, 
cusparseSolveAnalysisInfo_t info,
const void *f, cudaDataType
ftype,
void *x, cudaDataType xtype,
cudaDataType executiontype);
```

This function is an extended version of `cusparse<t>csrsv_solve()`. For detailed description of the functionality, see `cusparse<t>csrsv_solve()`.

This function does not support half-precision execution type, but it supports half-precision IO with single precision execution.

**Input specifically required by cusparseCsrsv_solveEx**

<table>
<thead>
<tr>
<th><strong>alphatype</strong></th>
<th>Data type of alpha.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>csrSortedValAtype</strong></td>
<td>Data type of csrSortedValA.</td>
</tr>
<tr>
<td><strong>ftype</strong></td>
<td>Data type of f.</td>
</tr>
<tr>
<td><strong>xtype</strong></td>
<td>Data type of x.</td>
</tr>
<tr>
<td><strong>executiontype</strong></td>
<td>Data type used for computation.</td>
</tr>
</tbody>
</table>
7.17. cusparse<t>csrsv2_bufferSize()

This function returns the size of the buffer used in csrsv2, a new sparse triangular linear system \( \mathbf{op}(\mathbf{A}) \mathbf{y} = \mathbf{a} \mathbf{x} \).

\( \mathbf{A} \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( \mathbf{csrValA} \), \( \mathbf{csrRowPtrA} \), and \( \mathbf{csrColIndA} \); \( \mathbf{x} \) and \( \mathbf{y} \) are the right-hand-side and the solution vectors; \( \mathbf{a} \) is a scalar; and
op(A) = \[
\begin{cases} 
A & \text{if trans} == \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
A^T & \text{if trans} == \text{CUSPARSE_OPERATION_TRANSPOSE} \\
A^H & \text{if trans} == \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE}
\end{cases}
\]

Although there are six combinations in terms of the parameter `trans` and the upper (lower) triangular part of A, `csrsv2_bufferSize()` returns the maximum size buffer of these combinations. The buffer size depends on the dimension and the number of nonzero elements of the matrix. If the user changes the matrix, it is necessary to call `csrsv2_bufferSize()` again to have the correct buffer size; otherwise, a segmentation fault may occur.

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation op(A).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types are CUSPARSE_DIAG_TYPE_UNIT and CUSPARSE_DIAG_TYPE_NON_UNIT.</td>
</tr>
<tr>
<td>csrValA</td>
<td><code>&lt;type&gt;</code> array of nnz (= csrRowPtrA(m) - csrRowPtrA(0) ) nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of m + 1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of nnz (= csrRowPtrA(m) - csrRowPtrA(0) ) column indices of the nonzero elements of matrix A.</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>record of internal states based on different algorithms.</td>
</tr>
<tr>
<td>pBufferSizesInBytes</td>
<td>number of bytes of the buffer used in the csrsv2_analysis and csrsv2_solve.</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, nnz&lt;0), base index is not 0 or 1.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.18. 

This function performs the analysis phase of \texttt{csrsrv2}, a new sparse triangular linear system \( \text{op}(A) \mathbf{y} = \mathbf{A} \mathbf{x} \).
A is an $m \times m$ sparse matrix that is defined in CSR storage format by the three arrays $\text{csrValA}$, $\text{csrRowPtrA}$, and $\text{csrColIndA}$; $x$ and $y$ are the right-hand-side and the solution vectors; $\alpha$ is a scalar; and

$$\text{op}(A) = \begin{cases} A & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\ A^T & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_TRANSPOSE} \\ A^H & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} \end{cases}$$

It is expected that this function will be executed only once for a given matrix and a particular operation type.

This function requires a buffer size returned by $\text{csrs2_bufferSize()}$. The address of pBuffer must be multiple of 128 bytes. If it is not, $\text{CUSPARSE_STATUS_INVALID_VALUE}$ is returned.

Function $\text{csrs2_analysis()}$ reports a structural zero and computes level information that is stored in opaque structure info. The level information can extract more parallelism for a triangular solver. However $\text{csrs2_solve()}$ can be done without level information. To disable level information, the user needs to specify the policy of the triangular solver as $\text{CUSPARSE_SOLVE_POLICY_NO_LEVEL}$.

Function $\text{csrs2_analysis()}$ always reports the first structural zero, even if the policy is $\text{CUSPARSE_SOLVE_POLICY_NO_LEVEL}$. No structural zero is reported if $\text{CUSPARSE_DIAG_TYPE_UNIT}$ is specified, even if $A(j,j)$ is missing for some $j$. The user needs to call $\text{cusparseXcsrs2_zeroPivot()}$ to know where the structural zero is.

It is the user’s choice whether to call $\text{csrs2_solve()}$ if $\text{csrs2_analysis()}$ reports a structural zero. In this case, the user can still call $\text{csrs2_solve()}$ which will return a numerical zero in the same position as the structural zero. However the result $x$ is meaningless.

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation $\text{op}(A)$.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix $A$.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix type is $\text{CUSPARSE_MATRIX_TYPE_GENERAL}$, while the supported diagonal types are $\text{CUSPARSE_DIAG_TYPE_UNIT}$ and $\text{CUSPARSE_DIAG_TYPE_NON_UNIT}$.</td>
</tr>
<tr>
<td>csrValA</td>
<td>$&lt;\text{type}&gt;$ array of $\text{nnz} =$ $\text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)$ nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of $\text{nnz} =$ $\text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)$ column indices of the nonzero elements of matrix $A$.</td>
</tr>
</tbody>
</table>
### info
- structure initialized using `cusparseCreateCsrsv2Info()`.

### policy
- The supported policies are `CUSPARSE_SOLVE_POLICY_NO_LEVEL` and `CUSPARSE_SOLVE_POLICY_USE_LEVEL`.

### pBuffer
- buffer allocated by the user, the size is returned by `csrsv2_bufferSize()`.

## Output

| info | structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged). |

## Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, nnz&lt;=0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.19. `cusparse<t>csrsv2_solve`()

```c
cusparseStatus_t
cusparseScsrsv2_solve(cusparseHandle_t handle,
                      cusparseOperation_t transA,
                      int m,
                      int nnz,
                      const float *alpha,
                      const cusparseMatDescr_t descra,
                      const float *csrValA,
                      const int *csrRowPtrA,
                      const int *csrColIndA,
                      csrsv2Info_t info,
                      float *x,
                      float *y,
                      cusparseSolvePolicy_t policy,
                      void *pBuffer);

cusparseStatus_t
cusparseDcsrsv2_solve(cusparseHandle_t handle,
                      cusparseOperation_t transA,
                      int m,
                      int nnz,
                      const double *alpha,
                      const cusparseMatDescr_t descra,
                      const double *csrValA,
                      const int *csrRowPtrA,
                      const int *csrColIndA,
                      csrsv2Info_t info,
                      const double *x,
                      double *y,
                      cusparseSolvePolicy_t policy,
                      void *pBuffer);

cusparseStatus_t
cusparseCcsrsv2_solve(cusparseHandle_t handle,
                      cusparseOperation_t transA,
                      int m,
                      int nnz,
                      const cuComplex *alpha,
                      const cusparseMatDescr_t descra,
                      const cuComplex *csrValA,
                      const int *csrRowPtrA,
                      const int *csrColIndA,
                      csrsv2Info_t info,
                      const cuComplex *x,
                      cuComplex *y,
                      cusparseSolvePolicy_t policy,
                      void *pBuffer);

cusparseStatus_t
cusparseZcsrsv2_solve(cusparseHandle_t handle,
                      cusparseOperation_t transA,
                      int m,
                      int nnz,
                      const cuDoubleComplex *alpha,
                      const cusparseMatDescr_t descra,
                      const cuDoubleComplex *csrValA,
                      const int *csrRowPtrA,
                      const int *csrColIndA,
                      csrsv2Info_t info,
                      const cuDoubleComplex *x,
                      cuDoubleComplex *y,
                      cusparseSolvePolicy_t policy,
                      void *pBuffer);
```
This function performs the solve phase of \texttt{csrsv2}, a new sparse triangular linear system 
\[\text{op}(A) * y = \alpha x.\]

\(A\) is an \(m \times m\) sparse matrix that is defined in CSR storage format by the three arrays \texttt{csrValA}, \texttt{csrRowPtrA}, and \texttt{csrColIndA}; \(x\) and \(y\) are the right-hand-side and the solution vectors; \(\alpha\) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE}} \\
A^T & \text{if \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE}} \\
A^H & \text{if \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE}}
\end{cases}
\]

This function may be executed multiple times for a given matrix and a particular operation type.

This function requires the buffer size returned by \texttt{csrsv2\_bufferSize()}. The address of \texttt{pBuffer} must be multiple of 128 bytes. If it is not, \texttt{CUSPARSE\_STATUS\_INVALID\_VALUE} is returned.

Although \texttt{csrsv2\_solve()} can be done without level information, the user still needs to be aware of consistency. If \texttt{csrsv2\_analysis()} is called with policy \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL}, \texttt{csrsv2\_solve()} can be run with or without levels. On the contrary, if \texttt{csrsv2\_analysis()} is called with \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}, \texttt{csrsv2\_solve()} can only accept \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}; otherwise, \texttt{CUSPARSE\_STATUS\_INVALID\_VALUE} is returned.

The level information may not improve the performance but spend extra time doing analysis. For example, a tridiagonal matrix has no parallelism. In this case, \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} performs better than \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL}. If the user has an iterative solver, the best approach is to do \texttt{csrsv2\_analysis()} with \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} once. Then do \texttt{csrsv2\_solve()} with \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} in the first run and with \texttt{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} in the second run, picking faster one to perform the remaining iterations.

Function \texttt{csrsv2\_solve()} reports the first numerical zero, including a structural zero. If \texttt{status} is 0, no numerical zero was found. Furthermore, no numerical zero is reported if \texttt{CUSPARSE\_DIAG\_TYPE\_UNIT} is specified, even if \(A(j, j)\) is zero for some \(j\). The user needs to call \texttt{cusparseXcsrsv2\_zeroPivot()} to know where the numerical zero is.
For example, suppose \( L \) is a lower triangular matrix with unit diagonal, the following code solves \( L\mathbf{y} = \mathbf{x} \) by level information.

```c
// Suppose that \( L \) is a \( m \times m \) sparse matrix represented by CSR format,
// \( L \) is lower triangular with unit diagonal.
// Assumption:
// - dimension of matrix \( L \) is \( m \),
// - matrix \( L \) has \( nnz \) number zero elements,
// - handle is already created by \( \text{cusparseCreate}() \),
// - \( (d\_\text{csrRowPtr}, d\_\text{csrColInd}, d\_\text{csrVal}) \) is CSR of \( L \) on device memory,
// - \( d\_x \) is right hand side vector on device memory,
// - \( d\_y \) is solution vector on device memory.

cusparseMatDescr_t descr = 0;
csrsv2Info_t info = 0;
int pBufferSize;
void *pBuffer = 0;
int structural_zero;
int numerical_zero;
const double alpha = 1.;
const cusparseSolvePolicy_t policy = CUSPARSE_SOLVE_POLICY_USE_LEVEL;
const cusparseOperation_t trans = CUSPARSE_OPERATION_NON_TRANSPOSE;

// step 1: create a descriptor which contains
// - matrix \( L \) is base-1
// - matrix \( L \) is lower triangular
// - matrix \( L \) has unit diagonal, specified by parameter CUSPARSE_DIAG_TYPE_UNIT
// (L may not have all diagonal elements.)
cusparseCreateMatDescr(&descr);
cusparseSetMatIndexBase(descr, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatFillMode(descr, CUSPARSE_FILL_MODE_LOWER);
cusparseSetMatDiagType(descr, CUSPARSE_DIAG_TYPE_UNIT);

// step 2: create an empty info structure

cusparseCreateCsrsv2Info(&info);

// step 3: query how much memory used in csrsv2, and allocate the buffer

cusparseDcsrsv2_bufferSize(handle, trans, m, nnz, descr,
                                        d_csrVal, d_csrRowPtr, d_csrColInd, &pBufferSize);
// pBuffer returned by cudaMalloc is automatically aligned to 128 bytes.
cudaMalloc((void**)&pBuffer, pBufferSize);

// step 4: perform analysis

cusparseDcsrsv2_analysis(handle, trans, m, nnz, descr,
                                      d_csrVal, d_csrRowPtr, d_csrColInd, info, policy, pBuffer);
// L has unit diagonal, so no structural zero is reported.
status = cusparseXcsrsv2_zeroPivot(handle, info, &structural_zero);
if (CUSPARSE_STATUS_ZERO_PIVOT == status){
    printf("L(%d,%d) is missing\n", structural_zero, structural_zero);
}

// step 5: solve \( L\mathbf{y} = \mathbf{x} \)
cusparseDcsrsv2_solve(handle, trans, m, nnz, &alpha, descr,
                                      d_csrVal, d_csrRowPtr, d_csrColInd, info,
                                      d_x, d_y, policy, pBuffer);
// L has unit diagonal, so no numerical zero is reported.
status = cusparseXcsrsv2_zeroPivot(handle, info, &numerical_zero);
if (CUSPARSE_STATUS_ZERO_PIVOT == status){
    printf("L(%d,%d) is zero\n", numerical_zero, numerical_zero);
}

// step 6: free resources

cudaFree(pBuffer);
cusparseDestroyCsrsv2Info(info);
cusparseDestroyMatDescr(descr);
cusparseDestroy(handle);
```
Remark: `csrsv2_solve()` needs more nonzeros per row to achieve good performance. It would perform better if more than 16 nonzeros per row in average.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation $\text{op}(A)$.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows and columns of matrix $A$.</td>
</tr>
<tr>
<td>alpha</td>
<td>$\text{&lt;type&gt; scalar used for multiplication.}$</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix type is \text{CUSPARSE_MATRIX_TYPE_GENERAL}, while the supported diagonal types are \text{CUSPARSE_DIAG_TYPE_UNIT} and \text{CUSPARSE_DIAG_TYPE_NON_UNIT}.</td>
</tr>
<tr>
<td>csrValA</td>
<td>$\text{&lt;type&gt; array of } nnz (= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0))$ nonzeros of matrix $A$.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of $nnz (= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0))$ column indices of the nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>info</td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
<tr>
<td>x</td>
<td>$\text{&lt;type&gt; right-hand-side vector of size } m$.</td>
</tr>
<tr>
<td>policy</td>
<td>The supported policies are \text{CUSPARSE_SOLVE_POLICY_NO_LEVEL} and \text{CUSPARSE_SOLVE_POLICY_USE_LEVEL}.</td>
</tr>
<tr>
<td>pBuffer</td>
<td>buffer allocated by the user, the size is return by \text{csrsv2_bufferSize}.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>y</td>
<td>$\text{&lt;type&gt; solution vector of size } m$.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ($m, nnz &lt; 0$).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MAPPING_ERROR</td>
<td>the texture binding failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.20. cusparseXcsrsv2_zeroPivot()

cusparseStatus_t cusparseXcsrsv2_zeroPivot(cusparseHandle_t handle,
    csrsv2Info_t info,
    int *position);

If the returned error code is CUSPARSE_STATUS_ZERO_PIVOT, position = j means
A(j,j) has either a structural zero or a numerical zero. Otherwise position = -1.

The position can be 0-based or 1-based, the same as the matrix.

Function cusparseXcsrsv2_zeroPivot() is a blocking call. It calls
cudaDeviceSynchronize() to make sure all previous kernels are done.

The position can be in the host memory or device memory. The user can set the proper
mode with cusparseSetPointerMode().

Input

| handle | handle to the cuSPARSE library context. |
| info   | info contains structural zero or numerical zero if the user already called csrsv2_analysis() or csrsv2_solve(). |

Output

| position | if no structural or numerical zero, position is -1; otherwise, if A(j,j) is missing or U(j,j) is zero, position = j. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully. |
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized. |
| CUSPARSE_STATUS_INVALID_VALUE | info is not valid. |
| CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed. |
7.21. **cusparse<
t>hybmv()**

```c
cusparseStatus_t
cusparseShybmv(cusparseHandle_t handle, cusparseOperation_t transA,
   const float           *alpha,
   const cusparseMatDescr_t descrA,
   const cusparseHybMat_t hybA, const float           *x,
   const float           *beta, float           *y)
cusparseStatus_t
cusparseDhybmv(cusparseHandle_t handle, cusparseOperation_t transA,
   const double          *alpha,
   const cusparseMatDescr_t descrA,
   const cusparseHybMat_t hybA, const double          *x,
   const double          *beta, double          *y)
cusparseStatus_t
cusparseChybmv(cusparseHandle_t handle, cusparseOperation_t transA,
   const cuComplex       *alpha,
   const cusparseMatDescr_t descrA,
   const cusparseHybMat_t hybA, const cuComplex       *x,
   const cuComplex       *beta, cuComplex       *y)
cusparseStatus_t
cusparseZhybmv(cusparseHandle_t handle, cusparseOperation_t transA,
   const cuDoubleComplex *alpha,
   const cusparseMatDescr_t descrA,
   const cusparseHybMat_t hybA, const cuDoubleComplex *x,
   const cuDoubleComplex *beta, cuDoubleComplex *y)
```

This function performs the matrix-vector operation

\[ y = \alpha \cdot \text{op}(A) \cdot x + \beta \cdot y \]

\( A \) is an \( m \times n \) sparse matrix that is defined in the HYB storage format by an opaque data structure \( \text{hybA} \), \( x \) and \( y \) are vectors, \( \alpha \) and \( \beta \) are scalars, and
\n\( \text{op}(A) = \{ A \; \text{if transA} = \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \}

Notice that currently only \( \text{op}(A) = A \) is supported.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation ( \text{op}(A) ) (currently only ( \text{op}(A) = A ) is supported).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix ( A ).</td>
</tr>
<tr>
<td>alpha</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is ( \text{CUSPARSE_MATRIX_TYPE_GENERAL} ).</td>
</tr>
<tr>
<td>hybA</td>
<td>the matrix ( A ) in HYB storage format.</td>
</tr>
<tr>
<td>x</td>
<td>(&lt;\text{type}&gt;) vector of ( n ) elements.</td>
</tr>
</tbody>
</table>
### beta
- `<type>` scalar used for multiplication. If `beta` is zero, `y` does not have to be a valid input.

### y
- `<type>` vector of `m` elements.

#### Output

| `y` | `<type>` updated vector. |

#### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the internally stored HYB format parameters are invalid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

### 7.22. cusparse<t>hybsv_analysis()

```c
#include <cusparses.h>

cusparsesStatus_t
 cusparseShybsv_analysis(cusparseHandle_t handle,
                          cusparseOperation_t transA,
                          const cusparseMatDescr_t descrA,
                          cusparseHybMat_t hybA,
                          cusparseSolveAnalysisInfo_t info)

cusparsesStatus_t
 cusparseDhybsv_analysis(cusparseHandle_t handle,
                          cusparseOperation_t transA,
                          const cusparseMatDescr_t descrA,
                          cusparseHybMat_t hybA,
                          cusparseSolveAnalysisInfo_t info)

cusparsesStatus_t
 cusparseChybsv_analysis(cusparseHandle_t handle,
                          cusparseOperation_t transA,
                          const cusparseMatDescr_t descrA,
                          cusparseHybMat_t hybA,
                          cusparseSolveAnalysisInfo_t info)

cusparsesStatus_t
 cusparseZhybsv_analysis(cusparseHandle_t handle,
                          cusparseOperation_t transA,
                          const cusparseMatDescr_t descrA,
                          cusparseHybMat_t hybA,
                          cusparseSolveAnalysisInfo_t info)
```

This function performs the analysis phase of the solution of a sparse triangular linear system

\[
\text{op}(A) \ast y = \sigma \ast x
\]
A is an $m \times m$ sparse matrix that is defined in HYB storage format by an opaque data structure hybA, x and y are the right-hand-side and the solution vectors, $a$ is a scalar, and

$$\text{op}(A) = \begin{cases} A & \text{if transA == CUSPARSE_OPERATION_NON_TRANSPOSE} \\ \end{cases}$$

Notice that currently only $\text{op}(A) = A$ is supported.

It is expected that this function will be executed only once for a given matrix and a particular operation type.

This function requires a significant amount of extra storage that is proportional to the matrix size. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

### Input

<table>
<thead>
<tr>
<th>Input</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation $\text{op}(A)$ (currently only $\text{op}(A) = A$ is supported).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix type is CUSPARSE_MATRIX_TYPE_TRIANGULAR and diagonal type USPARSE_DIAG_TYPE_NON_UNIT.</td>
</tr>
<tr>
<td>hybA</td>
<td>the matrix $A$ in HYB storage format.</td>
</tr>
<tr>
<td>info</td>
<td>structure initialized using cusparseCreateSolveAnalysisInfo().</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>Output</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged).</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the internally stored HYB format parameters are invalid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
7.23. `cusparse<t>hybsv_solve()`

Refer to the cuSPARSE Level 2 Function Reference for the full documentation. The function `cusparse<t>hybsv_solve()` performs the solve phase of the solution of a sparse triangular linear system:

\[ \text{op}(A) \cdot y = \alpha \cdot x \]

Where \( A \) is an \( m \times m \) sparse matrix that is defined in HYB storage format by an opaque data structure `hybA`, \( x \) and \( y \) are the right-hand-side and the solution vectors, \( \alpha \) is a scalar, and \( \text{op}(A) = \{ A \quad \text{if transA == CUSPARSE_OPERATION_NON_TRANSPOSE} \} \)

Notice that currently only \( \text{op}(A) = A \) is supported.

This function may be executed multiple times for a given matrix and a particular operation type.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

- **handle**
  - handle to the cuSPARSE library context.
### transA
the operation $\text{op}(A)$ (currently only $\text{op}(A) = A$ is supported).

### alpha
<type> scalar used for multiplication.

### descrA
the descriptor of matrix $A$.
The supported matrix type is `CUSPARSE_MATRIX_TYPE_TRIANGULAR` and the diagonal type is `CUSPARSE_DIAG_TYPE_NON_UNIT`.

### hybA
the matrix $A$ in HYB storage format.

### info
structure with information collected during the analysis phase (that should be passed to the solve phase unchanged).

### x
<type> right-hand-side vector of size $m$.

#### Output

| y | <type> solution vector of size $m$. |

### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>the internally stored hyb format parameters are invalid.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MAPPING_ERROR</code></td>
<td>the texture binding failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_EXECUTION_FAILED</code></td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

### 7.24. cusparseCsr2cscEx2()

```c
cusparseStatus_t
cusparseCsr2cscEx2(cusparseHandle_t handle,
int m,
int n,
int nnz,
const void* csrVal,
const int* csrRowPtr,
const int* csrColInd,
void* cscVal,
int* cscColPtr,
int* cscRowInd,
cudaDataType valType,
cusparseAction_t copyValues,
cusparseIndexBase_t idxBase,
cusparseCsr2CscAlg_t alg,
void* buffer)
```
This function converts a sparse matrix in CSR format (that is defined by the three arrays \texttt{csrVal}, \texttt{csrRowPtr}, and \texttt{csrColInd}) into a sparse matrix in CSC format (that is defined by arrays \texttt{cscVal}, \texttt{cscRowInd}, and \texttt{cscColPtr}). The resulting matrix can also be seen as the transpose of the original sparse matrix. Notice that this routine can also be used to convert a matrix in CSC format into a matrix in CSR format.

For alg \texttt{CUSPARSE_CSR2CSC\_ALG1}: it requires extra storage proportional to the number of nonzero values \texttt{nnz}. It provides in output always the same matrix.

For alg \texttt{CUSPARSE_CSR2CSC\_ALG2}: it requires extra storage proportional to the number of rows \texttt{m}. It does not ensure always the same ordering of CSC column indices and values. Also, it provides better performance than \texttt{CUSPARSE_CSR2CSC\_ALG1} for regular matrices.

It is executed asynchronously with respect to the host, and it may return control to the application on the host before the result is ready.

**Input**

| \texttt{handle} | handle to the cuSPARSE library context |
| \texttt{m} | number of rows of the CSR input matrix; number of columns of the CSC output matrix |
| \texttt{n} | number of columns of the CSR input matrix; number of rows of the CSC output matrix |
| \texttt{nnz} | number of nonzero elements of the CSR and CSC matrices |
| \texttt{csrVal} | value array of size \texttt{nnz} of the CSR matrix; of same type as \texttt{valType} |
| \texttt{csrRowPtr} | integer array of size \texttt{m} + 1 that contains the CSR row offsets |
| \texttt{csrColInd} | integer array of size \texttt{nnz} that contains the CSR column indices |
| \texttt{valType} | value type for both CSR and CSC matrices |
| \texttt{copyValues} | \texttt{CUSPARSE\_ACTION\_SYMBOLIC} or \texttt{CUSPARSE\_ACTION\_NUMERIC} |
| \texttt{idxBase} | Index base \texttt{CUSPARSE\_INDEX\_BASE\_ZERO} or \texttt{CUSPARSE\_INDEX\_BASE\_ONE}. |
| \texttt{alg} | algorithm implementation. see \texttt{cusparseCsr2CscAlg\_t} for possible values. |
| \texttt{buffer} | pointer to workspace buffer |

**Output**

| \texttt{cscVal} | value array of size \texttt{nnz} of the CSC matrix; of same type as \texttt{valType} |
| \texttt{cscColPtr} | integer array of size \texttt{n} + 1 that contains the CSC column offsets |
| \texttt{cscRowInd} | integer array of size \texttt{nnz} that contains the CSC row indices |
7.25. cusparseCsr2cscEx2_bufferSize()

cusparseStatus_t

cusparseCsr2cscEx2_bufferSize(cusparseHandle_t handle,
int m,
int n,
int nnz,
const void* csrVal,
const int* csrRowPtr,
const int* csrColInd,
void* cscVal,
int* cscColPtr,
int* cscRowInd,
cudaDataType valType,
cusparseAction_t copyValues,
cusparseIndexBase_t idxBase,
cusparseCsr2CscAlg_t alg,
size_t* bufferSize)

This function returns the size of the workspace needed by cusparseCsr2cscEx2(). User needs to allocate a buffer of this size and give that buffer to 
cusparseCsr2cscEx2() as an argument. All the arguments are similar to 
cusparseCsr2cscEx2() except the following output argument.

Output

| bufferSize | number of bytes of workspace needed by cusparseCsr2cscEx2(). |
This chapter describes sparse linear algebra functions that perform operations between sparse and (usually tall) dense matrices.

In particular, the solution of sparse triangular linear systems with multiple right-hand sides is implemented in two phases. First, during the analysis phase, the sparse triangular matrix is analyzed to determine the dependencies between its elements by calling the appropriate `csrsm_analysis()` function. The analysis is specific to the sparsity pattern of the given matrix and to the selected `cusparseOperation_t` type. The information from the analysis phase is stored in the parameter of type `cusparseSolveAnalysisInfo_t` that has been initialized previously with a call to `cusparseCreateSolveAnalysisInfo()`.

Second, during the solve phase, the given sparse triangular linear system is solved using the information stored in the `cusparseSolveAnalysisInfo_t` parameter by calling the appropriate `csrsm_solve()` function. The solve phase may be performed multiple times with different multiple right-hand sides, while the analysis phase needs to be performed only once. This is especially useful when a sparse triangular linear system must be solved for different sets of multiple right-hand sides one at a time, while its coefficient matrix remains the same.

Finally, once all the solves have completed, the opaque data structure pointed to by the `cusparseSolveAnalysisInfo_t` parameter can be released by calling `cusparseDestroySolveAnalysisInfo()`. For more information please refer to [3].
8.1. cusparse<t>csrmm()

cusparseStatus_t
cusparseScsrmm(cusparseHandle_t handle,
cusparseOperation_t transA,
int m,
int n,
int k,
int nnz,
const float *alpha,
const cusparseMatDescr_t descrA,
const float *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
const float *B,
int ldb,
const float *beta,
float *C,
int ldc)
cusparseStatus_t
cusparseDcsrmm(cusparseHandle_t handle,
cusparseOperation_t transA,
int m,
int n,
int k,
int nnz,
const double *alpha,
const cusparseMatDescr_t descrA,
const double *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
const double *B,
int ldb,
const double *beta,
double *C,
int ldc)
cusparseStatus_t
cusparseCcsrmm(cusparseHandle_t handle,
cusparseOperation_t transA,
int m,
int n,
int k,
int nnz,
const cuComplex *alpha,
const cusparseMatDescr_t descrA,
cusparseOperation_t descrA,
const cuComplex *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparseOperation_t *B,
int ldb,
cusparseStatus_t cusparseCcsrmm(cusparseHandle_t handle,
cusparseOperation_t descrA,
const cuComplex *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparseStatus_t cusparseZcsrmm(cusparseHandle_t handle,
cusparseOperation_t descrA,
const cuDoubleComplex *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparseStatus_t cusparseZcsrmm(cusparseHandle_t handle,
cusparseOperation_t descrA,
const cuDoubleComplex *csrValA,
This function performs one of the following matrix-matrix operations:

\[ C = \alpha \ast \text{op}(A) \ast B + \beta \ast C \]

\( A \) is an \( m \times k \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA}, \text{csrRowPtrA}, \) and \( \text{csrColIndA} \); \( B \) and \( C \) are dense matrices; \( \alpha \) and \( \beta \) are scalars; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if trans == CUSPARSE_OPERATION_NON_TRANSPOSE} \\
A^T & \text{if trans == CUSPARSE_OPERATION_TRANSPOSE} \\
A_H & \text{if trans == CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} 
\end{cases}
\]

When using the (conjugate) transpose of a general matrix or a Hermitian/symmetric matrix, this routine may produce slightly different results with the same input parameters during different runs of this function. For these matrix types it uses atomic operations to compute the final result; consequently, many threads may be adding floating point numbers to the same memory location without any specific ordering, which may produce slightly different results for each run.

If exactly the same output is required for any input when multiplying by the transpose of a general matrix, the following procedure can be used:

1. Convert the matrix from CSR to CSC format using one of the \( \text{csr2csc()} \) functions. Notice that by interchanging the rows and columns of the result you are implicitly transposing the matrix.

2. Call the \( \text{csrmm()} \) function with the \( \text{cusparseOperation_t} \) parameter set to \( \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \) and with the interchanged rows and columns of the matrix stored in CSC format. This (implicitly) multiplies the vector by the transpose of the matrix in the original CSR format.

This function requires no extra storage for the general matrices when operation \( \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \) is selected. It requires some extra storage for Hermitian/symmetric matrices and for the general matrices when an operation different from \( \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \) is selected. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation ( \text{op}(A) ).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of sparse matrix ( A ).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of dense matrices ( B ) and ( C ).</td>
</tr>
<tr>
<td>k</td>
<td>number of columns of sparse matrix ( A ).</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of sparse matrix ( A ).</td>
</tr>
<tr>
<td>alpha</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix types are \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, \texttt{CUSPARSE_MATRIX_TYPE_SYMMETRIC}, and \texttt{CUSPARSE_MATRIX_TYPE_HERMITIAN}. Also, the supported index bases are</td>
</tr>
</tbody>
</table>
### cuSPARSE Level 3 Function Reference

CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrValA</td>
<td><code>&lt;type&gt;</code> array of $nnz(=csrRowPtrA(m) - csrRowPtrA(0))$ nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>Integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>Integer array of $nnz(=csrRowPtrA(m) - csrRowPtrA(0))$ column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>B</td>
<td>Array of dimensions $(ldb, n)$.</td>
</tr>
<tr>
<td>ldb</td>
<td>Leading dimension of B. It must be at least $\max(1, k)$ if $\text{op}(A)=A$ and at least $\max(1, m)$ otherwise.</td>
</tr>
<tr>
<td>beta</td>
<td><code>&lt;type&gt;</code> scalar used for multiplication. If $\beta$ is zero, C does not have to be a valid input.</td>
</tr>
<tr>
<td>C</td>
<td>Array of dimensions $(ldc, n)$.</td>
</tr>
<tr>
<td>ldc</td>
<td>Leading dimension of C. It must be at least $\max(1, m)$ if $\text{op}(A)=A$ and at least $\max(1, k)$ otherwise.</td>
</tr>
</tbody>
</table>

#### Output

| C           | `<type>` updated array of dimensions $(ldc, n)$. |

#### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>The operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>The library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>The resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>Invalid parameters were passed ($m, n, k, nnz&lt;0$ or $\text{ldb}$ and $\text{ldc}$ are incorrect).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>The device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>The function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>An internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>The matrix type is not supported.</td>
</tr>
</tbody>
</table>
8.2. \texttt{cusparse<t>csrmm2()}

cusparseStatus_t cusparseScsrmm2(cusparseHandle_t handle, cusparseOperation_t transA, cusparseOperation_t transB, int m, int n, int k, int nnz, const float *alpha, const cusparseMatDescr_t descrA, const float *csrValA, const int *csrRowPtrA, const int *csrColIndA, const float *B, int ldb, const float *beta, float *C, int ldc)
cusparseStatus_t cusparseDcsrmm2(cusparseHandle_t handle, cusparseOperation_t transA, cusparseOperation_t transB, int m, int n, int k, int nnz, const double *alpha, const cusparseMatDescr_t descrA, const double *csrValA, const int *csrRowPtrA, const int *csrColIndA, const double *B, int ldb, const double *beta, double *C, int ldc)
cusparseStatus_t cusparseCcsrmm2(cusparseHandle_t handle, cusparseOperation_t transA, cusparseOperation_t transB, int m, int n, int k, int nnz, const cuComplex *alpha, const cusparseMatDescr_t descrA, const cuComplex *csrValA, const int *csrRowPtrA, const int *csrColIndA, const cuComplex *B, int ldb, const cuComplex *beta, cuComplex *C, int ldc)
cusparseStatus_t cusparseZcsrmm2(cusparseHandle_t handle, cusparseOperation_t transA, cusparseOperation_t transB, int m, int n, int k, int nnz, const cuDoubleComplex *alpha, const cusparseMatDescr_t descrA, const cuDoubleComplex *csrValA, const int *csrRowPtrA, const int *csrColIndA, const cuDoubleComplex *B, int ldb, const cuDoubleComplex *beta, cuDoubleComplex *C, int ldc)
This function performs one of the following matrix-matrix operations:

\[ C = \alpha \ast \text{op}(A) \ast \text{op}(B) + \beta \ast C \]

A is an \( m \times k \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA}, \text{csrRowPtrA}, \text{csrColIndA} \); B and C are dense matrices; \( \alpha \) and \( \beta \) are scalars; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{transA} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{transA} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if } \text{transA} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE}
\end{cases}
\]

and

\[
\text{op}(B) = \begin{cases} 
B & \text{if } \text{transB} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
B^T & \text{if } \text{transB} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
B^H & \text{not supported}
\end{cases}
\]

If \( \text{op}(B) = B \), \( \text{cusparse}<t>\text{csrmm2()} \) is the same as \( \text{cusparse}<t>\text{csrmm()} \); otherwise, only \( \text{op}(A) = A \) is supported and the matrix type must be \text{CUSPARSE\_MATRIX\_TYPE\_GENERAL}.

The motivation of \( \text{transpose}(B) \) is to improve the memory access of matrix B. The computational pattern of \( A\ast\text{transpose}(B) \) with matrix B in column-major order is equivalent to \( A\ast B \) with matrix B in row-major order.

In practice, no operation in iterative solver or eigenvalue solver uses \( A\ast\text{transpose}(B) \). However, we can perform \( A\ast\text{transpose}(\text{transpose}(B)) \) which is the same as \( A\ast B \).

For example, suppose A is \( m \times k \), B is \( k \times n \) and C is \( m \times n \), the following code shows usage of \text{cusparseDcsrmm}().

```c
// A is m\times k, B is k\times n and C is m\times n
const int ldb_B = k; // leading dimension of B
const int ldc = m; // leading dimension of C
// perform C:=alpha*A*B + beta*C
cusparseSetMatType(descrA, CUSPARSE\_MATRIX\_TYPE\_GENERAL );
cusparseDcsrmm(cusparse\_handle, 
    CUSPARSE\_OPERATION\_NON\_TRANSPOSE, 
m, n, k, nnz, alpha,
    descrA, csrValA, csrRowPtrA, csrColIndA, 
    B, ldb_B, 
    beta, C, ldc);
```
Instead of using $A*B$, our proposal is to transpose $B$ to $B^t$ first by calling 
\texttt{cublas<t>geam()}\texttt{, then to perform $A*\text{transpose}(B^t)$}.

\begin{verbatim}
// step 1: $B^t := \text{transpose}(B)$
double *Bt;
const int ldb_Bt = n; // leading dimension of $B$
cudaMalloc((void**)&Bt, sizeof(double)*ldb_Bt*k);
double one  = 1.0;
double zero = 0.0;
cublasSetPointerMode(cublas_handle, CUBLAS_POINTER_MODE_HOST);
cublasDgeam(cublas_handle, CUBLAS_OP_T, CUBLAS_OP_T,
n, k, &one, B, int ldb_B, &zero, Bt, int ldb_Bt, Bt, ldb_Bt);

// step 2: perform $C := \alpha A^t B^t + \beta C$
cusparseDcsrmm2(cusparse_handle,
    CUSPARSE_OPERATION_NON_TRANSPOSE,
    CUSPARSE_OPERATION_TRANSPOSE
    m, n, k, nnz, alpha,
descrA, csrValA, csrRowPtrA, csrColIndA,
    Bt, ldb_Bt,
    beta, C, ldc);
\end{verbatim}

Remark 1: \texttt{cublas<t>geam()} and \texttt{cusparse<t>csrmm2()} are memory bound. The complexity of \texttt{cublas<t>geam()} is $2*n*k$, and the minimum complexity of \texttt{cusparse<t>csrmm2()} is about $(nnz + nnz*n + 2*m*n)$. If $nnz$ per column ($=nnz/k$) is large, it is worth paying the extra cost on transposition because $A^t B^t$ may be 2× faster than $A*B$ if the sparsity pattern of $A$ is not good.

Remark 2: $A^t B^t$ is only supported on compute capability 2.0 and above.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation $\text{op}(A)$</td>
</tr>
<tr>
<td>transB</td>
<td>the operation $\text{op}(B)$</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of sparse matrix $A$.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of dense matrix $\text{op}(B)$ and $C$.</td>
</tr>
<tr>
<td>k</td>
<td>number of columns of sparse matrix $A$.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of sparse matrix $A$.</td>
</tr>
<tr>
<td>alpha</td>
<td>$\text{&lt;type&gt;}$ scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix types is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}. Also, the supported index bases are \texttt{CUSPARSE_INDEX_BASE_ZERO} and \texttt{CUSPARSE_INDEX_BASE_ONE}.</td>
</tr>
<tr>
<td>csrValA</td>
<td>$\text{&lt;type&gt;}$ array of $nnz$ ($= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)$) nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of $nnz$ ($= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)$) column indices of the nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>B</td>
<td>array of dimensions (ldb, n) if op(B)=B and (ldb, k) otherwise.</td>
</tr>
<tr>
<td>-----</td>
<td>-----------------------------------------------------------------</td>
</tr>
<tr>
<td>ldb</td>
<td>leading dimension of B. If op(B)=B, it must be at least max(1, k) if op(A)=A and at least max(1, m) otherwise. If op(B) != B, it must be at least max(1, n).</td>
</tr>
<tr>
<td>beta</td>
<td>&lt;type&gt; scalar used for multiplication. If beta is zero, C does not have to be a valid input.</td>
</tr>
<tr>
<td>C</td>
<td>array of dimensions (ldc, n).</td>
</tr>
<tr>
<td>ldc</td>
<td>leading dimension of C. It must be at least max(1, m) if op(A)=A and at least max(1, k) otherwise.</td>
</tr>
</tbody>
</table>

**Output**

| C   | <type> updated array of dimensions (ldc, n). |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n, k, nnz&lt;0 or ldb and ldc are incorrect).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>if op(B)=B, the device does not support double precision or if op(B) = transpose(B) the device is below compute capability 2.0.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>only CUSPARSE_MATRIX_TYPE_GENERAL is supported otherwise.</td>
</tr>
</tbody>
</table>
8.3. `cusparse<t>csrsm_analysis()`

This function performs the analysis phase of the solution of a sparse triangular linear system.
with multiple right-hand sides, where $A$ is an $m \times m$ sparse matrix that is defined in CSR storage format by the three arrays $\text{csrValA}$, $\text{csrRowPtrA}$, and $\text{csrColIndA}$; $B$ and $X$ are the right-hand-side and the solution dense matrices; $\alpha$ is a scalar; and

$$\text{op}(A) = \begin{cases} A & \text{if trans} = \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\ A^T & \text{if trans} = \text{CUSPARSE_OPERATION_TRANSPOSE} \\ A^{H} & \text{if trans} = \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} \end{cases}$$

It is expected that this function will be executed only once for a given matrix and a particular operation type.

This function requires a significant amount of extra storage that is proportional to the matrix size. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation $\text{op}(A)$.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix $A$.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix types are CUSPARSE_MATRIX_TYPE_TRIANGULAR and CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types are CUSPARSE_DIAG_TYPE_UNIT and CUSPARSE_DIAG_TYPE_NON_UNIT.</td>
</tr>
<tr>
<td>csrValA</td>
<td>$&lt;$type$&gt;$ array of $\text{nnz} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)$ nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of $\text{nnz} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)$ column indices of the nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>info</td>
<td>structure initialized using $\text{cusparseCreateSolveAnalysisInfo()}$.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged).</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>cuSPARSE Status Code</td>
<td>Description</td>
</tr>
<tr>
<td>----------------------</td>
<td>-------------</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, nnz&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
8.4. cusparse<t>csrsm_solve()

cusparseStatus_t
    cusparseScsrsm_solve(
        cusparseHandle_t handle,
        cusparseOperation_t transA,
        int m,
        int n,
        const float *alpha,
        const cusparseMatDescr_t descrA,
        const float *csrSortedValA,
        const int *csrSortedRowPtrA,
        const int *csrSortedColIndA,
        cusparseSolveAnalysisInfo_t info,
        const float *B,
        int ldb,
        float *X,
        int ldx)

cusparseStatus_t
    cusparseDcsrsm_solve(
        cusparseHandle_t handle,
        cusparseOperation_t transA,
        int m,
        int n,
        const double *alpha,
        const cusparseMatDescr_t descrA,
        const double *csrSortedValA,
        const int *csrSortedRowPtrA,
        const int *csrSortedColIndA,
        cusparseSolveAnalysisInfo_t info,
        const double *B,
        int ldb,
        double *X,
        int ldx)

cusparseStatus_t
    cusparseCcsrsm_solve(
        cusparseHandle_t handle,
        cusparseOperation_t transA,
        int m,
        int n,
        const cuComplex *alpha,
        const cusparseMatDescr_t descrA,
        const cuComplex *csrSortedValA,
        const int *csrSortedRowPtrA,
        const int *csrSortedColIndA,
        cusparseSolveAnalysisInfo_t info,
        const cuComplex *B,
        int ldb,
        cuComplex *X,
        int ldx)

cusparseStatus_t
    cusparseZcsrsm_solve(
        cusparseHandle_t handle,
        cusparseOperation_t transA,
        int m,
        int n,
        const cuDoubleComplex *alpha,
        const cusparseMatDescr_t descrA,
        const cuDoubleComplex *csrSortedValA,
        const int *csrSortedRowPtrA,
        const int *csrSortedColIndA,
This function performs the solve phase of the solution of a sparse triangular linear system

\[ \text{op}(A) \times X = \alpha \times B \]

with multiple right-hand sides, where \( A \) is an \( m \times n \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA} \), \( \text{csrRowPtrA} \), and \( \text{csrColIndA} \); \( B \) and \( X \) are the right-hand-side and the solution dense matrices; \( \alpha \) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if trans} = \text{CUSPARSEOPERATIONNONTRANSPOSE} \\
A^T & \text{if trans} = \text{CUSPARSEOPERATIONTRANSPOSE} \\
A^H & \text{if trans} = \text{CUSPARSEOPERATIONCONJUGATETRANSPOSE}
\end{cases}
\]

This function may be executed multiple times for a given matrix and a particular operation type.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation ( \text{op}(A) ).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows and columns of matrix ( A ).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix ( X ) and ( Y ).</td>
</tr>
<tr>
<td>alpha</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix types are ( \text{CUSPARSEMATRIXTYPE_TRIANGULAR} ) and ( \text{CUSPARSEMATRIXTYPEGENERAL} ), while the supported diagonal types are ( \text{CUSPARSEDIAGTYPEUNIT} ) and ( \text{CUSPARSEDIAGTYPENONUNIT} ).</td>
</tr>
<tr>
<td>csrValA</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnz} ) nonzeros of matrix ( A ).</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of ( m + 1 ) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of ( \text{nnz} ) column indices of the nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>info</td>
<td>structure with information collected during the analysis phase (that should be passed to the solve phase unchanged).</td>
</tr>
<tr>
<td>B</td>
<td>(&lt;\text{type}&gt;) right-hand-side array of dimensions ((\text{ldb}, n)).</td>
</tr>
<tr>
<td>ldb</td>
<td>leading dimension of ( B ) (that is ( \geq \max(1, m) )).</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>X</td>
<td>(&lt;\text{type}&gt;) solution array of dimensions ((\text{ldx}, n)).</td>
</tr>
<tr>
<td>ldx</td>
<td>leading dimension of ( X ) (that is ( \geq \max(1, m) )).</td>
</tr>
</tbody>
</table>
### Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MAPPING_ERROR</td>
<td>the texture binding failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
8.5. `cusparse<t>csrsm2_bufferSizeExt()`

cusparseStatus_t
cusparseScsrsm2_bufferSizeExt(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const float *alpha,
    const cusparseMatDescr_t descrA,
    const float *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const float *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    size_t *pBufferSize)

cusparseStatus_t
cusparseDcsrsm2_bufferSizeExt(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const double *alpha,
    const cusparseMatDescr_t descrA,
    const double *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const double *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    size_t *pBufferSize)

cusparseStatus_t
cusparseCcsrsm2_bufferSizeExt(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const cuComplex *alpha,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const cuComplex *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    size_t *pBufferSize)

cusparseStatus_t
cusparseZcsrsm2_bufferSizeExt(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const cuDoubleComplex *alpha,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const cuDoubleComplex *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    size_t *pBufferSize)
This function returns the size of the buffer used in \texttt{csrsm2}, a sparse triangular linear system \( \text{op}(A) \times \text{op}(X) = \alpha \text{op}(B) \).

\( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \texttt{csrValA}, \texttt{csrRowPtrA}, and \texttt{csrColIndA}; \( B \) and \( X \) are the right-hand-side matrix and the solution matrix; \( \alpha \) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
    A^T & \text{if trans} = \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
    A^T & \text{if trans} = \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
    A^H & \text{if trans} = \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE}
\end{cases}
\]

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>\texttt{handle}</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>\texttt{algo}</td>
<td>algo = 0 is non-block version; algo = 1 is block version.</td>
</tr>
<tr>
<td>\texttt{transA}</td>
<td>the operation ( \text{op}(A) ).</td>
</tr>
<tr>
<td>\texttt{transB}</td>
<td>the operation ( \text{op}(B) ).</td>
</tr>
<tr>
<td>\texttt{m}</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{nrhs}</td>
<td>number of columns of right hand side matrix ( \text{op}(B) ).</td>
</tr>
<tr>
<td>\texttt{nnz}</td>
<td>number of nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{alpha}</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication.</td>
</tr>
<tr>
<td>\texttt{descrA}</td>
<td>the descriptor of matrix ( A ). The supported matrix type is ( \text{CUSPARSE_MATRIX_TYPE_GENERAL} ), while the supported diagonal types are ( \text{CUSPARSE_DIAG_TYPE_UNIT} ) and ( \text{CUSPARSE_DIAG_TYPE_NON_UNIT} ).</td>
</tr>
<tr>
<td>\texttt{csrValA}</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnz} (= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0) ) ) nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{csrRowPtrA}</td>
<td>integer array of ( m + 1 ) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>\texttt{csrColIndA}</td>
<td>integer array of ( \text{nnz} (= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0) ) ) column indices of the nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{B}</td>
<td>(&lt;\text{type}&gt;) right-hand-side matrix. ( \text{op}(B) ) is of size ( m \times \text{nrhs} ).</td>
</tr>
<tr>
<td>\texttt{ldb}</td>
<td>leading dimension of ( B ) and ( X ).</td>
</tr>
<tr>
<td>\texttt{info}</td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
<tr>
<td>\texttt{policy}</td>
<td>The supported policies are ( \text{CUSPARSE_SOLVE_POLICY_NO_LEVEL} ) and ( \text{CUSPARSE_SOLVE_POLICY_USE_LEVEL} ).</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>\texttt{info}</td>
<td>record of internal states based on different algorithms.</td>
</tr>
</tbody>
</table>
### pBufferSize

number of bytes of the buffer used in the `csrsm2_analysis` and `csrsm2_solve`.

### Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, nnz &lt;= 0), base index is not 0 or 1.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
8.6. `cusparse<t>csrsm2_analysis()`

cusparseStatus_t
cusparseScsrsm2_analysis(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const float *alpha,
    const cusparseMatDescr_t descrA,
    const float *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const float *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)

cusparseStatus_t
cusparseDcsrsm2_analysis(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const double *alpha,
    const cusparseMatDescr_t descrA,
    const double *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const double *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)

cusparseStatus_t
cusparseCcsrsm2_analysis(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const cuComplex *alpha,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const cuComplex *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)

cusparseStatus_t
cusparseZcsrsm2_analysis(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const cuDoubleComplex *alpha,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    const cuDoubleComplex *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)
This function performs the analysis phase of \texttt{csrsm2}, a sparse triangular linear system \[ \text{op}(A) \ast \text{op}(X) = \alpha \text{op}(B). \]

\( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \texttt{csrValA}, \texttt{csrRowPtrA}, and \texttt{csrColIndA}; \( B \) and \( X \) are the right-hand-side matrix and the solution matrix; \( \alpha \) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
    A & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
    A^T & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
    A^H & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE}
\end{cases}
\]

It is expected that this function will be executed only once for a given matrix and a particular operation type.

This function requires a buffer size returned by \texttt{csrsm2\_bufferSize()}. The address of \texttt{pBuffer} must be multiple of 128 bytes. If it is not, \texttt{CUSPARSE\_STATUS\_INVALID\_VALUE} is returned.

Function \texttt{csrsm2\_analysis()} reports a structural zero and computes level information that is stored in opaque structure \texttt{info}. The level information can extract more parallelism for a triangular solver. However \texttt{csrsm2\_solve()} can be done without level information. To disable level information, the user needs to specify the policy of the triangular solver as \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}.

Function \texttt{csrsm2\_analysis()} always reports the first structural zero, even if the policy is \texttt{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}. No structural zero is reported if \texttt{CUSPARSE\_DIAG\_TYPE\_UNIT} is specified, even if \( A(j,j) \) is missing for some \( j \). The user needs to call \texttt{cusparseXcsrsm2\_zeroPivot()} to know where the structural zero is.

It is the user’s choice whether to call \texttt{csrsm2\_solve()} if \texttt{csrsm2\_analysis()} reports a structural zero. In this case, the user can still call \texttt{csrsm2\_solve()} which will return a numerical zero in the same position as the structural zero. However the result \( X \) is meaningless.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>algo</td>
<td>algo = 0 is non-block version; algo = 1 is block version.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation \text{op}(A).</td>
</tr>
<tr>
<td>transB</td>
<td>the operation \text{op}(B).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>nrhs</td>
<td>number of columns of matrix \text{op}(B).</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>alpha</td>
<td>\langle\text{type}\rangle scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, while the supported diagonal types are \texttt{CUSPARSE_DIAG_TYPE_UNIT} and \texttt{CUSPARSE_DIAG_TYPE_NON_UNIT}.</td>
</tr>
</tbody>
</table>
### csrValA

- `<type>` array of `nnz (= csrRowPtrA(m) - csrRowPtrA(0))` nonzero elements of matrix `A`.

### csrRowPtrA

- Integer array of `m + 1` elements that contains the start of every row and the end of the last row plus one.

### csrColIndA

- Integer array of `nnz (= csrRowPtrA(m) - csrRowPtrA(0))` column indices of the nonzero elements of matrix `A`.

### B

- `<type>` right-hand-side matrix. `op(B)` is of size `m-by-nrhs`.

### ldb

- Leading dimension of `B` and `X`.

### info

- Structure initialized using `cusparseCreateCsrsv2Info()`.

### policy

- The supported policies are `CUSPARSE_SOLVE_POLICY_NO_LEVEL` and `CUSPARSE_SOLVE_POLICY_USE_LEVEL`.

### pBuffer

- Buffer allocated by the user, the size is returned by `csrsm2_bufferSize()`.

### Output

- `info` structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged).

### Status Returned

- **CUSPARSE_STATUS_SUCCESS**
  - The operation completed successfully.

- **CUSPARSE_STATUS_NOT_INITIALIZED**
  - The library was not initialized.

- **CUSPARSE_STATUS_INVALID_VALUE**
  - Invalid parameters were passed (`m, nnz <= 0`).

- **CUSPARSE_STATUS_EXECUTION_FAILED**
  - The function failed to launch on the GPU.

- **CUSPARSE_STATUS_INTERNAL_ERROR**
  - An internal operation failed.
8.7. cuSparse<t>csrsm2_solve()

cusparseStatus_t
 cusparseScsrsm2_solve(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const float *alpha,
    const cusparseMatDescr_t descrA,
    const float *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    float *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)

cusparseStatus_t
 cusparseDcsrsm2_solve(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const double *alpha,
    const cusparseMatDescr_t descrA,
    const double *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    double *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)

cusparseStatus_t
 cusparseCcsrsm2_solve(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const cuComplex *alpha,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    cuComplex *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)

cusparseStatus_t
 cusparseZcsrsm2_solve(
    cusparseHandle_t handle,
    int algo,
    cusparseOperation_t transA,
    cusparseOperation_t transB,
    int m,
    int nrhs,
    int nnz,
    const cuDoubleComplex *alpha,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrSortedValA,
    const int *csrSortedRowPtrA,
    const int *csrSortedColIndA,
    cuDoubleComplex *B,
    int ldb,
    csrsm2Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer)
This function performs the solve phase of \( \text{csrsm2} \), a sparse triangular linear system 
\[ \text{op}(A) \ast \text{op}(X) = \alpha \text{op}(B). \]

\( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA}, \text{csrRowPtrA}, \) and \( \text{csrColIndA} \); \( B \) and \( X \) are the right-hand-side matrix and the solution matrix; \( \alpha \) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{transA} = \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
A^T & \text{if } \text{transA} = \text{CUSPARSE_OPERATION_TRANSPOSE} \\
A^H & \text{if } \text{transA} = \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE}
\end{cases}
\]

\( \text{transB} \) acts on both matrix \( B \) and matrix \( X \), only \( \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \) and \( \text{CUSPARSE_OPERATION_TRANSPOSE} \). The operation is in-place, matrix \( B \) is overwritten by matrix \( X \).

\( \text{ldb} \) must be not less than \( m \) if \( \text{transB} = \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \). Otherwise, \( \text{ldb} \) must be not less than \( \text{nrhs} \).

This function requires the buffer size returned by \( \text{csrsm2\_bufferSize()} \). The address of \( \text{pBuffer} \) must be multiple of 128 bytes. If it is not, \( \text{CUSPARSE\_STATUS\_INVALID\_VALUE} \) is returned.

Although \( \text{csrsm2\_solve()} \) can be done without level information, the user still needs to be aware of consistency. If \( \text{csrsm2\_analysis()} \) is called with policy \( \text{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} \), \( \text{csrsm2\_solve()} \) can be run with or without levels. On the contrary, if \( \text{csrsm2\_analysis()} \) is called with \( \text{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} \), \( \text{csrsm2\_solve()} \) can only accept \( \text{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} \); otherwise, \( \text{CUSPARSE\_STATUS\_INVALID\_VALUE} \) is returned.

The level information may not improve the performance but spend extra time doing analysis. For example, a tridiagonal matrix has no parallelism. In this case, \( \text{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} \) performs better than \( \text{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} \). If the user has an iterative solver, the best approach is to do \( \text{csrsm2\_analysis()} \) with \( \text{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} \) once. Then do \( \text{csrsm2\_solve()} \) with \( \text{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL} \) in the first run and with \( \text{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL} \) in the second run, picking faster one to perform the remaining iterations.

Function \( \text{csrsm2\_solve()} \) reports the first numerical zero, including a structural zero. If \( \text{status} \) is 0, no numerical zero was found. Furthermore, no numerical zero is reported if \( \text{CUSPARSE\_DIAG\_TYPE\_UNIT} \) is specified, even if \( A(j,j) \) is zero for some \( j \). The user needs to call \( \text{cusparseXcsrsm2\_zeroPivot()} \) to know where the numerical zero is.

\( \text{csrsm2} \) provides two algorithms specified by the parameter \( \text{algo} \). \( \text{algo}=0 \) is non-block version and \( \text{algo}=1 \) is block version. non-block version is memory-bound, limited by bandwidth. block version partitions the matrix into small tiles and applies desne operations. Although it has more flops than non-block version, it may be faster if non-block version already reaches maximum bandwidth.

Appendix H shows an example of \( \text{csrsm2} \).

**Input**

| handle | handle to the cuSPARSE library context. |

---

www.nvidia.com

cuSPARSE Library
<table>
<thead>
<tr>
<th>algo</th>
<th>algo = 0 is non-block version; algo = 1 is block version.</th>
</tr>
</thead>
<tbody>
<tr>
<td>transA</td>
<td>the operation ( \text{op}(A) ).</td>
</tr>
<tr>
<td>transB</td>
<td>the operation ( \text{op}(B) ).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows and columns of matrix ( A ).</td>
</tr>
<tr>
<td>nrhs</td>
<td>number of columns of matrix ( \text{op}(B) ).</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzeros of matrix ( A ).</td>
</tr>
<tr>
<td>alpha</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is \text{CUSPARSE_MATRIX_TYPE_GENERAL}, while the supported diagonal types are \text{CUSPARSE_DIAG_TYPE_UNIT} and \text{CUSPARSE_DIAG_TYPE_NON_UNIT}.</td>
</tr>
<tr>
<td>csrValA</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnz} ) nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of ( m + 1 ) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of ( \text{nnz} ) column indices of the nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>B</td>
<td>(&lt;\text{type}&gt;) right-hand-side matrix. ( \text{op}(B) ) is of size ( m )-by-( \text{nrhs} ).</td>
</tr>
<tr>
<td>ldb</td>
<td>leading dimension of ( B ) and ( X ).</td>
</tr>
<tr>
<td>info</td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
<tr>
<td>policy</td>
<td>The supported policies are \text{CUSPARSE_SOLVE_POLICY_NO_LEVEL} and \text{CUSPARSE_SOLVE_POLICY_USE_LEVEL}.</td>
</tr>
<tr>
<td>pBuffer</td>
<td>buffer allocated by the user, the size is returned by \text{csrsm2_bufferSize}.</td>
</tr>
</tbody>
</table>

**Output**

| \( X \)  | \(<\text{type}>\) solution matrix, \( \text{op}(X) \) is of size \( m \)-by-\( \text{nrhs} \). |

**Status Returned**

| \( \text{CUSPARSE\_STATUS\_SUCCESS} \) | the operation completed successfully. |
| \( \text{CUSPARSE\_STATUS\_NOT\_INITIALIZED} \) | the library was not initialized. |
| \( \text{CUSPARSE\_STATUS\_INVALID\_VALUE} \) | invalid parameters were passed \( (m, \text{nnz} \leq 0) \). |
| \( \text{CUSPARSE\_STATUS\_EXECUTION\_FAILED} \) | the function failed to launch on the GPU. |
| \( \text{CUSPARSE\_STATUS\_INTERNAL\_ERROR} \) | an internal operation failed. |
8.8. cusparseXcsrsm2_zeroPivot()

cusparseStatus_t
cusparseXcsrsm2_zeroPivot(cusparseHandle_t handle,
    csrsm2Info_t info,
    int *position);

If the returned error code is CUSPARSE_STATUS_ZERO_PIVOT, position=j means
A(j,j) has either a structural zero or a numerical zero. Otherwise position=-1.

The position can be 0-based or 1-based, the same as the matrix.

Function cusparseXcsrsm2_zeroPivot() is a blocking call. It calls
cudaDeviceSynchronize() to make sure all previous kernels are done.

The position can be in the host memory or device memory. The user can set the proper
mode with cusparseSetPointerMode().

Input

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>info contains structural zero or numerical zero if the user already called csrsm2_analysis() or csrsm2_solve().</td>
</tr>
</tbody>
</table>

Output

| position | if no structural or numerical zero, position is -1; otherwise, if A(j,j) is missing or U(j,j) is zero, position=j. |

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>info is not valid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
8.9. `cusparse<t>bsrmm()`

```c
 cusparseStatus_t
cusparseSbsrmm(cusparseHandle_t handle,
              cusparseDirection_t dirA,
              cusparseOperation_t transA,
              cusparseOperation_t transB,
              int mb,
              int n,
              int kb,
              int nnzb,
              const float *alpha,
              const cusparseMatDescr_t descrA,
              const float *bsrValA,
              const int *bsrRowPtrA,
              const int *bsrColIndA,
              const int blockDim,
              const float *B,
              const int ldb,
              const float *beta,
              float *C,
              int ldc)
```

```c
 cusparseStatus_t
cusparseDbsrmm(cusparseHandle_t handle,
              cusparseDirection_t dirA,
              cusparseOperation_t transA,
              cusparseOperation_t transB,
              int mb,
              int n,
              int kb,
              int nnzb,
              const double *alpha,
              const cusparseMatDescr_t descrA,
              const double *bsrValA,
              const int *bsrRowPtrA,
              const int *bsrColIndA,
              const int blockDim,
              const double *B,
              const int ldb,
              const double *beta,
              double *C,
              int ldc)
```

```c
 cusparseStatus_t
cusparseCbsrmm(cusparseHandle_t handle,
              cusparseDirection_t dirA,
              cusparseOperation_t transA,
              cusparseOperation_t transB,
              int mb,
              int n,
              int kb,
              int nnzb,
              const cuComplex *alpha,
              const cusparseMatDescr_t descrA,
              const cuComplex *bsrValA,
              const int *bsrRowPtrA,
              const int *bsrColIndA,
              const int blockDim,
              const cuComplex *B,
              const int ldb,
              const cuComplex *beta,
              cuComplex *C,
              int ldc)
```

```c
 cusparseStatus_t
cusparseZbsrmm(cusparseHandle_t handle,
              cusparseDirection_t dirA,
              cusparseOperation_t transA,
              cusparseOperation_t transB,
              int mb,
              int n,
              int kb,
              int nnzb,
              const cuDoubleComplex *alpha,
              const cusparseMatDescr_t descrA,
              const cuDoubleComplex *bsrValA,
              const int *bsrRowPtrA,
              const int *bsrColIndA,
              const int blockDim,
              const cuDoubleComplex *B,
              const int ldb,
              const cuDoubleComplex *beta,
              cuDoubleComplex *C,
              int ldc)
```
This function performs one of the following matrix-matrix operations:

\[ C = \alpha \cdot \text{op}(A) \cdot \text{op}(B) + \beta \cdot C \]

\( A \) is an \( mb \times kb \) sparse matrix that is defined in BSR storage format by the three arrays \( \text{bsrValA} \), \( \text{bsrRowPtrA} \), and \( \text{bsrColIndA} \); \( B \) and \( C \) are dense matrices; \( \alpha \) and \( \beta \) are scalars; and

\[
\text{op}(A) = \begin{cases} A & \text{if } \text{transA} = \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{transA} = \text{CUSPARSE\_OPERATION\_TRANSPOSE} \text{ (not supported)} \\
A^H & \text{if } \text{transA} = \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE} \text{ (not supported)}
\end{cases}
\]

and

\[
\text{op}(B) = \begin{cases} B & \text{if } \text{transB} = \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
B^T & \text{if } \text{transB} = \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
B^H & \text{if } \text{transB} = \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE} \text{ (not supported)}
\end{cases}
\]

The matrix type must be \text{CUSPARSE\_MATRIX\_TYPE\_GENERAL}.

The motivation of \text{transpose}(B) is to improve memory access of matrix \( B \). The computational pattern of \( A^*\text{transpose}(B) \) with matrix \( B \) in column-major order is equivalent to \( A*B \) with matrix \( B \) in row-major order.

In practice, no operation in an iterative solver or eigenvalue solver uses \( A^*\text{transpose}(B) \). However, we can perform \( A^*\text{transpose}(\text{transpose}(B)) \) which is the same as \( A*B \). For example, suppose \( A \) is \( mb \times kb \), \( B \) is \( k \times n \) and \( C \) is \( m \times n \), the following code shows usage of \text{cusparseDbsrmm}().

```c
// A is mb*kb, B is k*n and C is m*n
const int m = mb*blockSize;
const int k = kb*blockSize;
const int ldb_B = k; // leading dimension of B
const int ldc   = m; // leading dimension of C
// perform C:=alpha*A*B + beta*C
cusparseSetMatType(descrA, CUSPARSE\_MATRIX\_TYPE\_GENERAL );
cusparseDbsrmm(cusparse_handle,
    CUSPARSE\_DIRECTION\_COLUMN,
    CUSPARSE\_OPERATION\_NON\_TRANSPOSE,
    CUSPARSE\_OPERATION\_NON\_TRANSPOSE,
    mb, n, kb, nnzb, alpha,
    descrA, bsrValA, bsrRowPtrA, bsrColIndA, blockSize,
    B, ldb_B,
    beta, C, ldc);
```
Instead of using $A*B$, our proposal is to transpose $B$ to $B^t$ by first calling 
\texttt{cublas<
\langle T\rangle>geam()}, and then to perform $A^tB^t$.

// step 1: $B^t := \text{transpose}(B)$
const int m = mb*blockSize;
const int k = kb*blockSize;
double *Bt;
const int ldb_Bt = n; // leading dimension of $B$
cudaMalloc((void**)&Bt, sizeof(double)*ldb_Bt*k);
double one  = 1.0;
double zero = 0.0;
cublasSetPointerMode(cublas_handle, CUBLAS_POINTER_MODE_HOST);
cublasDgeam(cublas_handle, CUBLAS_OP_T, CUBLAS_OP_T,
n, k, &one, B, int ldb_B, &zero, B, int ldb_B, Bt, ldb_Bt);

// step 2: perform $\text{C}:=\alpha A^tB^t + \beta \text{C}$
cusparseDbsrmm(cusparse_handle,
CUSPARSE_DIRECTION_COLUMN,
CUSPARSE_OPERATION_NON_TRANSPOSE,
CUSPARSE_OPERATION_TRANSPOSE,
mb, n, kb, nnzb, alpha,
descrA, bsrValA, bsrRowPtrA, bsrColIndA, blockSize,
Bt, ldb_Bt,
beta, C, ldc);

Function \texttt{bsrmm()} is only supported on compute capability 2.0 and above.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>dir</td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation $\text{op}(A)$.</td>
</tr>
<tr>
<td>transB</td>
<td>the operation $\text{op}(B)$.</td>
</tr>
<tr>
<td>mb</td>
<td>number of block rows of sparse matrix $A$.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of dense matrix $\text{op}(B)$ and $A$.</td>
</tr>
<tr>
<td>kb</td>
<td>number of block columns of sparse matrix $A$.</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of non-zero blocks of sparse matrix $A$.</td>
</tr>
<tr>
<td>alpha</td>
<td>&lt;type&gt; scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>bsrValA</td>
<td>&lt;type&gt; array of $\text{nnzb} = \text{bsrRowPtrA}(\text{mb}) - \text{bsrRowPtrA}(0)$ nonzero blocks of matrix $A$.</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>integer array of $\text{mb} + 1$ elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td>bsrColIndA</td>
<td>integer array of $\text{nnzb} = \text{bsrRowPtrA}(\text{mb}) - \text{bsrRowPtrA}(0)$ column indices of the nonzero blocks of matrix $A$.</td>
</tr>
</tbody>
</table>
blockDim | block dimension of sparse matrix \( A \), larger than zero.

\( B \) | array of dimensions \((\text{ldb}, n)\) if \( \text{op}(B) = B \) and \((\text{ldb}, k)\) otherwise.

\( \text{ldb} \) | leading dimension of \( B \). If \( \text{op}(B) = B \), it must be at least \( \max(1, k) \); if \( \text{op}(B) \neq B \), it must be at least \( \max(1, n) \).

\( \beta \) | \(<\text{type}>\) scalar used for multiplication. If \( \beta \) is zero, \( C \) does not have to be a valid input.

\( C \) | array of dimensions \((\text{ldc}, n)\).

\( \text{ldc} \) | leading dimension of \( C \). It must be at least \( \max(1, m) \) if \( \text{op}(A) = A \) and at least \( \max(1, k) \) otherwise.

Output

\( C \) | \(<\text{type}>\) updated array of dimensions \((\text{ldc}, n)\).

Status Returned

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully.
<table>
<thead>
<tr>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>Either invalid parameters were passed ((m, n, k, nnz&lt;0; \text{or } \text{ldb and ldc are incorrect})). Either invalid or unsupported operations were passed ((\text{op}(A) \text{ is different from CUSPARSE_OPERATION_NON_TRANSPOSE, or } \text{op}(B) \text{ is different from CUSPARSE_OPERATION_NON_TRANSPOSE or CUSPARSE_OPERATION_TRANSPOSE})).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>if device is below compute capability 2.0.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>only CUSPARSE_MATRIX_TYPE_GENERAL is supported otherwise.</td>
</tr>
</tbody>
</table>
8.10. \texttt{cusparse<t>bsrsm2\_bufferSize()}

cusparseStatus_t
cusparseSbsrsm2\_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
float *bsrSortedValA,
const int *bsrSortedRowPtrA,
const int *bsrSortedColIndA,
int blockDim,
bsrsm2Info_t info,
int *pBufferSizeInBytes)

cusparseStatus_t
cusparseDbsrsm2\_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
double *bsrSortedValA,
const int *bsrSortedRowPtrA,
const int *bsrSortedColIndA,
int blockDim,
bsrsm2Info_t info,
int *pBufferSizeInBytes)

cusparseStatus_t
cusparseCbsrsm2\_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
cuComplex *bsrSortedValA,
const int *bsrSortedRowPtrA,
const int *bsrSortedColIndA,
int blockDim,
bsrsm2Info_t info,
int *pBufferSizeInBytes)

cusparseStatus_t
cusparseZbsrsm2\_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
cuDoubleComplex *bsrSortedValA,
const int *bsrSortedRowPtrA,
const int *bsrSortedColIndA,
int blockDim,
bsrsm2Info_t info,
int *pBufferSizeInBytes)
This function returns size of buffer used in \texttt{bsrsm2()}, a new sparse triangular linear system \( \text{op}(A) \cdot \text{op}(X) = \alpha \cdot \text{op}(B) \).

\( A \) is an \((mb \times \text{blockDim}) \times (mb \times \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \texttt{bsrValA}, \texttt{bsrRowPtrA}, and \texttt{bsrColIndA}; \( B \) and \( X \) are the right-hand-side and the solution matrices; \( \alpha \) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE}
\end{cases}
\]

Although there are six combinations in terms of parameter \( \text{trans} \) and the upper (and lower) triangular part of \( A \), \texttt{bsrsm2\_bufferSize()} returns the maximum size of the buffer among these combinations. The buffer size depends on dimension \( mb, \text{blockDim} \) and the number of nonzeros of the matrix, \( nnzb \). If the user changes the matrix, it is necessary to call \texttt{bsrsm2\_bufferSize()} again to get the correct buffer size, otherwise a segmentation fault may occur.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>\texttt{handle}</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>\texttt{dirA}</td>
<td>storage format of blocks, either \texttt{CUSPARSE_DIRECTION_ROW} or \texttt{CUSPARSE_DIRECTION_COLUMN}.</td>
</tr>
<tr>
<td>\texttt{transA}</td>
<td>the operation ( \text{op}(A) ).</td>
</tr>
<tr>
<td>\texttt{transX}</td>
<td>the operation ( \text{op}(X) ).</td>
</tr>
<tr>
<td>\texttt{mb}</td>
<td>number of block rows of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{n}</td>
<td>number of columns of matrix ( \text{op}(B) ) and ( \text{op}(X) ).</td>
</tr>
<tr>
<td>\texttt{nnzb}</td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{descrA}</td>
<td>the descriptor of matrix ( A ). The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, while the supported diagonal types are \texttt{CUSPARSE_DIAG_TYPE_UNIT} and \texttt{CUSPARSE_DIAG_TYPE_NON_UNIT}.</td>
</tr>
<tr>
<td>\texttt{bsrValA}</td>
<td>(&lt;\text{type}&gt;) array of ( nnzb ) nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{bsrRowPtrA}</td>
<td>integer array of ( mb + 1 ) elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td>\texttt{bsrColIndA}</td>
<td>integer array of ( nnzb ) column indices of the nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>\texttt{blockDim}</td>
<td>block dimension of sparse matrix ( A ); larger than zero.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>\texttt{info}</td>
<td>record internal states based on different algorithms.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>number of bytes of the buffer used in bsrsm2_analysis() and bsrsm2_solve().</td>
</tr>
<tr>
<td>-------------------</td>
<td>--------------------------------------------------------------------------</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (mb, n, nnzb&lt;=0); base index is not 0 or 1.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
8.11. `cusparse<t>bsrsm2_analysis()`

cusparseStatus_t
cusparseSbsrsm2_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
const float *bsrSortedVal,
const int *bsrSortedRowPtr,
const int *bsrSortedColInd,
int blockDim,
bsrsm2Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer)

cusparseStatus_t
cusparseDbsrsm2_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
const double *bsrSortedVal,
const int *bsrSortedRowPtr,
const int *bsrSortedColInd,
int blockDim,
bsrsm2Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer)

cusparseStatus_t
cusparseCbsrsm2_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
const cuComplex *bsrSortedVal,
const int *bsrSortedRowPtr,
const int *bsrSortedColInd,
int blockDim,
bsrsm2Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer)

cusparseStatus_t
cusparseZbsrsm2_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
cusparseOperation_t transA,
cusparseOperation_t transX,
int mb,
int n,
int nnzb,
const cusparseMatDescr_t descrA,
const cuDoubleComplex *bsrSortedVal,
const int *bsrSortedRowPtr,
const int *bsrSortedColInd,
This function performs the analysis phase of \texttt{bsrsm2()}, a new sparse triangular linear system \(\text{op}(A) \star \text{op}(X) = \alpha \text{op}(B)\).

\(A\) is an \((mb \times \text{blockDim}) \times (mb \times \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \texttt{bsrValA}, \texttt{bsrRowPtrA}, and \texttt{bsrColIndA}; \(B\) and \(X\) are the right-hand-side and the solution matrices; \(\alpha\) is a scalar; and

\[
\text{op}(A) = \begin{cases} 
A & \text{if trans} \equiv \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
A^T & \text{if trans} \equiv \text{CUSPARSE_OPERATION_TRANSPOSE} \\
A^H & \text{if trans} \equiv \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE}
\end{cases}
\]

and

\[
\text{op}(X) = \begin{cases} 
X & \text{if transX} \equiv \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
X^T & \text{if transX} \equiv \text{CUSPARSE_OPERATION_TRANSPOSE} \\
X^H & \text{if transX} \equiv \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} \text{ (not supported)}
\end{cases}
\]

and \(\text{op}(B)\) and \(\text{op}(X)\) are equal.

The block of BSR format is of size \(\text{blockDim} \times \text{blockDim}\), stored in column-major or row-major as determined by parameter \texttt{dirA}, which is either \texttt{CUSPARSE_DIRECTION_ROW} or \texttt{CUSPARSE_DIRECTION_COLUMN}. The matrix type must be \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, and the fill mode and diagonal type are ignored.

It is expected that this function will be executed only once for a given matrix and a particular operation type.

This function requires the buffer size returned by \texttt{bsrsm2_bufferSize()}. The address of \texttt{pBuffer} must be multiple of 128 bytes. If not, \texttt{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Function \texttt{bsrsm2_analysis()} reports a structural zero and computes the level information stored in opaque structure \texttt{info}. The level information can extract more parallelism during a triangular solver. However \texttt{bsrsm2_solve()} can be done without level information. To disable level information, the user needs to specify the policy of the triangular solver as \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}.

Function \texttt{bsrsm2_analysis()} always reports the first structural zero, even if the parameter \texttt{policy} is \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}. Besides, no structural zero is reported if \texttt{CUSPARSE_DIAG_TYPE_UNIT} is specified, even if block \(A(j,j)\) is missing for some \(j\). The user must call \texttt{cusparseXbsrsm2_query_zero_pivot()} to know where the structural zero is.

Even when \texttt{bsrsm2_analysis()} reports a structural zero, the user can still call asynchronously \texttt{bsrsm2_solve()}. In this case, the solve will return a numerical zero in the same position as the structural zero but this result \(X\) is meaningless.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>dirA</td>
<td>storage format of blocks, either \texttt{CUSPARSE_DIRECTION_ROW} or \texttt{CUSPARSE_DIRECTION_COLUMN}.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation (\text{op}(A)).</td>
</tr>
<tr>
<td>transX</td>
<td>the operation (\text{op}(B)) and (\text{op}(X)).</td>
</tr>
</tbody>
</table>
mb  
number of block rows of matrix A.

n  
number of columns of matrix op(B) and op(X).

nnzb  
number of non-zero blocks of matrix A.

descrA  
the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types are CUSPARSE_DIAG_TYPE_UNIT and CUSPARSE_DIAG_TYPE_NON_UNIT.

bsrValA  
<type> array of nnzb (= bsrRowPtrA(mb) - bsrRowPtrA(0)) nonzero blocks of matrix A.

bsrRowPtrA  
integer array of mb + 1 elements that contains the start of every block row and the end of the last block row plus one.

bsrColIndA  
integer array of nnzb (= bsrRowPtrA(mb) - bsrRowPtrA(0)) column indices of the nonzero blocks of matrix A.

blockDim  
block dimension of sparse matrix A; larger than zero.

info  
structure initialized using cusparseCreateBsrsm2Info.

policy  
The supported policies are CUSPARSE_SOLVE_POLICY_NO_LEVEL and CUSPARSE_SOLVE_POLICY_USE_LEVEL.

 pBuffer  
buffer allocated by the user; the size is return by bsrsm2_bufferSize().

Output

descrA  
structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged).

Status Returned

CUSPARSE_STATUS_SUCCESS  
the operation completed successfully.

CUSPARSE_STATUS_NOT_INITIALIZED  
the library was not initialized.

CUSPARSE_STATUS_ALLOC_FAILED  
the resources could not be allocated.

CUSPARSE_STATUS_INVALID_VALUE  
Either invalid parameters were passed (mb, n, nnzb<=0).
Either invalid or unsupported operations were passed (op(B) and op(X) are different from CUSPARSE_OPERATION_NON_TRANSPOSE or CUSPARSE_OPERATION_TRANSPOSE).

CUSPARSE_STATUS_ARCH_MISMATCH  
the device only supports compute capability 2.0 and above.

CUSPARSE_STATUS_EXECUTION_FAILED  
the function failed to launch on the GPU.

CUSPARSE_STATUS_INTERNAL_ERROR  
an internal operation failed.
| CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported. |
8.12. `cusparse<t>bsrsm2_solve()`

```
cusparseStatus_t
cusparseSbsrsm2_solve(cusparseHandle_t handle,
            cusparseDirection_t dirA,
            cusparseOperation_t transA,
            cusparseOperation_t transX,
            int mb,
            int n,
            int nnzb,
            const float *alpha,
            const cusparseMatDescr_t descrA,
            const float *bsrSortedVal,
            const int *bsrSortedRowPtr,
            const int *bsrSortedColInd,
            int blockDim,
            bsrsm2Info_t info,
            const float *B,
            int ldb,
            float *X,
            int ldx,
            cusparseSolvePolicy_t policy,
            void *pBuffer)
```

```
cusparseStatus_t
cusparseDbsrsm2_solve(cusparseHandle_t handle,
            cusparseDirection_t dirA,
            cusparseOperation_t transA,
            cusparseOperation_t transX,
            int mb,
            int n,
            int nnzb,
            const double *alpha,
            const cusparseMatDescr_t descrA,
            const double *bsrSortedVal,
            const int *bsrSortedRowPtr,
            const int *bsrSortedColInd,
            int blockDim,
            bsrsm2Info_t info,
            const double *B,
            int ldb,
            double *X,
            int ldx,
            cusparseSolvePolicy_t policy,
            void *pBuffer)
```

```
cusparseStatus_t
cusparseCbsrsm2_solve(cusparseHandle_t handle,
            cusparseDirection_t dirA,
            cusparseOperation_t transA,
            cusparseOperation_t transX,
            int mb,
            int n,
            int nnzb,
            const cuComplex *alpha,
            const cusparseMatDescr_t descrA,
            const cuComplex *bsrSortedVal,
            const int *bsrSortedRowPtr,
            const int *bsrSortedColInd,
            int blockDim,
            bsrsm2Info_t info,
            const cuComplex *B,
            int ldb,
            cuComplex *X,
            int ldx,
            cusparseSolvePolicy_t policy,
            void *pBuffer)
```

```
cusparseStatus_t
cusparseZbsrsm2_solve(cusparseHandle_t handle,
            cusparseDirection_t dirA,
            cusparseOperation_t transA,
            cusparseOperation_t transX,
            int mb,
            int n,
            int nnzb,
            const cuDoubleComplex *alpha,
            const cusparseMatDescr_t descrA,
            const cuDoubleComplex *bsrSortedVal,
            const int *bsrSortedRowPtr,
            const int *bsrSortedColInd,
            int blockDim,
            bsrsm2Info_t info,
            const cuDoubleComplex *B,
            int ldb,
            cuDoubleComplex *X,
            int ldx,
            cusparseSolvePolicy_t policy,
            void *pBuffer)
```
This function performs the solve phase of the solution of a sparse triangular linear system:

\[ \text{op}(A) \ast \text{op}(X) = \alpha \ast \text{op}(B) \]

\( A \) is an \((mb \ast \text{blockDim}) \times (mb \ast \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \( \text{bsrValA}, \text{bsrRowPtrA} \), and \( \text{bsrColIndA} \); \( B \) and \( X \) are the right-hand-side and the solution matrices; \( \alpha \) is a scalar, and

\[
\text{op}(A) = \begin{cases} A & \text{if } \text{transA} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\ A^T & \text{if } \text{transA} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\ A^H & \text{if } \text{transA} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE} \end{cases}
\]

and

\[
\text{op}(X) = \begin{cases} X & \text{if } \text{transX} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\ X^T & \text{if } \text{transX} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\ X^H & \text{not supported} \end{cases}
\]

Only \( \text{op}(A) = A \) is supported.

\( \text{op}(B) \) and \( \text{op}(X) \) must be performed in the same way. In other words, if \( \text{op}(B) = B \), \( \text{op}(X) = X \).

The block of BSR format is of size \( \text{blockDim} \ast \text{blockDim} \), stored as column-major or row-major as determined by parameter \( \text{dirA} \), which is either \text{CUSPARSE\_DIRECTION\_ROW} or \text{CUSPARSE\_DIRECTION\_COLUMN}. The matrix type must be \text{CUSPARSE\_MATRIX\_TYPE\_GENERAL}, and the fill mode and diagonal type are ignored. Function \text{bsrsm02\_solve}() can support an arbitrary \( \text{blockDim} \).

This function may be executed multiple times for a given matrix and a particular operation type.

This function requires the buffer size returned by \text{bsrsm2\_bufferSize}().

The address of \( \text{pBuffer} \) must be multiple of 128 bytes. If it is not, \text{CUSPARSE\_STATUS\_INVALID\_VALUE} is returned.

Although \text{bsrsm2\_solve}() can be done without level information, the user still needs to be aware of consistency. If \text{bsrsm2\_analysis}() is called with policy \text{CUSPARSE\_SOLVE\_POLICY\_USE\_LEVEL}, \text{bsrsm2\_solve}() can be run with or without levels. On the other hand, if \text{bsrsm2\_analysis}() is called with \text{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}, \text{bsrsm2\_solve}() can only accept \text{CUSPARSE\_SOLVE\_POLICY\_NO\_LEVEL}; otherwise, \text{CUSPARSE\_STATUS\_INVALID\_VALUE} is returned.

Function \text{bsrsm02\_solve}() has the same behavior as \text{bsrsv02\_solve}(), reporting the first numerical zero, including a structural zero. The user must call \text{cusparseXbsrsm2\_query\_zero\_pivot}() to know where the numerical zero is.

The motivation of \( \text{transpose}(X) \) is to improve the memory access of matrix \( X \). The computational pattern of \( \text{transpose}(X) \) with matrix \( X \) in column-major order is equivalent to \( X \) with matrix \( X \) in row-major order.

In-place is supported and requires that \( B \) and \( X \) point to the same memory block, and \( \text{ldb} = \text{ldx} \).
**handle**

handle to the cuSPARSE library context.

**dirA**

storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.

**transA**

the operation \( \text{op}(A) \).

**transX**

the operation \( \text{op}(B) \) and \( \text{op}(X) \).

**mb**

number of block rows of matrix \( A \).

**n**

number of columns of matrix \( \text{op}(B) \) and \( \text{op}(X) \).

**nnzb**

number of non-zero blocks of matrix \( A \).

**alpha**

<type> scalar used for multiplication.

**descrA**

the descriptor of matrix \( A \). The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL, while the supported diagonal types are CUSPARSE_DIAG_TYPE_UNIT and CUSPARSE_DIAG_TYPE_NON_UNIT.

**bsrValA**

<type> array of \( \text{nnzb} (= \text{bsrRowPtrA}(mb) - \text{bsrRowPtrA}(0)) \) non-zero blocks of matrix \( A \).

**bsrRowPtrA**

integer array of \( \text{mb} \) elements that contains the start of every block row and the end of the last block row plus one.

**bsrColIndA**

integer array of \( \text{nnzb} (= \text{bsrRowPtrA}(mb) - \text{bsrRowPtrA}(0)) \) column indices of the nonzero blocks of matrix \( A \).

**blockDim**

block dimension of sparse matrix \( A \); larger than zero.

**info**

structure initialized using cusparseCreateBsrsm2Info().

**B**

<type> right-hand-side array.

**ldb**

leading dimension of \( B \). If \( \text{op}(B) = B \), \( \text{ldb} >= (mb*\text{blockDim}) \); otherwise, \( \text{ldb} >= n \).

**ldx**

leading dimension of \( X \). If \( \text{op}(X) = X \), then \( \text{ldx} >= (mb*\text{blockDim}) \); otherwise, \( \text{ldx} >= n \).

**policy**

the supported policies are CUSPARSE_SOLVE_POLICY_NO_LEVEL and CUSPARSE_SOLVE_POLICY_USE_LEVEL.

**pBuffer**

buffer allocated by the user; the size is returned by bsrsm2_bufferSize().

**Output**

\( X \)

<type> solution array with leading dimensions \( \text{ldx} \).

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
</tbody>
</table>
8.13. `cusparseXbsrsm2_zeroPivot()`

```c
    cusparseStatus_t
cusparseXbsrsm2_zeroPivot(cusparseHandle_t handle,
                     bsrsm2Info_t info,
                     int *position);
```

If the returned error code is `CUSPARSE_STATUS_ZERO_PIVOT`, `position=j` means $A(j,j)$ is either a structural zero or a numerical zero (singular block). Otherwise `position=-1`.

The `position` can be 0-base or 1-base, the same as the matrix.

Function `cusparseXbsrsm2_zeroPivot()` is a blocking call. It calls `cudaDeviceSynchronize()` to make sure all previous kernels are done.

The `position` can be in the host memory or device memory. The user can set the proper mode with `cusparseSetPointerMode()`.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
</table>
| info       | info contains a structural zero or a numerical zero if the user already called `bsrsm2_analysis()` or `bsrsm2_solve()`.

**Output**

| position   | if no structural or numerical zero, `position` is -1; otherwise, if $A(j,j)$ is missing or $U(j,j)$ is zero, `position=j`. |

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td><code>info</code> is not valid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
8.14. cusparse<t>gemmi()

cusparseStatus_t
cusparseSgemmi(cusparseHandle_t handle,
    int m,
    int n,
    int k,
    int nnz,
    const float *alpha,
    const float *A,
    int lda,
    const float *cscValB,
    const int *cscColPtrB,
    const int *cscRowIndB,
    const float *beta,
    float *C,
    int ldc);

cusparseStatus_t
cusparseDgemmi(cusparseHandle_t handle,
    int m,
    int n,
    int k,
    int nnz,
    const double *alpha,
    const double *A,
    int lda,
    const double *cscValB,
    const int *cscColPtrB,
    const int *cscRowIndB,
    const double *beta,
    double *C,
    int ldc);

cusparseStatus_t
cusparseCgemmi(cusparseHandle_t handle,
    int m,
    int n,
    int k,
    int nnz,
    const cuComplex *alpha,
    const cuComplex *A,
    int lda,
    const cuComplex *cscValB,
    const int *cscColPtrB,
    const int *cscRowIndB,
    const cuComplex *beta,
    cuComplex *C,
    int ldc);

cusparseStatus_t
cusparseZgemmi(cusparseHandle_t handle,
    int m,
    int n,
    int k,
    int nnz,
    const cuDoubleComplex *alpha,
    const cuDoubleComplex *A,
    int lda,
    const cuDoubleComplex *cscValB,
    const int *cscColPtrB,
    const int *cscRowIndB,
    const cuDoubleComplex *beta,
    cuDoubleComplex *C,
    int ldc);
This function performs the following matrix-matrix operations:

\[ C = \alpha \cdot A \cdot B + \beta \cdot C \]

\( A \) and \( C \) are dense matrices; \( B \) is a \( k \times n \) sparse matrix that is defined in CSC storage format by the three arrays \( \text{cscValB} \), \( \text{cscColPtrB} \), and \( \text{cscRowIndB} \); \( \alpha \) and \( \beta \) are scalars; and

Remark: \( B \) is base-0.

**Input**

<table>
<thead>
<tr>
<th>Input</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrices ( B ) and ( C ).</td>
</tr>
<tr>
<td>k</td>
<td>number of columns of matrix ( A ).</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of sparse matrix ( B ).</td>
</tr>
<tr>
<td>alpha</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication.</td>
</tr>
<tr>
<td>A</td>
<td>array of dimensions ((\text{lda}, k)).</td>
</tr>
<tr>
<td>lda</td>
<td>leading dimension of ( A ). It must be at least ( m ).</td>
</tr>
<tr>
<td>cscValB</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnz} = (\text{cscColPtrB}(k) - \text{cscColPtrB}(0) ) nonzero elements of matrix ( B ).</td>
</tr>
<tr>
<td>cscColPtrB</td>
<td>integer array of ( k + 1 ) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>cscRowIndB</td>
<td>integer array of ( \text{nnz} = (\text{cscColPtrB}(k) - \text{cscColPtrB}(0) ) column indices of the nonzero elements of matrix ( B ).</td>
</tr>
<tr>
<td>beta</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication. If ( \beta ) is zero, ( C ) does not have to be a valid input.</td>
</tr>
<tr>
<td>C</td>
<td>array of dimensions ((\text{ldc}, n)).</td>
</tr>
<tr>
<td>ldc</td>
<td>leading dimension of ( C ). It must be at least ( m ).</td>
</tr>
</tbody>
</table>

**Output**

| C       | \(<\text{type}>\) updated array of dimensions \((\text{ldc}, n)\).         |

**Status Returned**

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>\text{CUSPARSE_STATUS_SUCCESS}</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>\text{CUSPARSE_STATUS_NOT_INITIALIZED}</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>\text{CUSPARSE_STATUS_INVALID_VALUE}</td>
<td>invalid parameters were passed (( m, n, k, \text{nnz}&lt;0 ) or ( \text{lda} ) and ( \text{ldc} ) are incorrect).</td>
</tr>
<tr>
<td>\text{CUSPARSE_STATUS_EXECUTION_FAILED}</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>\text{CUSPARSE_STATUS_INTERNAL_ERROR}</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
This chapter describes the extra routines used to manipulate sparse matrices.
9.1. cusparse<t>csrgeam()

cusparseStatus_t
cusparseXcsrgeamNnz(cusparseHandle_t handle,
   int m,
   int n,
   const cusparseMatDescr_t descrA,
   int nnzA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   const cusparseMatDescr_t descrB,
   int nnzB,
   const int *csrRowPtrB,
   const int *csrColIndB,
   const cusparseMatDescr_t descrC,
   int *csrRowPtrC,
   int *nnzTotalDevHostPtr)

cusparseStatus_t
cusparseScsrgeam(cusparseHandle_t handle,
   int m,
   int n,
   const float *alpha,
   const cusparseMatDescr_t descrA,
   int nnzA,
   const float *csrValA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   const float *beta,
   const cusparseMatDescr_t descrB,
   int nnzB,
   const float *csrValB,
   const int *csrRowPtrB,
   const int *csrColIndB,
   const cusparseMatDescr_t descrC,
   float *csrValC,
   int *csrRowPtrC,
   int *csrColIndC)

cusparseStatus_t
cusparseDcsrgeam(cusparseHandle_t handle,
   int m,
   int n,
   const double *alpha,
   const cusparseMatDescr_t descrA,
   int nnzA,
   const double *csrValA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   const double *beta,
   const cusparseMatDescr_t descrB,
   int nnzB,
   const double *csrValB,
   const int *csrRowPtrB,
   const int *csrColIndB,
   const cusparseMatDescr_t descrC,
   double *csrValC,
   int *csrRowPtrC,
   int *csrColIndC)

cusparseStatus_t
cusparseCcsrgeam(cusparseHandle_t handle,
   int m,
   int n,
   const cuComplex *alpha,
   const cusparseMatDescr_t descrA,
   int nnzA,
   const cuComplex *csrValA,
This function performs the following matrix-matrix operation:

$$C = \alpha A + \beta B$$

where $A$, $B$, and $C$ are $m \times n$ sparse matrices (defined in CSR storage format by the three arrays $\text{csrValA}$, $\text{csrValB}$, $\text{csrValC}$, $\text{csrRowPtrA}$, $\text{csrRowPtrB}$, $\text{csrRowPtrC}$, and $\text{csrColIndA}$, $\text{csrColIndB}$, $\text{csrColIndC}$ respectively), and $\alpha$ and $\beta$ are scalars. Since $A$ and $B$ have different sparsity patterns, cuSPARSE adopts a two-step approach to complete sparse matrix $C$. In the first step, the user allocates $\text{csrRowPtrC}$ of $m + 1$ elements and uses function $\text{cusparseXcsrgeamNnz()}$ to determine $\text{csrRowPtrC}$ and the total number of nonzero elements. In the second step, the user gathers $\text{nnzC}$ (number of nonzero elements of matrix $C$) from either ($\text{nnzC}=\text{nnzTotalDevHostPtr}$) or ($\text{nnzC}=\text{csrRowPtrC}(m)-\text{csrRowPtrC}(0)$) and allocates $\text{csrValC}$, $\text{csrColIndC}$ of $\text{nnzC}$ elements respectively, then finally calls function $\text{cusparse}[S|D|C|Z]\text{csrgeam()}$ to complete matrix $C$.

The general procedure is as follows:

```c
int baseC, nnzC;

// nnzTotalDevHostPtr points to host memory
int *nnzTotalDevHostPtr = &nnzC;
cusparseSetPointerMode(handle, CUSPARSE_POINTER_MODE_HOST);
cudaMalloc((void**)&csrRowPtrC, sizeof(int)*(m+1));
cusparseXcsrgeamNnz(handle, m, n,
    descrA, nnzA, csrRowPtrA, csrColIndA,
    descrB, nnzB, csrRowPtrB, csrColIndB,
    descrC, csrRowPtrC, nnzTotalDevHostPtr);
if (NULL != nnzTotalDevHostPtr){
    nnzC = *nnzTotalDevHostPtr;
} else{
    cudaMemcpy(&nnzC, csrRowPtrC+m, sizeof(int), cudaMemcpyDeviceToHost);
    cudaMemcpy(&baseC, csrRowPtrC, sizeof(int), cudaMemcpyDeviceToHost);
    nnzC -= baseC;
}
cudaMalloc((void**)&csrColIndC, sizeof(int)*nnzC);
cudaMalloc((void**)&csrValC, sizeof(float)*nnzC);
cusparseScsrgeam(handle, m, n,
    alpha,
    descrA, nnzA,
    csrValA, csrRowPtrA, csrColIndA,
    beta,
    descrB, nnzB,
    csrValB, csrRowPtrB, csrColIndB,
    descrC,
    csrValC, csrRowPtrC, csrColIndC);
```

Several comments on $\text{csrgeam()}$:

- The other three combinations, NT, TN, and TT, are not supported by cuSPARSE. In order to do any one of the three, the user should use the routine $\text{csr2csc()}$ to convert $A | B$ to $A^T | B^T$.
- Only $\text{CUSPARSE_MATRIX_TYPE GENERAL}$ is supported. If either $A$ or $B$ is symmetric or Hermitian, then the user must extend the matrix to a full one and reconfigure the $\text{MatrixType}$ field of the descriptor to $\text{CUSPARSE_MATRIX_TYPE_GENERAL}$.
- If the sparsity pattern of matrix $C$ is known, the user can skip the call to function $\text{cusparseXcsrgeamNnz()}$. For example, suppose that the user has an iterative algorithm which would update $A$ and $B$ iteratively but keep the sparsity patterns.
The user can call function `cusparseXcsrgemMnnz()` once to set up the sparsity pattern of $C$, then call function `cusparse*[S|D|C|Z]geam()` only for each iteration.

- The pointers `alpha` and `beta` must be valid.
- When `alpha` or `beta` is zero, it is not considered a special case by cuSPARSE. The sparsity pattern of $C$ is independent of the value of `alpha` and `beta`. If the user wants $C = 0 \times A + 1 \times B^T$, then `csr2csc()` is better than `csrgem()`.

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>number of rows of sparse matrix $A, B, C$.</td>
</tr>
<tr>
<td><code>n</code></td>
<td>number of columns of sparse matrix $A, B, C$.</td>
</tr>
<tr>
<td><code>alpha</code></td>
<td><code>&lt;type&gt;</code> scalar used for multiplication.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix $A$. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> only.</td>
</tr>
<tr>
<td><code>nnzA</code></td>
<td>number of nonzero elements of sparse matrix $A$.</td>
</tr>
<tr>
<td><code>csrValA</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnzA</code> (= <code>csrRowPtrA(m) - csrRowPtrA(0)</code>) nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td><code>csrRowPtrA</code></td>
<td>integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><code>csrColIndA</code></td>
<td>integer array of <code>nnzA</code> (= <code>csrRowPtrA(m) - csrRowPtrA(0)</code>) column indices of the nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td><code>beta</code></td>
<td><code>&lt;type&gt;</code> scalar used for multiplication. If <code>beta</code> is zero, $y$ does not have to be a valid input.</td>
</tr>
<tr>
<td><code>descrB</code></td>
<td>the descriptor of matrix $B$. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> only.</td>
</tr>
<tr>
<td><code>nnzB</code></td>
<td>number of nonzero elements of sparse matrix $B$.</td>
</tr>
<tr>
<td><code>csrValB</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnzB</code> (= <code>csrRowPtrB(m) - csrRowPtrB(0)</code>) nonzero elements of matrix $B$.</td>
</tr>
<tr>
<td><code>csrRowPtrB</code></td>
<td>integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><code>csrColIndB</code></td>
<td>integer array of <code>nnzB</code> (= <code>csrRowPtrB(m) - csrRowPtrB(0)</code>) column indices of the nonzero elements of matrix $B$.</td>
</tr>
<tr>
<td><code>descrC</code></td>
<td>the descriptor of matrix $C$. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> only.</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>csrValC</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnzC</code> (= <code>csrRowPtrC(m) - csrRowPtrC(0)</code>) nonzero elements of matrix $C$.</td>
</tr>
<tr>
<td><code>csrRowPtrC</code></td>
<td>integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><strong>csrColIndC</strong></td>
<td>integer array of ( \text{nnzC} = \text{csrRowPtrC}(m) - \text{csrRowPtrC}(0) ) column indices of the nonzero elements of matrix ( C ).</td>
</tr>
<tr>
<td>------------------------</td>
<td>----------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td><strong>nnzTotalDevHostPtr</strong></td>
<td>total number of nonzero elements in device or host memory. It is equal to ((\text{csrRowPtrC}(m) - \text{csrRowPtrC}(0))).</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th><strong>CUSPARSE_STATUS_SUCCESS</strong></th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>CUSPARSE_STATUS_NOT_INITIALIZED</strong></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_ALLOC_FAILED</strong></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_INVALID_VALUE</strong></td>
<td>invalid parameters were passed ((m, n, \text{nnz}&lt;0, \text{IndexBase of descrA, descrB, descrC is not base-0 or base-1, or alpha or beta is nil})).</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_ARCH_MISMATCH</strong></td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_EXECUTION_FAILED</strong></td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</strong></td>
<td>the matrix type is not supported.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_INTERNAL_ERROR</strong></td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
9.2. `cusparse<t>csrgeam2()`

cusparseStatus_t
cusparseScsrgeam2_bufferSizeExt(  
cusparseHandle_t handle,  
int m,  
int n,  
const float *alpha,  
const cusparseMatDescr_t descrA,  
int nnzA,  
const float *csrSortedValA,  
const int *csrSortedRowPtrA,  
const int *csrSortedColIndA,  
const float *beta,  
const cusparseMatDescr_t descrB,  
int nnzB,  
const float *csrSortedValB,  
const int *csrSortedRowPtrB,  
const int *csrSortedColIndB,  
const cusparseMatDescr_t descrC,  
const float *csrSortedValC,  
const int *csrSortedRowPtrC,  
const int *csrSortedColIndC,  
sizet *pBufferSizeInBytes );

cusparseStatus_t
cusparseDcsrgeam2_bufferSizeExt(  
cusparseHandle_t handle,  
int m,  
int n,  
const double *alpha,  
const cusparseMatDescr_t descrA,  
int nnzA,  
const double *csrSortedValA,  
const int *csrSortedRowPtrA,  
const int *csrSortedColIndA,  
const double *beta,  
const cusparseMatDescr_t descrB,  
int nnzB,  
const double *csrSortedValB,  
const int *csrSortedRowPtrB,  
const int *csrSortedColIndB,  
const cusparseMatDescr_t descrC,  
const double *csrSortedValC,  
const int *csrSortedRowPtrC,  
const int *csrSortedColIndC,  
sizet *pBufferSizeInBytes );

cusparseStatus_t
cusparseCcsrgeam2_bufferSizeExt(  
cusparseHandle_t handle,  
int m,  
int n,  
const cuComplex *alpha,  
const cusparseMatDescr_t descrA,  
int nnzA,  
const cuComplex *csrSortedValA,  
const int *csrSortedRowPtrA,  
const int *csrSortedColIndA,  
const cuComplex *beta,  
const cusparseMatDescr_t descrB,  
int nnzB,  
const cuComplex *csrSortedValB,  
const int *csrSortedRowPtrB,  
const cuComplex *csrSortedValC,  
const int *csrSortedRowPtrC,  
const int *csrSortedColIndC,  
const cuComplex *csrSortedValC,  
const int *csrSortedRowPtrC,  
const int *csrSortedColIndC,  
sizet *pBufferSizeInBytes );

cusparseStatus_t
cusparseZcsrgeam2_bufferSizeExt(  
cusparseHandle_t handle,  
int m,  
int n,  
const cuDoubleComplex *alpha,  
const cusparseMatDescr_t descrA,  
int nnzA,  
const cuDoubleComplex *csrSortedValA,  
const int *csrSortedRowPtrA,  
const int *csrSortedColIndA,  
const cuDoubleComplex *beta,  
const cusparseMatDescr_t descrB,  
int nnzB,  
const cuDoubleComplex *csrSortedValB,  
const int *csrSortedRowPtrB,  
const cuDoubleComplex *csrSortedValC,  
const int *csrSortedRowPtrC,  
const int *csrSortedColIndC,  
sizet *pBufferSizeInBytes );
This function performs the following matrix-matrix operation

\[ C = \alpha A + \beta B \]

where \( A, B, \) and \( C \) are \( m \times n \) sparse matrices (defined in CSR storage format by the three arrays \( \text{csrValA} | \text{csrValB} | \text{csrValC}, \text{csrRowPtrA} | \text{csrRowPtrB} | \text{csrRowPtrC}, \) and \( \text{csrColIndA} | \text{csrColIndB} | \text{csrColIndC} \) respectively), and \( \alpha \) and \( \beta \) are scalars. Since \( A \) and \( B \) have different sparsity patterns, cuSPARSE adopts a two-step approach to complete the sparse matrix \( C \). In the first step, the user allocates \( \text{csrRowPtrC} \) of \( m+1 \) elements and uses function \( \text{cusparseXcsrgeam2Nnz()} \) to determine \( \text{csrRowPtrC} \) and the total number of nonzero elements. In the second step, the user gathers \( nnzC \) (number of nonzero elements of matrix \( C \)) from either \( (nnzC=*\text{nnzTotalDevHostPtr}) \) or \( (nnzC=\text{csrRowPtrC}(m)-\text{csrRowPtrC}(0)) \) and allocates \( \text{csrValC}, \text{csrColIndC} \) of \( nnzC \) elements respectively, then finally calls function \( \text{cusparse}[S|D|C]Z\text{csrgeam2()} \) to complete matrix \( C \).

The general procedure is as follows:

```c
int baseC, nnzC;
/* alpha, nnzTotalDevHostPtr points to host memory */
size_t BufferSizeInBytes;
char *buffer = NULL;
int *nnzTotalDevHostPtr = &nnzC;
cusparseSetPointerMode(handle, CUSPARSE_POINTER_MODE_HOST);
cudaMalloc((void**)&csrRowPtrC, sizeof(int)*(m+1));
/* prepare buffer */
cusparseScsrgeam2_bufferSizeExt(handle, m, n,
     alpha,
     descrA, nnzA,
     csrValA, csrRowPtrA, csrColIndA,
     beta,
     descrB, nnzB,
     csrValB, csrRowPtrB, csrColIndB,
     descrC,
     csrValC, csrRowPtrC, csrColIndC
     &BufferSizeInBytes);
cudaMalloc((void**)&buffer, sizeof(char)*BufferSizeInBytes);
 cusparseXcsrgeam2Nnz(handle, m, n,
     descrA, nnzA, csrRowPtrA, csrColIndA,
     descrB, nnzB, csrRowPtrB, csrColIndB,
     descrC, csrRowPtrC, nnzTotalDevHostPtr,
     buffer);
if (NULL != nnzTotalDevHostPtr){
    nnzC = *nnzTotalDevHostPtr;
} else{
    cudaMemcpy(&nnzC, csrRowPtrC+m, sizeof(int), cudaMemcpyDeviceToHost);
    cudaMemcpy(&baseC, csrRowPtrC, sizeof(int), cudaMemcpyDeviceToHost);
    nnzC -= baseC;
}
cudaMalloc((void**)&csrColIndC, sizeof(int)*nnzC);
cudaMalloc((void**)&csrValC, sizeof(float)*nnzC);
cusparseScsrgeam2(handle, m, n,
     alpha,
     descrA, nnzA,
     csrValA, csrRowPtrA, csrColIndA,
     beta,
     descrB, nnzB,
     csrValB, csrRowPtrB, csrColIndB,
     descrC,
     csrValC, csrRowPtrC, csrColIndC
     buffer);
```
Several comments on \texttt{csrgeam2}:

- The other three combinations, NT, TN, and TT, are not supported by cuSPARSE. In order to do any one of the three, the user should use the routine \texttt{csr2csc} to convert \(A|B\) to \(A^T|B^T\).
- Only \texttt{CUSPARSE\_MATRIX\_TYPE\_GENERAL} is supported. If either \(A\) or \(B\) is symmetric or Hermitian, then the user must extend the matrix to a full one and reconfigure the \texttt{MatrixType} field of the descriptor to \texttt{CUSPARSE\_MATRIX\_TYPE\_GENERAL}.
- If the sparsity pattern of matrix \(C\) is known, the user can skip the call to function \texttt{cusparseXcsrgeam2Nnz()}. For example, suppose that the user has an iterative algorithm which would update \(A\) and \(B\) iteratively but keep the sparsity patterns. The user can call function \texttt{cusparseXcsrgeam2Nnz()} once to set up the sparsity pattern of \(C\), then call function \texttt{cusparse[S|D|C|Z]geam()} only for each iteration.
- The pointers \texttt{alpha} and \texttt{beta} must be valid.
- When \texttt{alpha} or \texttt{beta} is zero, it is not considered a special case by cuSPARSE. The sparsity pattern of \(C\) is independent of the value of \texttt{alpha} and \texttt{beta}. If the user wants \(C = 0 \times A + 1 \times B^T\), then \texttt{csr2csc()} is better than \texttt{csrgeam2()}. 
- \texttt{csrgeam2()} is the same as \texttt{csrgeam()} except \texttt{csrgeam2()} needs explicit buffer where \texttt{csrgeam()} allocates the buffer internally.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of sparse matrix (A,B,C).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of sparse matrix (A,B,C).</td>
</tr>
<tr>
<td>alpha</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix (A). The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL} only.</td>
</tr>
<tr>
<td>nnzA</td>
<td>number of nonzero elements of sparse matrix (A).</td>
</tr>
<tr>
<td>csrValA</td>
<td>(&lt;\text{type}&gt;) array of (\text{nnzA} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)) nonzero elements of matrix (A).</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of (m + 1) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of (\text{nnzA} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)) column indices of the nonzero elements of matrix (A).</td>
</tr>
<tr>
<td>beta</td>
<td>(&lt;\text{type}&gt;) scalar used for multiplication. If \texttt{beta} is zero, (y) does not have to be a valid input.</td>
</tr>
<tr>
<td>descrB</td>
<td>the descriptor of matrix (B). The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL} only.</td>
</tr>
<tr>
<td>nnzB</td>
<td>number of nonzero elements of sparse matrix (B).</td>
</tr>
<tr>
<td>csrValB</td>
<td>(&lt;\text{type}&gt;) array of (\text{nnzB} = \text{csrRowPtrB}(m) - \text{csrRowPtrB}(0)) nonzero elements of matrix (B).</td>
</tr>
</tbody>
</table>
### csrRowPtrB
- **integer array of \( m + 1 \) elements that contains the start of every row and the end of the last row plus one.**

### csrColIndB
- **integer array of \( \text{nnzB} = \text{csrRowPtrB}(m) - \text{csrRowPtrB}(0) \) column indices of the nonzero elements of matrix B.**

### descrC
- **the descriptor of matrix C. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL only.**

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrValC</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnzC} = \text{csrRowPtrC}(m) - \text{csrRowPtrC}(0) ) nonzero elements of matrix C.</td>
</tr>
<tr>
<td>csrRowPtrC</td>
<td>( \text{integer array of } m + 1 \text{ elements that contains the start of every row and the end of the last row plus one.} )</td>
</tr>
<tr>
<td>csrColIndC</td>
<td>( \text{integer array of } \text{nnzC} = \text{csrRowPtrC}(m) - \text{csrRowPtrC}(0) ) column indices of the nonzero elements of matrix C.</td>
</tr>
<tr>
<td>nnzTotalDevHostPtr</td>
<td>( \text{total number of nonzero elements in device or host memory. It is equal to } (\text{csrRowPtrC}(m) - \text{csrRowPtrC}(0)). )</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ((m,n,\text{nnz}&lt;0, \text{IndexBase of descrA, descrB, descrC is not base-0 or base-1, or alpha or beta is nil}).)</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
9.3. cusparse<t>csrgemm()

cusparseStatus_t
CUSParsExcsrgemmNnz(cusparseHandle_t handle,
cusparseOperation_t transA,
cusparseOperation_t transB,
int m,
int n,
int k,
const cusparseMatDescr_t descrA,
const int nnzA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparseMatDescr_t descrB,
const int nnzB,
const int *csrRowPtrB,
const int *csrColIndB,
cusparseMatDescr_t descrC,
int *csrRowPtrC,
int *nnzTotalDevHostPtr )

cusparseStatus_t
CUSParsExcsrgemm(cusparseHandle_t handle,
cusparseOperation_t transA,
cusparseOperation_t transB,
int m,
int n,
int k,
const cusparseMatDescr_t descrA,
const int nnzA,
const float *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparseMatDescr_t descrB,
const int nnzB,
const float *csrValB,
const int *csrRowPtrB,
const int *csrColIndB,
cusparseMatDescr_t descrC,
float *csrValC,
const int *csrRowPtrC,
int *csrColIndC )

cusparseStatus_t
CUSParsExcsrgemm(cusparseHandle_t handle,
cusparseOperation_t transA,
cusparseOperation_t transB,
int m,
int n,
int k,
const cusparseMatDescr_t descrA,
const int nnzA,
const double *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparseMatDescr_t descrB,
const int nnzB,
const double *csrValB,
const int *csrRowPtrB,
const int *csrColIndB,
cusparseMatDescr_t descrC,
double *csrValC,
const int *csrRowPtrC,
int *csrColIndC )

cusparseStatus_t
CUSParsExcsrgemm(cusparseHandle_t handle,
cusparseOperation_t transA,
This function performs the following matrix-matrix operation:

\[ C = \text{op}(A) \times \text{op}(B) \]

where \( \text{op}(A) \), \( \text{op}(B) \) and \( C \) are \( m \times k \), \( k \times n \), and \( m \times n \) sparse matrices (defined in CSR storage format by the three arrays \( \text{csrValA} \mid \text{csrValB} \mid \text{csrValC} \), \( \text{csrRowPtrA} \mid \text{csrRowPtrB} \mid \text{csrRowPtrC} \), and \( \text{csrColIndA} \mid \text{csrColIndB} \mid \text{csrColIndC} \) respectively. The operation is defined by

\[
\text{op}(A) = \begin{cases} 
A & \text{if trans} = \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
A^T & \text{if trans} = \text{CUSPARSE_OPERATION_TRANSPOSE}
\end{cases}
\]

There are four versions, NN, NT, TN, and TT. NN stands for \( C = A \times B \), NT stands for \( C = A \times B^T \), TN stands for \( C = A^T \times B \) and TT stands for \( C = A^T \times B^T \).

The cuSPARSE library adopts a two-step approach to complete sparse matrix. In the first step, the user allocates \( \text{csrRowPtrC} \) of \( m+1 \) elements and uses the function \( \text{cusparseXcsrgemmNnz()} \) to determine \( \text{csrRowPtrC} \) and the total number of nonzero elements. In the second step, the user gathers \( \text{nnzC} \) (the number of nonzero elements of matrix \( C \)) from either \( \text{nnzC} = *\text{nnzTotalDevHostPtr} \) or \( \text{nnzC} = \text{csrRowPtrC}(m) - \text{csrRowPtrC}(0) \) and allocates \( \text{csrValC} \) and \( \text{csrColIndC} \) of \( \text{nnzC} \) elements respectively, then finally calls function \( \text{cusparse[S|D|C|Z]csrgemm()} \) to complete matrix \( C \).

The general procedure is as follows:

```c
int baseC, nnzC;
// nnzTotalDevHostPtr points to host memory
int *nnzTotalDevHostPtr = &nnzC;
cusparseSetPointerMode(handle, CUSPARSE_POINTER_MODE_HOST);
cudaMalloc((void**)&csrRowPtrC, sizeof(int)*(m+1));
cusparseXcsrgemmNnz(handle, transA, transB, m, n, k,
    descrA, nnzA, csrRowPtrA, csrColIndA,
    descrB, nnzB, csrRowPtrB, csrColIndB,
    descrC, csrRowPtrC, nnzTotalDevHostPtr);
if (NULL != nnzTotalDevHostPtr){
    nnzC = *nnzTotalDevHostPtr;
} else{
    cudaMemcpy(&nnzC, csrRowPtrC+m, sizeof(int), cudaMemcpyDeviceToHost);
    cudaMemcpy(&baseC, csrRowPtrC, sizeof(int), cudaMemcpyDeviceToHost);
    nnzC -= baseC;
}
cudaMalloc((void**)&csrColIndC, sizeof(int)*nnzC);
cudaMalloc((void**)&csrValC, sizeof(float)*nnzC);
cusparseScsrgemm(handle, transA, transB, m, n, k,
    descrA, nnzA, csrValA, csrRowPtrA, csrColIndA,
    descrB, nnzB, csrValB, csrRowPtrB, csrColIndB,
    descrC, csrValC, csrRowPtrC, csrColIndC);
```

Several comments on \( \text{csrgemm()} \):

- Although NN, NT, TN and TT are supported, only the NN version is implemented. For the NT, TN and TT versions, \( \text{csr2csc()} \) is used to transpose the relevant matrices, followed by a call to the NN version of \( \text{csrgemm()} \).
- The NN version needs working space of size \( \text{nnzA} \) integers at least.
Only **CUSPARSE_MATRIX_TYPE_GENERAL** is supported. If either A or B is symmetric or Hermitian, the user must extend the matrix to a full one and reconfigure the `MatrixType` field descriptor to **CUSPARSE_MATRIX_TYPE_GENERAL**.

Only devices of compute capability 2.0 or above are supported.

**Input**

<table>
<thead>
<tr>
<th>Handle</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation op(A)</td>
</tr>
<tr>
<td>transB</td>
<td>the operation op(B)</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of sparse matrix op(A) and C.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of sparse matrix op(B) and C.</td>
</tr>
<tr>
<td>k</td>
<td>number of columns/rows of sparse matrix op(A) / op(B).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is <strong>CUSPARSE_MATRIX_TYPE_GENERAL</strong> only.</td>
</tr>
<tr>
<td>nnzA</td>
<td>number of nonzero elements of sparse matrix A.</td>
</tr>
<tr>
<td>csrValA</td>
<td>&lt;type&gt; array of nnzA (= csrRowPtrA(m) - csrRowPtrA(0)) nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of ( m + 1 ) elements that contains the start of every row and the end of the last row plus one. ( m = m ) if transA == <strong>CUSPARSE_OPERATION_NON_TRANSPOSE</strong>, otherwise ( m = k. )</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of nnzA (= csrRowPtrA(m) - csrRowPtrA(0)) column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>descrB</td>
<td>the descriptor of matrix B. The supported matrix type is <strong>CUSPARSE_MATRIX_TYPE_GENERAL</strong> only.</td>
</tr>
<tr>
<td>nnzB</td>
<td>number of nonzero elements of sparse matrix B.</td>
</tr>
<tr>
<td>csrValB</td>
<td>&lt;type&gt; array of nnzB nonzero elements of matrix B.</td>
</tr>
<tr>
<td>csrRowPtrB</td>
<td>integer array of ( k + 1 ) elements that contains the start of every row and the end of the last row plus one. ( k = k ) if transB == <strong>CUSPARSE_OPERATION_NON_TRANSPOSE</strong>, otherwise ( k = n. )</td>
</tr>
<tr>
<td>csrColIndB</td>
<td>integer array of nnzB column indices of the nonzero elements of matrix B.</td>
</tr>
<tr>
<td>descrC</td>
<td>the descriptor of matrix C. The supported matrix type is <strong>CUSPARSE_MATRIX_TYPE_GENERAL</strong> only.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Handle</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrValC</td>
<td>&lt;type&gt; array of nnzC (= csrRowPtrC(m) - csrRowPtrC(0)) nonzero elements of matrix C.</td>
</tr>
<tr>
<td>Variable</td>
<td>Description</td>
</tr>
<tr>
<td>---------------------</td>
<td>------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>csrRowPtrC</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndC</td>
<td>integer array of nnzC (= csrRowPtrC(m) - csrRowPtrC(0)) column indices of the nonzero elements of matrix C.</td>
</tr>
<tr>
<td>nnzTotalDevHostPtr</td>
<td>total number of nonzero elements in device or host memory. It is equal to (csrRowPtrC(m) - csrRowPtrC(0)).</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n, k&lt;0; IndexBase of descrA, descrB, descrC is not base-0 or base-1; or alpha or beta is nil)).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
9.4. cusparse<t>csrgemm2()

cusparseStatus_t
cusparseScsrgemm2_bufferSizeExt(cusparseHandle_t handle,
   int m,
   int n,
   int k,
   const float *alpha,
   const cusparseMatDescr_t descrA,
   int nnzA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   const cusparseMatDescr_t descrB,
   int nnzB,
   const int *csrRowPtrB,
   const int *csrColIndB,
   const double *beta,
   const cusparseMatDescr_t descrD,
   int nnzD,
   const int *csrRowPtrD,
   const int *csrColIndD,
   csrgemm2Info_t info,
   size_t *pBufferSizeInBytes );

cusparseStatus_t
cusparseDcsrgemm2_bufferSizeExt(cusparseHandle_t handle,
   int m,
   int n,
   int k,
   const double *alpha,
   const cusparseMatDescr_t descrA,
   int nnzA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   const cusparseMatDescr_t descrB,
   int nnzB,
   const int *csrRowPtrB,
   const int *csrColIndB,
   const double *beta,
   const cusparseMatDescr_t descrD,
   int nnzD,
   const int *csrRowPtrD,
   const int *csrColIndD,
   csrgemm2Info_t info,
   size_t *pBufferSizeInBytes );

cusparseStatus_t
cusparseScsrgemm2_bufferSizeExt(cusparseHandle_t handle,
   int m,
   int n,
   int k,
   const cuComplex *alpha,
   const cusparseMatDescr_t descrA,
   int nnzA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   const cusparseMatDescr_t descrB,
   int nnzB,
   const int *csrRowPtrB,
   const int *csrColIndB,
   const cuComplex *beta,
   const cusparseMatDescr_t descrD,
   int nnzD,
   const int *csrRowPtrD,
   const int *csrColIndD,
   csrgemm2Info_t info,
   size_t *pBufferSizeInBytes );
This function performs following matrix-matrix operation:

\[ C = \alpha A \ast B + \beta D \]

where \( A, B, D \) and \( C \) are \( m \times k, k \times n, m \times n \) and \( m \times n \) sparse matrices (defined in CSR storage format by the three arrays \( \text{csrValA} | \text{csrValB} | \text{csrValD} | \text{csrValC}, \text{csrRowPtrA} | \text{csrRowPtrB} | \text{csrRowPtrD} | \text{csrRowPtrC}, \text{csrColIndA} | \text{csrColIndB} | \text{csrColIndD} | \text{csrColIndC} \) respectively.

We provide \text{csrgemm2} as a generalization of \text{csrgemm}. It provides more operations in terms of \( \alpha \) and \( \beta \). For example, \( C = -A \ast B + D \) can be done by \text{csrgemm2}.

The \text{csrgemm2} uses \( \alpha \) and \( \beta \) to support the following operations:

<table>
<thead>
<tr>
<th>( \alpha )</th>
<th>( \beta )</th>
<th>operation</th>
</tr>
</thead>
<tbody>
<tr>
<td>NULL</td>
<td>NULL</td>
<td>invalid</td>
</tr>
<tr>
<td>NULL</td>
<td>!NULL</td>
<td>( C = \beta D, A ) \text{ and } B ) are not used</td>
</tr>
<tr>
<td>!NULL</td>
<td>NULL</td>
<td>( C = \alpha A \ast B, D ) is not used</td>
</tr>
<tr>
<td>!NULL</td>
<td>!NULL</td>
<td>( C = \alpha A \ast B + \beta D )</td>
</tr>
</tbody>
</table>

The numerical value of \( \alpha \) and \( \beta \) only affects the numerical values of \( C \), not its sparsity pattern. For example, if \( \alpha \) and \( \beta \) are not zero, the sparsity pattern of \( C \) is union of \( A \ast B \) and \( D \), independent of numerical value of \( \alpha \) and \( \beta \).

The following table shows different operations according to the value of \( m, n \) and \( k \):

<table>
<thead>
<tr>
<th>( m, n, k )</th>
<th>operation</th>
</tr>
</thead>
<tbody>
<tr>
<td>( m &lt; 0 ) or ( n &lt; 0 ) or ( k &lt; 0 )</td>
<td>invalid</td>
</tr>
<tr>
<td>( m ) is 0 or ( n ) is 0</td>
<td>do nothing</td>
</tr>
<tr>
<td>( m &gt; 0 ) and ( n &gt; 0 ) and ( k ) is 0</td>
<td>invalid if ( \beta ) is zero; ( C = \beta D ) if ( \beta ) is not zero.</td>
</tr>
<tr>
<td>( m &gt; 0 ) and ( n &gt; 0 ) and ( k &gt; 0 )</td>
<td>( C = \beta D ) if ( \alpha ) is zero. ( C = \alpha A \ast B ) if ( \beta ) is zero. ( C = \alpha A \ast B + \beta D ) if ( \alpha ) and ( \beta ) are not zero.</td>
</tr>
</tbody>
</table>

This function requires the buffer size returned by \text{csrgemm2_bufferSizeExt}().

The address of \( p\text{Buffer} \) must be multiple of 128 bytes. If it is not, \text{CUSPARSE_STATUS_INVALID_VALUE} is returned.

The cuSPARSE library adopts a two-step approach to complete sparse matrix. In the first step, the user allocates \( \text{csrRowPtrC} \) of \( m+1 \) elements and uses the function \text{cusparseXcsrgemm2Nnz}() to determine \( \text{csrRowPtrC} \) and the total number of nonzero elements. In the second step, the user gathers \( \text{nnzC} \) (the number of nonzero elements of matrix \( C \)) from either \( \text{nnzC} = *\text{nnzTotalDevHostPtr} \) or \( \text{nnzC} = \text{csrRowPtrC}(m) - \text{csrRowPtrC}(0) \) and allocates \( \text{csrValC} \) and \( \text{csrColIndC} \) of \( \text{nnzC} \) elements respectively, then finally calls function \text{cusparse[S]D[C[Z]csrgemm2}() to evaluate matrix \( C \).
The general procedure of $C = -A \times B + D$ is as follows:

```c
// assume matrices A, B and D are ready.
int baseC, nnzC;
csrmm2Info_t info = NULL;
size_t bufferSize;
void *buffer = NULL;
// nnzTotalDevHostPtr points to host memory
int *nnzTotalDevHostPtr = &nnzC;
double alpha = -1.0;
double beta  = 1.0;
cusparseSetPointerMode(handle, CUSPARSE_POINTER_MODE_HOST);

// step 1: create an opaque structure
cusparseCreateCsrmm2Info(&info);

// step 2: allocate buffer for csrmm2Nnz and csrmm2

cusparseCcsrmm2_bufferSizeExt(handle, m, n, k, &alpha,
    descrA, nnzA, csrRowPtrA, csrColIndA,
    descrB, nnzB, csrRowPtrB, csrColIndB,
    &beta,
    descrD, nnzD, csrRowPtrD, csrColIndD,
    info,
    &bufferSize);
cudaMalloc(&buffer, bufferSize);

// step 3: compute csrRowPtrC
cudaMalloc((void**)&csrRowPtrC, sizeof(int)*(m+1));

cusparseXcsrmm2Nnz(handle, m, n, k,
    descrA, nnzA, csrRowPtrA, csrColIndA,
    descrB, nnzB, csrRowPtrB, csrColIndB,
    descrD, nnzD, csrRowPtrD, csrColIndD,
    descrC, csrRowPtrC, nnzTotalDevHostPtr,
    info, buffer);
if (NULL != nnzTotalDevHostPtr){
    nnzC = *nnzTotalDevHostPtr;
} else{
    cudaMemcpy(&nnzC, csrRowPtrC+m, sizeof(int), cudaMemcpyDeviceToHost);
    cudaMemcpy(&baseC, csrRowPtrC, sizeof(int), cudaMemcpyDeviceToHost);
    nnzC -= baseC;
}

// step 4: finish sparsity pattern and value of C
cudaMalloc((void**)&csrColIndC, sizeof(int)*nnzC);
cudaMalloc((void**)&csrValC, sizeof(double)*nnzC);
// Remark: set csrValC to null if only sparsity pattern is required.
cusparseDcsrmm2(handle, m, n, k, &alpha,
    descrA, nnzA, csrValA, csrRowPtrA, csrColIndA,
    descrB, nnzB, csrValB, csrRowPtrB, csrColIndB,
    &beta,
    descrD, nnzD, csrValD, csrRowPtrD, csrColIndD,
    descrC, csrValC, csrRowPtrC, csrColIndC,
    info, buffer);

// step 5: destroy the opaque structure
cusparseDestroyCsrmm2Info(info);
```

Several comments on `csrmm2()`:

- Only the NN version is supported. For other modes, the user has to transpose $A$ or $B$ explicitly.
- Only `CUSPARSE_MATRIX_TYPE_GENERAL` is supported. If either $A$ or $B$ is symmetric or Hermitian, the user must extend the matrix to a full one and reconfigure the `MatrixType` field descriptor to `CUSPARSE_MATRIX_TYPE_GENERAL`. 
- if `csrValC` is zero, only sparsity pattern of `C` is calculated.
- Only devices of compute capability 2.0 or above are supported.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of sparse matrix <code>A</code>, <code>D</code> and <code>C</code>.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of sparse matrix <code>B</code>, <code>D</code> and <code>C</code>.</td>
</tr>
<tr>
<td>k</td>
<td>number of columns/rows of sparse matrix <code>A / B</code>.</td>
</tr>
<tr>
<td>alpha</td>
<td><code>&lt;type&gt;</code> scalar used for multiplication.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix <code>A</code>. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> only.</td>
</tr>
<tr>
<td>nnzA</td>
<td>number of nonzero elements of sparse matrix <code>A</code>.</td>
</tr>
<tr>
<td>csrValA</td>
<td><code>&lt;type&gt;</code> array of <code>nnzA</code> nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of <code>m+1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of <code>nnzA</code> column indices of the nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td>descrB</td>
<td>the descriptor of matrix <code>B</code>. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> only.</td>
</tr>
<tr>
<td>nnzB</td>
<td>number of nonzero elements of sparse matrix <code>B</code>.</td>
</tr>
<tr>
<td>csrValB</td>
<td><code>&lt;type&gt;</code> array of <code>nnzB</code> nonzero elements of matrix <code>B</code>.</td>
</tr>
<tr>
<td>csrRowPtrB</td>
<td>integer array of <code>k+1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndB</td>
<td>integer array of <code>nnzB</code> column indices of the nonzero elements of matrix <code>B</code>.</td>
</tr>
<tr>
<td>descrD</td>
<td>the descriptor of matrix <code>D</code>. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> only.</td>
</tr>
<tr>
<td>nnzD</td>
<td>number of nonzero elements of sparse matrix <code>D</code>.</td>
</tr>
<tr>
<td>csrValD</td>
<td><code>&lt;type&gt;</code> array of <code>nnzD</code> nonzero elements of matrix <code>D</code>.</td>
</tr>
<tr>
<td>csrRowPtrD</td>
<td>integer array of <code>m+1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndD</td>
<td>integer array of <code>nnzD</code> column indices of the nonzero elements of matrix <code>D</code>.</td>
</tr>
<tr>
<td>beta</td>
<td><code>&lt;type&gt;</code> scalar used for multiplication.</td>
</tr>
<tr>
<td>descrC</td>
<td>the descriptor of matrix <code>C</code>. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code> only.</td>
</tr>
<tr>
<td>info</td>
<td>structure with information used in <code>csrgemm2Nnz</code> and <code>csrgemm2</code>.</td>
</tr>
</tbody>
</table>
pBuffer | buffer allocated by the user; the size is returned by `csrgemm2_bufferSizeExt`.

Output

| csrValC | `<type>` array of `nnzC` nonzero elements of matrix C. |
| csrRowPtrC | integer array of `m+1` elements that contains the start of every row and the end of the last row plus one. |
| csrColIndC | integer array of `nnzC` column indices of the nonzero elements of matrix C. |
| pBufferSizeInBytes | number of bytes of the buffer used in `csrgemm2Nnnz` and `csrgemm2`. |
| nnzTotalDevHostPtr | total number of nonzero elements in device or host memory. It is equal to `(csrRowPtrC(m) - csrRowPtrC(0))`. |

Status Returned

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully. |
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized. |
| CUSPARSE_STATUS_ALLOC_FAILED | the resources could not be allocated. |
| CUSPARSE_STATUS_INVALID_VALUE | invalid parameters were passed (`m,n,k<0; IndexBase of descrA, descrB, descrD, descrC` is not base-0 or base-1). |
| CUSPARSE_STATUS_ARCH_MISMATCH | the device does not support double precision. |
| CUSPARSE_STATUS_EXECUTION_FAILED | the function failed to launch on the GPU. |
| CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported. |
| CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed. |
This chapter describes the routines that implement different preconditioners.

In particular, the incomplete factorizations are implemented in two phases. First, during the analysis phase, the sparse triangular matrix is analyzed to determine the dependencies between its elements by calling the appropriate `csrsv_analysis()` function. The analysis is specific to the sparsity pattern of the given matrix and the selected `cusparseOperation_t` type. The information from the analysis phase is stored in the parameter of type `cusparseSolveAnalysisInfo_t` that has been initialized previously with a call to `cusparseCreateSolveAnalysisInfo()`.

Second, during the numerical factorization phase, the given coefficient matrix is factorized using the information stored in the `cusparseSolveAnalysisInfo_t` parameter by calling the appropriate `csrilu0()` or `csric0()` function.

The analysis phase is shared across the sparse triangular solve, and the incomplete factorization and must be performed only once. The resulting information can be passed to the numerical factorization and the sparse triangular solve multiple times.

Finally, once the incomplete factorization and all the sparse triangular solves have completed, the opaque data structure pointed to by the `cusparseSolveAnalysisInfo_t` parameter can be released by calling `cusparseDestroySolveAnalysisInfo()`.

### 10.1. Incomplete Cholesky Factorization: level 0

Different algorithms for $ic0$ are discussed in this section.
10.1.1. cusparse<t>csric0()

This function computes the incomplete-Cholesky factorization with 0 fill-in and no pivoting:

\[ op(A) = R^T R \]

\( A \) is an \( m \times m \) Hermitian/symmetric positive definite sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValM}, \text{csrRowPtrA}, \text{csrColIndA} \); and

\[ op(A) =\begin{cases} A & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\ A^T & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_TRANSPOSE} \\ A^H & \text{if } \text{trans} == \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} \end{cases} \]

Notice that only a lower or upper Hermitian/symmetric part of the matrix \( A \) is actually stored. It is overwritten by the lower or upper triangular factors \( R^T \) and \( R \), respectively. A call to this routine must be preceded by a call to the \text{csrsv_analysis()} routine.
The matrix descriptor for `csrsv_analysis()` and `csric0()` must be the same. Otherwise, runtime error would occur.

This function requires some extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>handle</strong></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><strong>trans</strong></td>
<td>the operation <code>op</code> <code>(A)</code>.</td>
</tr>
<tr>
<td><strong>m</strong></td>
<td>number of rows and columns of matrix <code>A</code>.</td>
</tr>
<tr>
<td><strong>descrA</strong></td>
<td>the descriptor of matrix <code>A</code>. The supported matrix types are <code>CUSPARSE_MATRIX_TYPE_SYMMETRIC</code> and <code>CUSPARSE_MATRIX_TYPE_HERMITIAN</code>. Also, the supported index bases are <code>CUSPARSE_INDEX_BASE_ZERO</code> and <code>CUSPARSE_INDEX_BASE_ONE</code>.</td>
</tr>
<tr>
<td><strong>csrValM</strong></td>
<td><code>&lt;type&gt;</code> array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td><strong>csrRowPtrA</strong></td>
<td>integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><strong>csrColIndA</strong></td>
<td>integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> column indices of the nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td><strong>info</strong></td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>csrValM</strong></td>
<td><code>&lt;type&gt;</code> matrix containing the incomplete-Cholesky lower or upper triangular factor.</td>
</tr>
</tbody>
</table>

**Status Returned**

- `CUSPARSE_STATUS_SUCCESS` | the operation completed successfully. |
- `CUSPARSE_STATUS_NOT_INITIALIZED` | the library was not initialized. |
- `CUSPARSE_STATUS_ALLOC_FAILED` | the resources could not be allocated. |
- `CUSPARSE_STATUS_INVALID_VALUE` | invalid parameters were passed (`m<0`). |
- `CUSPARSE_STATUS_ARCH_MISMATCH` | the device does not support double precision. |
- `CUSPARSE_STATUS_EXECUTION_FAILED` | the function failed to launch on the GPU. |
- `CUSPARSE_STATUS_INTERNAL_ERROR` | an internal operation failed. |
- `CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED` | the matrix type is not supported. |
10.1.2. cusparse<t>csric02_bufferSize()

```c
cusparseStatus_t
cusparseScsric02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csric02Info_t info,
    int *pBufferSizeInBytes);

cusparseStatus_t
cusparseDcsric02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csric02Info_t info,
    int *pBufferSizeInBytes);

cusparseStatus_t
cusparseCcsric02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csric02Info_t info,
    int *pBufferSizeInBytes);

cusparseStatus_t
cusparseZcsric02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csric02Info_t info,
    int *pBufferSizeInBytes);
```

This function returns size of buffer used in computing the incomplete-Cholesky factorization with 0 fill-in and no pivoting:

\[ A \approx LL^H \]

\( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( csrValA, csrRowPtrA, \) and \( csrColIndA. \)
The buffer size depends on dimension $m$ and $nnz$, the number of nonzeros of the matrix. If the user changes the matrix, it is necessary to call `csric02_bufferSize()` again to have the correct buffer size; otherwise, a segmentation fault may occur.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>$m$</td>
<td>number of rows and columns of matrix $A$.</td>
</tr>
<tr>
<td>$nnz$</td>
<td>number of nonzeros of matrix $A$.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. 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>.</td>
</tr>
<tr>
<td>csrValA</td>
<td><code>&lt;type&gt;</code> array of $nnz$ ($= csrRowPtrA(m) - csrRowPtrA(0)$) nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of $nnz$ ($= csrRowPtrA(m) - csrRowPtrA(0)$) column indices of the nonzero elements of matrix $A$.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>record internal states based on different algorithms.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>number of bytes of the buffer used in <code>csric02_analysis()</code> and <code>csric02()</code>.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed ($m,nnz$$\leq$$0$), base index is not 0 or 1.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.1.3. cusparse<t>csric02_analysis()

```c
cusparseStatus_t
cusparseScsric02_analysis(cusparseHandle_t handle,
int m,
int nnz,
const cusparseMatDescr_t descrA,
const float *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
csic02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseDcsric02_analysis(cusparseHandle_t handle,
int m,
int nnz,
const cusparseMatDescr_t descrA,
const double *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
csic02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseCcsric02_analysis(cusparseHandle_t handle,
int m,
int nnz,
const cusparseMatDescr_t descrA,
cuComplex *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
csic02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseZcsric02_analysis(cusparseHandle_t handle,
int m,
int nnz,
const cusparseMatDescr_t descrA,
cuDoubeComplex *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
csic02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);
```

This function performs the analysis phase of the incomplete-Cholesky factorization with 0 fill-in and no pivoting:

\[ A \approx LL^H \]

\( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA}, \text{csrRowPtrA}, \) and \( \text{csrColIndA} \).
This function requires a buffer size returned by `csric02_bufferSize()`. The address of `pBuffer` must be a multiple of 128 bytes. If not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

Function `csric02_analysis()` reports a structural zero and computes level information stored in the opaque structure `info`. The level information can extract more parallelism during incomplete Cholesky factorization. However `csric02()` can be done without level information. To disable level information, the user must specify the policy of `csric02_analysis()` and `csric02()` as `CUSPARSE_SOLVE_POLICY_NO_LEVEL`.

Function `csric02_analysis()` always reports the first structural zero, even if the policy is `CUSPARSE_SOLVE_POLICY_NO_LEVEL`. The user needs to call `cusparseXcsric02_zeroPivot()` to know where the structural zero is.

It is the user's choice whether to call `csric02()` if `csric02_analysis()` reports a structural zero. In this case, the user can still call `csric02()`, which will return a numerical zero at the same position as the structural zero. However the result is meaningless.

**Input**

<table>
<thead>
<tr>
<th>Input</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>number of rows and columns of matrix A.</td>
</tr>
<tr>
<td><code>nnz</code></td>
<td>number of nonzeros of matrix A.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix A. 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>.</td>
</tr>
<tr>
<td><code>csrValA</code></td>
<td><code>&lt;type&gt;</code> array of nnz ( = csrRowPtrA(m) - csrRowPtrA(0) ) nonzero elements of matrix A.</td>
</tr>
<tr>
<td><code>csrRowPtrA</code></td>
<td>integer array of m + 1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><code>csrColIndA</code></td>
<td>integer array of nnz ( = csrRowPtrA(m) - csrRowPtrA(0) ) column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td><code>info</code></td>
<td>structure initialized using <code>cusparseCreateCsric02Info()</code>.</td>
</tr>
<tr>
<td><code>policy</code></td>
<td>the supported policies are <code>CUSPARSE_SOLVE_POLICY_NO_LEVEL</code> and <code>CUSPARSE_SOLVE_POLICY_USE_LEVEL</code>.</td>
</tr>
<tr>
<td><code>pBuffer</code></td>
<td>buffer allocated by the user; the size is returned by <code>csric02_bufferSize()</code>.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Output</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>info</code></td>
<td>number of bytes of the buffer used in <code>csric02_analysis()</code> and <code>csric02()</code>.</td>
</tr>
</tbody>
</table>

**Status Returned**
<table>
<thead>
<tr>
<th>cuSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>cuSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>cuSPARSE_STATUS_ALOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>cuSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, nnz&lt;=0), base index is not 0 or 1.</td>
</tr>
<tr>
<td>cuSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>cuSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>cuSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.1.4. \texttt{cusparse<t>csric02()}

\texttt{cusparseStatus_t}
\texttt{cusparseScsric02(cusparseHandle_t handle,}
\texttt{    int m,}
\texttt{    int nnz,}
\texttt{    const cusparseMatDescr_t descrA,}
\texttt{    float *csrValA_valM,}
\texttt{    const int *csrRowPtrA,}
\texttt{    const int *csrColIndA,}
\texttt{    csric02Info_t info,}
\texttt{    cusparseSolvePolicy_t policy,}
\texttt{    void *pBuffer);}

\texttt{cusparseStatus_t}
\texttt{cusparseDcsric02(cusparseHandle_t handle,}
\texttt{    int m,}
\texttt{    int nnz,}
\texttt{    const cusparseMatDescr_t descrA,}
\texttt{    double *csrValA_valM,}
\texttt{    const int *csrRowPtrA,}
\texttt{    const int *csrColIndA,}
\texttt{    csric02Info_t info,}
\texttt{    cusparseSolvePolicy_t policy,}
\texttt{    void *pBuffer);}

\texttt{cusparseStatus_t}
\texttt{cusparseCcsric02(cusparseHandle_t handle,}
\texttt{    int m,}
\texttt{    int nnz,}
\texttt{    const cusparseMatDescr_t descrA,}
\texttt{    cuComplex *csrValA_valM,}
\texttt{    const int *csrRowPtrA,}
\texttt{    const int *csrColIndA,}
\texttt{    csric02Info_t info,}
\texttt{    cusparseSolvePolicy_t policy,}
\texttt{    void *pBuffer);}

\texttt{cusparseStatus_t}
\texttt{cusparseZcsric02(cusparseHandle_t handle,}
\texttt{    int m,}
\texttt{    int nnz,}
\texttt{    const cusparseMatDescr_t descrA,}
\texttt{    cuDoubleComplex *csrValA_valM,}
\texttt{    const int *csrRowPtrA,}
\texttt{    const int *csrColIndA,}
\texttt{    csric02Info_t info,}
\texttt{    cusparseSolvePolicy_t policy,}
\texttt{    void *pBuffer);}\

This function performs the solve phase of the computing the incomplete-Cholesky factorization with 0 fill-in and no pivoting:

\[
A \approx LL^H
\]
This function requires a buffer size returned by `csric02_bufferSize()`. The address of `pBuffer` must be a multiple of 128 bytes. If not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

Although `csric02()` can be done without level information, the user still needs to be aware of consistency. If `csric02_analysis()` is called with policy `CUSPARSE_SOLVE_POLICY_USE_LEVEL`, `csric02()` can be run with or without levels. On the other hand, if `csric02_analysis()` is called with `CUSPARSE_SOLVE_POLICY_NO_LEVEL`, `csric02()` can only accept `CUSPARSE_SOLVE_POLICY_NO_LEVEL`; otherwise, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

Function `csric02()` reports the first numerical zero, including a structural zero. The user must call `cusparseXcsric02_zeroPivot()` to know where the numerical zero is.

Function `csric02()` only takes the lower triangular part of matrix `A` to perform factorization. The matrix type must be `CUSPARSE_MATRIX_TYPE_GENERAL`, the fill mode and diagonal type are ignored, and the strictly upper triangular part is ignored and never touched. It does not matter if `A` is Hermitian or not. In other words, from the point of view of `csric02()` `A` is Hermitian and only the lower triangular part is provided.

In practice, a positive definite matrix may not have incomplete cholesky factorization. To the best of our knowledge, only matrix `M` can guarantee the existence of incomplete cholesky factorization. If `csric02()` failed cholesky factorization and reported a numerical zero, it is possible that incomplete cholesky factorization does not exist.

For example, suppose `A` is a real `m × m` matrix, the following code solves the precondition system `M*y = x` where `M` is the product of Cholesky factorization `L` and its transpose.
\[ M = L L^H \]

// Suppose that A is m x m sparse matrix represented by CSR format,
// Assumption:
// - handle is already created by cusparseCreate(),
// - (d_csrRowPtr, d_csrColInd, d_csrVal) is CSR of A on device memory,
// - d_x is right hand side vector on device memory,
// - d_y is solution vector on device memory.
// - d_z is intermediate result on device memory.

cusparseMatDescr_t descr_M = 0;
cusparseMatDescr_t descr_L = 0;
csric02Info_t info_M = 0;
csrsv2Info_t info_L = 0;
csrsv2Info_t info_Lt = 0;
int pBufferSize_M;
int pBufferSize_L;
int pBufferSize_Lt;
int pBufferSize;
void *pBuffer = 0;
int structural_zero;
int numerical_zero;
const double alpha = 1.;
const cusparseSolvePolicy_t policy_M = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const cusparseSolvePolicy_t policy_L = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const cusparseSolvePolicy_t policy_Lt = CUSPARSE_SOLVE_POLICY_USE_LEVEL;
const cusparseOperation_t trans_L = CUSPARSE_OPERATION_NON_TRANSPOSE;
const cusparseOperation_t trans_Lt = CUSPARSE_OPERATION_TRANSPOSE;

// step 1: create a descriptor which contains
// - matrix M is base-1
// - matrix L is base-1
// - matrix L is lower triangular
// - matrix L has non-unit diagonal
cusparseCreateMatDescr(&descr_M);
cusparseSetMatIndexBase(descr_M, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descr_M, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseCreateMatDescr(&descr_L);
cusparseSetMatIndexBase(descr_L, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descr_L, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatFillMode(descr_L, CUSPARSE_FILL_MODE_LOWER);
cusparseSetMatDiagType(descr_L, CUSPARSE_DIAG_TYPE_NON_UNIT);

// step 2: create a empty info structure
// we need one info for csric02 and two info's for csrsv2
cusparseCreateCsric02Info(&info_M);
cusparseCreateCsrsv2Info(&info_L);
cusparseCreateCsrsv2Info(&info_Lt);

// step 3: query how much memory used in csric02 and csrsv2, and allocate the
// buffer
cusparseDcsric02_bufferSize(handle, m, nnz,
   descr_M, d_csrVal, d_csrRowPtr, d_csrColInd, info_M, &bufferSize_M);
cusparseDcsrsv2_bufferSize(handle, trans_L, m, nnz,
   descr_L, d_csrVal, d_csrRowPtr, d_csrColInd, info_L, &pBufferSize_L);
cusparseDcsrsv2_bufferSize(handle, trans_Lt, m, nnz,
   descr_L, d_csrVal, d_csrRowPtr, d_csrColInd, info_Lt, &pBufferSize_Lt);
pBufferSize = max(bufferSize_M, max(pBufferSize_L, pBufferSize_Lt));

// pBuffer returned by cudaMalloc is automatically aligned to 128 bytes.
cudaMalloc((void**)&pBuffer, pBufferSize);

// step 4: perform analysis of incomplete Cholesky on M
// perform analysis of triangular solve on L
// The lower triangular part of M has the same sparsity pattern as L, so
// we can do analysis of csric02 and csrsv2 simultaneously.
cusparseDcsric02_analysis(handle, m, nnz, descr_M,
   d_csrVal, d_csrRowPtr, d_csrColInd, info_M,
   policy_M, pBuffer);
status = cusparseXcsric02_zeroPivot(handle, info_M, &structural_zero);
if (CUSPARSE_STATUS_ZERO_PIVOT == status){
   printf("A(%d,%d) is missing\n", structural_zero, structural_zero);
}

cusparseDcsrsv2_analysis(handle, trans_L, m, nnz, &alpha, descr_L,
   d_csrVal, d_csrRowPtr, d_csrColInd, info_L, policy_L, pBuffer);

cusparseDcsrsv2_analysis(handle, trans_Lt, m, nnz, &alpha, descr_L,
   d_csrVal, d_csrRowPtr, d_csrColInd, info_Lt, policy_Lt, pBuffer);

// step 6: free resources
cudaFree(pBuffer);
cusparseDestroyMatDescr(descr_M);
cusparseDestroyMatDescr(descr_L);
cusparseDestroyCsric02Info(info_M);
cusparseDestroyCsrsv2Info(info_L);
cusparseDestroyCsrsv2Info(info_Lt);
cusparseDestroy(handle);
## Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows and columns of matrix $A$.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzeros of matrix $A$.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix $A$. 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>.</td>
</tr>
<tr>
<td>csrValA_valM</td>
<td>$&lt;\text{type}&gt;$ array of $\text{nnz} ( = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0))$ nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of $m + 1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of $\text{nnz} ( = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0))$ column indices of the nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>info</td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
<tr>
<td>policy</td>
<td>the supported policies are <code>CUSPARSE_SOLVE_POLICY_NO_LEVEL</code> and <code>CUSPARSE_SOLVE_POLICY_USE_LEVEL</code>.</td>
</tr>
<tr>
<td>pBuffer</td>
<td>buffer allocated by the user; the size is returned by <code>csric02_bufferSize()</code>.</td>
</tr>
</tbody>
</table>

## Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrValA_valM</td>
<td>$&lt;\text{type}&gt;$ matrix containing the incomplete-Cholesky lower triangular factor.</td>
</tr>
</tbody>
</table>

## Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed ($m, \text{nnz} \leq 0$), base index is not 0 or 1.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.1.5. `cusparseXcsric02_zeroPivot()`

cusparseStatus_t
cusparseXcsric02_zeroPivot(cusparseHandle_t handle,
    csric02Info_t info,
    int *position);

If the returned error code is `CUSPARSE_STATUS_ZERO_PIVOT`, `position=j` means
$A(j,j)$ has either a structural zero or a numerical zero; otherwise, `position=-1`.

The `position` can be 0-based or 1-based, the same as the matrix.

Function `cusparseXcsric02_zeroPivot()` is a blocking call. It calls
`cudaDeviceSynchronize()` to make sure all previous kernels are done.

The `position` can be in the host memory or device memory. The user can set proper
mode with `cusparseSetPointerMode()`.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>info contains structural zero or numerical zero if the user already called <code>csric02_analysis()</code> or <code>csric02()</code>.</td>
</tr>
</tbody>
</table>

**Output**

| position | if no structural or numerical zero, `position` is -1; otherwise, if $A(j,j)$ is missing or $L(j,j)$ is zero, `position=j`. |

**Status Returned**

<table>
<thead>
<tr>
<th><code>CUSPARSE_STATUS_SUCCESS</code></th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>info is not valid.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
10.1.6. cusparse<t>bsric02_bufferSize()

cusparseStatus_t
cusparseSbsric02_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
float *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsric02Info_t info,
int *pBufferSizeInBytes);

cusparseStatus_t
cusparseDbsric02_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
double *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsric02Info_t info,
int *pBufferSizeInBytes);

cusparseStatus_t
cusparseCbsric02_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
cuComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsric02Info_t info,
int *pBufferSizeInBytes);

cusparseStatus_t
cusparseZbsric02_bufferSize(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
cuDoubComplex *bsrVaIA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsric02Info_t info,
int *pBufferSizeInBytes);

This function returns the size of a buffer used in computing the incomplete-Cholesky factorization with 0 fill-in and no pivoting.
\[ A \approx LL^H \]

\( A \) is an \((mb \times \text{blockDim}) \times (mb \times \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \( bsrValA, bsrRowPtrA, \) and \( bsrColIndA \).

The buffer size depends on the dimensions of \( mb, \text{blockDim}, \) and the number of nonzero blocks of the matrix \( nnzb \). If the user changes the matrix, it is necessary to call \( \text{bsric02_bufferSize()} \) again to have the correct buffer size; otherwise, a segmentation fault may occur.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>dirA</td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td>mb</td>
<td>number of block rows and block columns of matrix ( A ).</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>bsrValA</td>
<td>(&lt;\text{type}&gt;) array of ( nnzb ) nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>integer array of ( mb + 1 ) elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td>bsrColIndA</td>
<td>integer array of ( nnzb ) column indices of the nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>blockDim</td>
<td>block dimension of sparse matrix ( A ), larger than zero.</td>
</tr>
</tbody>
</table>

**Output**

| info | record internal states based on different algorithms. |
| pBufferSizeInBytes | number of bytes of the buffer used in \( \text{bsric02_analysis()} \) and \( \text{bsric02()} \). |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (( mb, nnzb \leq 0 )); the base index is not 0 or 1.</td>
</tr>
<tr>
<td>cuSPARSE Status Code</td>
<td>Description</td>
</tr>
<tr>
<td>----------------------------------------</td>
<td>------------------------------------------------------------------</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.1.7. `cusparse<t>bsric02_analysis()`

```c
cusparseStatus_t
cusparseSbsric02_analysis(cusparseHandle_t handle,  
cusparseDirection_t dirA,  
int mb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const float *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int blockDim,  
bsric02Info_t info,  
cusparseSolvePolicy_t policy,  
void *pBuffer);
```

```c
cusparseStatus_t
cusparseDbsric02_analysis(cusparseHandle_t handle,  
cusparseDirection_t dirA,  
int mb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const double *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int blockDim,  
bsric02Info_t info,  
cusparseSolvePolicy_t policy,  
void *pBuffer);
```

```c
cusparseStatus_t
cusparseCbsric02_analysis(cusparseHandle_t handle,  
cusparseDirection_t dirA,  
int mb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const cuComplex *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int blockDim,  
bsric02Info_t info,  
cusparseSolvePolicy_t policy,  
void *pBuffer);
```

```c
cusparseStatus_t
cusparseZbsric02_analysis(cusparseHandle_t handle,  
cusparseDirection_t dirA,  
int mb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const cuDoubleComplex *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int blockDim,  
bsric02Info_t info,  
cusparseSolvePolicy_t policy,  
void *pBuffer);
```
This function performs the analysis phase of the incomplete-Cholesky factorization with 0 fill-in and no pivoting

\[ A \approx LL^H \]

\( A \) is an \((mb*blockDim) \times (mb*blockDim)\) sparse matrix that is defined in BSR storage format by the three arrays \( bsrValA, bsrRowPtrA \), and \( bsrColIndA \). The block in BSR format is of size \( blockDim*blockDim \), stored as column-major or row-major as determined by parameter \( dirA \), which is either \texttt{CUSPARSE_DIRECTION_COLUMN} or \texttt{CUSPARSE_DIRECTION_ROW}. The matrix type must be \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, and the fill mode and diagonal type are ignored.

This function requires a buffer size returned by \texttt{bsric02_bufferSize90}. The address of \texttt{pBuffer} must be a multiple of 128 bytes. If it is not, \texttt{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Function \texttt{bsric02_analysis()} reports structural zero and computes level information stored in the opaque structure \texttt{info}. The level information can extract more parallelism during incomplete Cholesky factorization. However \texttt{bsric02()} can be done without level information. To disable level information, the user needs to specify the parameter \texttt{policy} of \texttt{bsric02[_analysis]} as \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}.

Function \texttt{bsric02_analysis()} always reports the first structural zero, even when parameter \texttt{policy} is \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}. The user must call \texttt{cusparseXbsric02_zeroPivot()} to know where the structural zero is.

It is the user’s choice whether to call \texttt{bsric02()} if \texttt{bsric02_analysis()} reports a structural zero. In this case, the user can still call \texttt{bsric02()}, which returns a numerical zero in the same position as the structural zero. However the result is meaningless.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>dirA</td>
<td>storage format of blocks, either \texttt{CUSPARSE_DIRECTION_ROW} or \texttt{CUSPARSE_DIRECTION_COLUMN}.</td>
</tr>
<tr>
<td>mb</td>
<td>number of block rows and block columns of matrix ( A ).</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}. Also, the supported index bases are \texttt{CUSPARSE_INDEX_BASE_ZERO} and \texttt{CUSPARSE_INDEX_BASE_ONE}.</td>
</tr>
<tr>
<td>bsrValA</td>
<td>(&lt;\text{type}&gt;) array of ( nnzb = bsrRowPtrA(mb) - bsrRowPtrA(0) ) nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>integer array of ( mb + 1 ) elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td>bsrColIndA</td>
<td>integer array of ( nnzb = bsrRowPtrA(mb) - bsrRowPtrA(0) ) column indices of the nonzero blocks of matrix ( A ).</td>
</tr>
</tbody>
</table>
### blockDim
- **block dimension of sparse matrix A; must be larger than zero.**

### info
- **structure initialized using cusparseCreateBsrc02Info().**

### policy
- **the supported policies are CUSPARSE_SOLVE_POLICY_NO_LEVEL and CUSPARSE_SOLVE_POLICY_USE_LEVEL.**

### pBuffer
- **buffer allocated by the user; the size is returned by bsric02_bufferSize().**

### Output
- **info**
  - **structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged).**

### Status Returned
- **CUSPARSE_STATUS_SUCCESS**
  - the operation completed successfully.
- **CUSPARSE_STATUS_NOT_INITIALIZED**
  - the library was not initialized.
- **CUSPARSE_STATUS_ALLOC_FAILED**
  - the resources could not be allocated.
- **CUSPARSE_STATUS_INVALID_VALUE**
  - invalid parameters were passed (`mb, nnzb <= 0`); the base index is not 0 or 1.
- **CUSPARSE_STATUS_ARCH_MISMATCH**
  - the device only supports compute capability 2.0 and above.
- **CUSPARSE_STATUS_INTERNAL_ERROR**
  - an internal operation failed.
- **CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED**
  - the matrix type is not supported.
10.1.8. `cusparse<t>bsric02()`

cusparseStatus_t

`cusparseSbsric02(cusparseHandle_t handle,
    cusparseDirection_t dirA,
    int mb,
    int nnzb,
    const cusparseMatDescr_t descrA,
    float *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    bsric02Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer);`

cusparseStatus_t

`cusparseDbsric02(cusparseHandle_t handle,
    cusparseDirection_t dirA,
    int mb,
    int nnzb,
    const cusparseMatDescr_t descrA,
    double *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    bsric02Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer);`

cusparseStatus_t

`cusparseCbsric02(cusparseHandle_t handle,
    cusparseDirection_t dirA,
    int mb,
    int nnzb,
    const cusparseMatDescr_t descrA,
    cuComplex *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    bsric02Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer);`

cusparseStatus_t

`cusparseZbsric02(cusparseHandle_t handle,
    cusparseDirection_t dirA,
    int mb,
    int nnzb,
    const cusparseMatDescr_t descrA,
    cuDoubleComplex *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    bsric02Info_t info,
    cusparseSolvePolicy_t policy,
    void *pBuffer);`
This function performs the solve phase of the incomplete-Cholesky factorization with 0 fill-in and no pivoting

\[ A \approx LL^H \]

\( A \) is an \((mb*\text{blockDim}) \times (mb*\text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \( \text{bsrValA}, \text{bsrRowPtrA}, \) and \( \text{bsrColIndA} \).

The block in BSR format is of size \( \text{blockDim} \times \text{blockDim} \), stored as column-major or row-major as determined by parameter \( \text{dirA} \), which is either \text{CUSPARSE_DIRECTION_COLUMN} or \text{CUSPARSE_DIRECTION_ROW}.

The matrix type must be \text{CUSPARSE_MATRIX_TYPE_GENERAL}, and the fill mode and diagonal type are ignored.

This function requires a buffer size returned by \text{bsric02_bufferSize()}. The address of \( \text{pBuffer} \) must be a multiple of 128 bytes. If it is not, \text{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Although \text{bsric02()} can be done without level information, the user must be aware of consistency. If \text{bsric02_analysis()} is called with policy \text{CUSPARSE_SOLVE_POLICY_USE_LEVEL}, \text{bsric02()} can be run with or without levels. On the other hand, if \text{bsric02_analysis()} is called with \text{CUSPARSE_SOLVE_POLICY_NO_LEVEL}, \text{bsric02()} can only accept \text{CUSPARSE_SOLVE_POLICY_NO_LEVEL}; otherwise, \text{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Function \text{bsric02()} has the same behavior as \text{csric02()}.

Function \text{bsric02()} reports the first numerical zero, including a structural zero. The user must call \text{cusparseXbsric02_zeroPivot()} to know where the numerical zero is.

For example, suppose \( A \) is a real m-by-m matrix, where \( m=mb*\text{blockDim} \).

The following code solves precondition system \( M*y = x \), where \( M \) is the product of Cholesky factorization \( L \) and its transpose.
// Suppose that A is m x m sparse matrix represented by BSR format,
// The number of block rows/columns is mb, and
// the number of nonzero blocks is nnzb.
// Assumption:
// - handle is already created by cusparseCreate(),
// - (d_bsrRowPtr, d_bsrColInd, d_bsrVal) is BSR of A on device memory,
// - d_x is right hand side vector on device memory,
// - d_y is solution vector on device memory.
// - d_z is intermediate result on device memory.
// - d_x, d_y and d_z are of size m.
cusparseMatDescr_t descr_M = 0;
cusparseMatDescr_t descr_L = 0;
bsric02Info_t info_M  = 0;
bsrsv2Info_t  info_L  = 0;
bsrsv2Info_t  info_Lt = 0;
int pBufferSize_M;
int pBufferSize_L;
int pBufferSize_Lt;
int pBufferSize;
void *pBuffer = 0;
int structural_zero;
int numerical_zero;
const double alpha = 1.;
const cusparseSolvePolicy_t policy_M  = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const cusparseSolvePolicy_t policy_L  = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const cusparseSolvePolicy_t policy_Lt = CUSPARSE_SOLVE_POLICY_USE_LEVEL;
const cusparseOperation_t trans_L  = CUSPARSE_OPERATION_NON_TRANSPOSE;
const cusparseOperation_t trans_Lt = CUSPARSE_OPERATION_TRANSPOSE;
const cusparseDirection_t dir = CUSPARSE_DIRECTION_COLUMN;

// step 1: create a descriptor which contains
// - matrix M is base-1
// - matrix L is base-1
// - matrix L is lower triangular
// - matrix L has non-unit diagonal
cusparseCreateMatDescr(&descr_M);
cusparseCreateMatDescr(&descr_L);
cusparseSetMatIndexBase(descr_M, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatIndexBase(descr_L, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descr_M, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatType(descr_L, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatFillMode(descr_L, CUSPARSE_FILL_MODE_LOWER);
cusparseSetMatDiagType(descr_L, CUSPARSE_DIAG_TYPE_NON_UNIT);

// step 2: create a empty info structure
// we need one info for bsric02 and two info's for bsrsv2
cusparseCreateBsric02Info(&info_M);
cusparseCreateBsrsv2Info(&info_L);
cusparseCreateBsrsv2Info(&info_Lt);

// step 3: query how much memory used in bsric02 and bsrsv2, and allocate the
// buffer
cusparseBbsric02_bufferSize(handle, dir, mb, nnzb,
  descr_M, d_bsrVal, d_bsrRowPtr, d_bsrColInd, blockDim, info_M,
  &bufferSize_M);
cusparseBbsrsv2_bufferSize(handle, dir, trans_L, mb, nnzb,
  descr_L, d_bsrVal, d_bsrRowPtr, d_bsrColInd, blockDim, info_L,
  &bufferSize_L);
cusparseBbsrsv2_bufferSize(handle, dir, trans_Lt, mb, nnzb,
  descr_L, d_bsrVal, d_bsrRowPtr, d_bsrColInd, blockDim, info_Lt,
  &bufferSize_Lt);
pBufferSize = max(bufferSize_M, max(pBufferSize_L, pBufferSize_Lt));

// pBuffer returned by cudaMalloc is automatically aligned to 128 bytes.
cudaMalloc((void**)&pBuffer, pBufferSize);

// step 4: perform analysis of incomplete Cholesky on M
// perform analysis of triangular solve on L
// perform analysis of triangular solve on L'
// The lower triangular part of M has the same sparsity pattern as L'

M = LL^H
Input

| **handle** | handle to the cuSPARSE library context. |
| **dirA**   | storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN. |
| **mb**     | number of block rows and block columns of matrix A. |
| **nnzb**   | number of nonzero blocks of matrix A. |
| **descrA** | the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE. |
| **bsrValA**| <type> array of nnzb (= bsrRowPtrA(mb) - bsrRowPtrA(0)) nonzero blocks of matrix A. |
| **bsrRowPtrA** | integer array of mb + 1 elements that contains the start of every block row and the end of the last block row plus one. |
| **bsrColIndA** | integer array of nnzb (= bsrRowPtrA(mb) - bsrRowPtrA(0)) column indices of the nonzero blocks of matrix A. |
| **blockDim** | block dimension of sparse matrix A, larger than zero. |
| **info**   | structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged). |
| **policy** | the supported policies are CUSPARSE_SOLVE_POLICY_NO_LEVEL and CUSPARSE_SOLVE_POLICY_USE_LEVEL. |
| **pBuffer** | buffer allocated by the user, the size is returned by bsric02_bufferSize(). |

Output

| **bsrValA** | <type> matrix containing the incomplete-Cholesky lower triangular factor. |

Status Returned

| **CUSPARSE_STATUS_SUCCESS** | the operation completed successfully. |
| **CUSPARSE_STATUS_NOT_INITIALIZED** | the library was not initialized. |
| **CUSPARSE_STATUS_ALLOC_FAILED** | the resources could not be allocated. |
| **CUSPARSE_STATUS_INVALID_VALUE** | invalid parameters were passed (mb, nnzb<=0); the base index is not 0 or 1. |
| **CUSPARSE_STATUS_ARCH_MISMATCH** | the device only supports compute capability 2.0 and above. |
| **CUSPARSE_STATUS_INTERNAL_ERROR** | an internal operation failed. |
CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported.

### 10.1.9. cusparseXbsric02_zeroPivot()

```c
cusparseStatus_t
cusparseXbsric02_zeroPivot(cusparseHandle_t handle,
                           bsric02Info_t info,
                           int *position);
```

If the returned error code is CUSPARSE_STATUS_ZERO_PIVOT, position=\(j\) means \(A(j,j)\) has either a structural zero or a numerical zero (the block is not positive definite). Otherwise position=\(-1\).

The position can be 0-based or 1-based, the same as the matrix.

Function `cusparseXbsric02_zeroPivot()` is a blocking call. It calls `cudaDeviceSynchronize()` to make sure all previous kernels are done.

The position can be in the host memory or device memory. The user can set the proper mode with `cusparseSetPointerMode()`.

**Input**

| handle | handle to the cuSPARSE library context. |
| info   | info contains a structural zero or a numerical zero if the user already called bsric02_analysis() or bsric02(). |

**Output**

| position | if no structural or numerical zero, position is \(-1\), otherwise if \(A(j,j)\) is missing or \(L(j,j)\) is not positive definite, position=\(j\). |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully. |
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized. |
| CUSPARSE_STATUS_INVALID_VALUE | info is not valid. |
| CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed. |

### 10.2. Incomplete LU Factorization: level 0

Different algorithms for ilu0 are discussed in this section.
10.2.1. cusparse<t>csrilu0()

This function computes the incomplete-LU factorization with 0 fill-in and no pivoting:

\[ op(A) = LU \]

\( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValM} \), \( \text{csrRowPtrA} \), and \( \text{csrColIndA} \), and

\[
op(A) = \begin{cases} 
A & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_NON\_TRANSPOSE} \\
A^T & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_TRANSPOSE} \\
A^H & \text{if } \text{trans} == \text{CUSPARSE\_OPERATION\_CONJUGATE\_TRANSPOSE} 
\end{cases}
\]

Notice that the diagonal of lower triangular factor \( L \) is unitary and need not be stored. Therefore, the input matrix is overwritten with the resulting lower and upper triangular factors \( L \) and \( U \), respectively.

A call to this routine must be preceded by a call to the \texttt{csrsv\_analysis()} routine.

The matrix descriptor for \texttt{csrsv\_analysis()} and \texttt{csrilu0()} must be the same. Otherwise, runtime error would occur.
This function requires some extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>trans</code></td>
<td>the operation <code>op(A)</code>.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>number of rows and columns of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix <code>A</code>. 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>.</td>
</tr>
<tr>
<td><code>csrValM</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>csrRowPtrA</code></td>
<td>integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><code>csrColIndA</code></td>
<td>integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> column indices of the nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>info</code></td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>csrValM</code></td>
<td><code>&lt;type&gt;</code> matrix containing the incomplete-LU lower and upper triangular factors.</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed (<code>m&lt;0</code>).</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_EXECUTION_FAILED</code></td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.2.2. cusparseCsrilu0Ex()

cusparseStatus_t CUSPARSEAPI cusparseCsrilu0Ex(cusparseHandle_t handle, cusparseOperation_t trans, int m, const cusparseMatDescr_t descrA, void *csrSortedValA_ValM, cudaDataType csrSortedValA_ValMtype, const int *csrSortedRowPtrA, const int *csrSortedColIndA, cusparseSolveAnalysisInfo_t info, cudaDataType executiontype);

This function is an extended version of cusparse<t>csrilu0(). For detailed description of the functionality, see cusparse<t>csrilu0().

This function does not support half-precision execution type, but it supports half-precision IO with single precision execution.

**Input specifically required by cusparseCsrilu0Ex**

<table>
<thead>
<tr>
<th>csrSortedValA_ValMtype</th>
<th>Data type of csrSortedValA_ValM.</th>
</tr>
</thead>
<tbody>
<tr>
<td>executiontype</td>
<td>Data type used for computation.</td>
</tr>
</tbody>
</table>
10.2.3. cusparse\textless t\textgreater csrilu02_numericBoost()

The user can use a boost value to replace a numerical value in incomplete LU factorization. The \texttt{tol} is used to determine a numerical zero, and the \texttt{boost\_val} is used to replace a numerical zero. The behavior is

if \( \texttt{tol} \geq \texttt{fabs(A(j,j))} \), then \( \texttt{A(j,j)}=\texttt{boost\_val} \).

To enable a boost value, the user has to set parameter \texttt{enable\_boost} to \texttt{1} before calling \texttt{csrilu02()}. To disable a boost value, the user can call \texttt{csrilu02\_numericBoost()} again with parameter \texttt{enable\_boost=0}.

If \texttt{enable\_boost=0}, \texttt{tol} and \texttt{boost\_val} are ignored.

Both \texttt{tol} and \texttt{boost\_val} can be in the host memory or device memory. The user can set the proper mode with \texttt{cusparseSetPointerMode()}.

\textbf{Input}

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>structure initialized using \texttt{cusparseCreateCsrilu02Info()}</td>
</tr>
<tr>
<td>enable_boost</td>
<td>disable boost by \texttt{enable_boost=0}; otherwise, boost is enabled.</td>
</tr>
<tr>
<td>tol</td>
<td>tolerance to determine a numerical zero.</td>
</tr>
<tr>
<td>boost_val</td>
<td>boost value to replace a numerical zero.</td>
</tr>
</tbody>
</table>
Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>info or pointer mode is not valid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>

10.2.4. cusparse<t>csrilu02_bufferSize()

cusparseStatus_t
cusparseScsrilu02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csrilu02Info_t info,
    int *pBufferSizeInBytes);

cusparseStatus_t
cusparseDcsrilu02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csrilu02Info_t info,
    int *pBufferSizeInBytes);

cusparseStatus_t
cusparseCcsrilu02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csrilu02Info_t info,
    int *pBufferSizeInBytes);

cusparseStatus_t
cusparseZcsrilu02_bufferSize(cusparseHandle_t handle,
    int m,
    int nnz,
    const cusparseMatDescr_t descrA,
    cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    csrilu02Info_t info,
    int *pBufferSizeInBytes);

This function returns size of the buffer used in computing the incomplete-LU factorization with 0 fill-in and no pivoting:
A is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays 
\( \text{csrValA} \), \( \text{csrRowPtrA} \), and \( \text{csrColIndA} \).

The buffer size depends on the dimension \( m \) and \( \text{nnz} \), the number of nonzeros of the matrix. If the user changes the matrix, it is necessary to call \text{csrilu02_bufferSize()} \ again to have the correct buffer size; otherwise, a segmentation fault may occur.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>( m )</td>
<td>number of rows and columns of matrix ( A ).</td>
</tr>
<tr>
<td>( \text{nnz} )</td>
<td>number of nonzeros of matrix ( A ).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is \text{CUSPARSE_MATRIX_TYPE_GENERAL}. Also, the supported index bases are \text{CUSPARSE_INDEX_BASE_ZERO} and \text{CUSPARSE_INDEX_BASE_ONE}.</td>
</tr>
<tr>
<td>( \text{csrValA} )</td>
<td>&lt;type&gt; array of ( \text{nnz} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0) ) nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>( \text{csrRowPtrA} )</td>
<td>integer array of ( m + 1 ) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>( \text{csrColIndA} )</td>
<td>integer array of ( \text{nnz} = \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0) ) column indices of the nonzero elements of matrix ( A ).</td>
</tr>
</tbody>
</table>

**Output**

| info | record internal states based on different algorithms. |
| \( \text{ pBufferSizeInBytes} \) | number of bytes of the buffer used in \text{csrilu02_analysis()} and \text{csrilu02()}. |

**Status Returned**

| \text{CUSPARSE_STATUS_SUCCESS} | the operation completed successfully. |
| \text{CUSPARSE_STATUS_NOT_INITIALIZED} | the library was not initialized. |
| \text{CUSPARSE_STATUS_ALLOC_FAILED} | the resources could not be allocated. |
| \text{CUSPARSE_STATUS_INVALID_VALUE} | invalid parameters were passed (\( m, \text{nnz}<0 \)), base index is not 0 or 1. |
| \text{CUSPARSE_STATUS_ARCH_MISMATCH} | the device only supports compute capability 2.0 and above. |
| \text{CUSPARSE_STATUS_INTERNAL_ERROR} | an internal operation failed. |
| \text{CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED} | the matrix type is not supported. |
10.2.5. cusparse<t>csrilu02_analysis()

```c
cusparseStatus_t
cusparseScsrilu02_analysis(cusparseHandle_t handle,
   int m,
   int nnz,
   const cusparseMatDescr_t descrA,
   const float *csrValA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   csrilu02Info_t info,
   cusparseSolvePolicy_t policy,
   void *pBuffer);

cusparseStatus_t
cusparseDcsrilu02_analysis(cusparseHandle_t handle,
   int m,
   int nnz,
   const cusparseMatDescr_t descrA,
   const double *csrValA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   csrilu02Info_t info,
   cusparseSolvePolicy_t policy,
   void *pBuffer);

cusparseStatus_t
cusparseCcsrilu02_analysis(cusparseHandle_t handle,
   int m,
   int nnz,
   const cusparseMatDescr_t descrA,
   const cuComplex *csrValA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   csrilu02Info_t info,
   cusparseSolvePolicy_t policy,
   void *pBuffer);

cusparseStatus_t
cusparseZcsrilu02_analysis(cusparseHandle_t handle,
   int m,
   int nnz,
   const cusparseMatDescr_t descrA,
   const cuDoubleComplex *csrValA,
   const int *csrRowPtrA,
   const int *csrColIndA,
   csrilu02Info_t info,
   cusparseSolvePolicy_t policy,
   void *pBuffer);
```

This function performs the analysis phase of the incomplete-LU factorization with 0 fill-in and no pivoting:

\[ A \approx LU \]

\( A \) is an \( m \times m \) sparse matrix that is defined in CSR storage format by the three arrays \( \text{csrValA} \), \( \text{csrRowPtrA} \), and \( \text{csrColIndA} \).
This function requires the buffer size returned by `csrilu02_bufferSize()`. The address of `pBuffer` must be a multiple of 128 bytes. If not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

Function `csrilu02_analysis()` reports a structural zero and computes level information stored in the opaque structure `info`. The level information can extract more parallelism during incomplete LU factorization; however `csrilu02()` can be done without level information. To disable level information, the user must specify the policy of `csrilu02()` as `CUSPARSE_SOLVE_POLICY_NO_LEVEL`.

It is the user’s choice whether to call `csrilu02()` if `csrilu02_analysis()` reports a structural zero. In this case, the user can still call `csrilu02()`, which will return a numerical zero at the same position as the structural zero. However the result is meaningless.

**Input**

| handle | handle to the cuSPARSE library context. |
| m | number of rows and columns of matrix A. |
| nnz | number of nonzeros of matrix A. |
| descrA | the descriptor of matrix A. The supported matrix type is `CUSPARSE_MATRIX_TYPE_GENERAL`. Also, the supported index bases are `CUSPARSE_INDEX_BASE_ZERO` and `CUSPARSE_INDEX_BASE_ONE`. |
| csrValA | `<type> array of nnz (= csrRowPtrA(m) - csrRowPtrA(0))` nonzero elements of matrix A. |
| csrRowPtrA | integer array of `m + 1` elements that contains the start of every row and the end of the last row plus one. |
| csrColIndA | integer array of `nnz (= csrRowPtrA(m) - csrRowPtrA(0))` column indices of the nonzero elements of matrix A. |
| info | structure initialized using `cusparseCreateCsrilu02Info()` |
| policy | the supported policies are `CUSPARSE_SOLVE_POLICY_NO_LEVEL` and `CUSPARSE_SOLVE_POLICY_USE_LEVEL`. |
| pBuffer | buffer allocated by the user, the size is returned by `csrilu02_bufferSize()` |

**Output**

| info | structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged). |

**Status Returned**

<p>| <code>CUSPARSE_STATUS_SUCCESS</code> | the operation completed successfully. |
| <code>CUSPARSE_STATUS_NOT_INITIALIZED</code> | the library was not initialized. |</p>
<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m,nnz&lt;0), base index is not 0 or 1.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.2.6. cusparse<t>csrilu02()

\[
\text{cusparseStatus_t}
\]

\[
\text{cusparseScsrilu02(cusparseHandle_t handle, int m, int nnz, const cusparseMatDescr_t descrA, float *csrValA_valM, const int *csrRowPtrA, const int *csrColIndA, csrilu02Info_t info, cusparseSolvePolicy_t policy, void *pBuffer);}
\]

\[
\text{cusparseStatus_t}
\]

\[
\text{cusparseDcsrilu02(cusparseHandle_t handle, int m, int nnz, const cusparseMatDescr_t descrA, double *csrValA_valM, const int *csrRowPtrA, const int *csrColIndA, csrilu02Info_t info, cusparseSolvePolicy_t policy, void *pBuffer);}
\]

\[
\text{cusparseStatus_t}
\]

\[
\text{cusparseCcsrilu02(cusparseHandle_t handle, int m, int nnz, const cusparseMatDescr_t descrA, cuComplex *csrValA_valM, const int *csrRowPtrA, const int *csrColIndA, csrilu02Info_t info, cusparseSolvePolicy_t policy, void *pBuffer);}
\]

\[
\text{cusparseStatus_t}
\]

\[
\text{cusparseZcsrilu02(cusparseHandle_t handle, int m, int nnz, const cusparseMatDescr_t descrA, cuDoubleComplex *csrValA_valM, const int *csrRowPtrA, const int *csrColIndA, csrilu02Info_t info, cusparseSolvePolicy_t policy, void *pBuffer);}
\]

This function performs the solve phase of the incomplete-LU factorization with 0 fill-in and no pivoting:

\[
A \approx LU
\]

\(A\) is an \(m \times m\) sparse matrix that is defined in CSR storage format by the three arrays \(\text{csrValA_valM}, \text{csrRowPtrA},\) and \(\text{csrColIndA}\).
This function requires a buffer size returned by `csrilu02_bufferSize()`. The address of `pBuffer` must be a multiple of 128 bytes. If not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

The matrix type must be `CUSPARSE_MATRIX_TYPE_GENERAL`. The fill mode and diagonal type are ignored.

Although `csrilu02()` can be done without level information, the user still needs to be aware of consistency. If `csrilu02_analysis()` is called with policy `CUSPARSE_SOLVE_POLICY_USE_LEVEL`, `csrilu02()` can be run with or without levels. On the other hand, if `csrilu02_analysis()` is called with `CUSPARSE_SOLVE_POLICY_NO_LEVEL`, `csrilu02()` can only accept `CUSPARSE_SOLVE_POLICY_NO_LEVEL`; otherwise, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

Function `csrilu02()` reports the first numerical zero, including a structural zero. The user must call `cusparseXcsrilu02_zeroPivot()` to know where the numerical zero is.
For example, suppose $A$ is a real $m \times m$ matrix, the following code solves precondition system $M^*y = x$ where $M$ is the product of LU factors $L$ and $U$.

```c
// Suppose that A is m x m sparse matrix represented by CSR format,
// Assumption:
// - handle is already created by cusparseCreate(),
// - (d_csrRowPtr, d_csrColInd, d_csrVal) is CSR of A on device memory,
// - d_x is right hand side vector on device memory,
// - d_y is solution vector on device memory.
// - d_z is intermediate result on device memory.

cusparseMatDescr_t descr_M = 0;
cusparseMatDescr_t descr_L = 0;
cusparseMatDescr_t descr_U = 0;
csrilu02Info_t info_M = 0;
csrsv2Info_t info_L = 0;
csrsv2Info_t info_U = 0;
int pBufferSize_M;
int pBufferSize_L;
int pBufferSize_U;
int pBufferSize;
void *pBuffer = 0;
int structural_zero;
int numerical_zero;
const double alpha = 1.;
const cusparseSolvePolicy_t policy_M = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const cusparseSolvePolicy_t policy_L = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const cusparseSolvePolicy_t policy_U = CUSPARSE_SOLVE_POLICY_USE_LEVEL;
const cusparseOperation_t trans_L = CUSPARSE_OPERATION_NON_TRANSPOSE;
const cusparseOperation_t trans_U = CUSPARSE_OPERATION_NON_TRANSPOSE;

cusparseCreateMatDescr(&descr_M);
cusparseSetMatIndexBase(descr_M, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descr_M, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseCreateMatDescr(&descr_L);
cusparseSetMatIndexBase(descr_L, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descr_L, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatFillMode(descr_L, CUSPARSE_FILL_MODE_LOWER);
cusparseSetMatDiagType(descr_L, CUSPARSE_DIAG_TYPE_UNIT);
cusparseCreateMatDescr(&descr_U);
cusparseSetMatIndexBase(descr_U, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descr_U, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatFillMode(descr_U, CUSPARSE_FILL_MODE_UPPER);
cusparseSetMatDiagType(descr_U, CUSPARSE_DIAG_TYPE_NON_UNIT);

cusparseCreateCsrilu02Info(&info_M);
cusparseCreateCsrsv2Info(&info_L);
cusparseCreateCsrsv2Info(&info_U);

cusparseDcsrilu02_bufferSize(handle, m, nnz,
    descr_M, d_csrVal, d_csrRowPtr, d_csrColInd, info_M, &pBufferSize_M);
cusparseDcsrsv2_bufferSize(handle, trans_L, m, nnz,
    descr_L, d_csrVal, d_csrRowPtr, d_csrColInd, info_L, &pBufferSize_L);
cusparseDcsrsv2_bufferSize(handle, trans_U, m, nnz,
    descr_U, d_csrVal, d_csrRowPtr, d_csrColInd, info_U, &pBufferSize_U);
pBufferSize = max(pBufferSize_M, max(pBufferSize_L, pBufferSize_U));

cusparseDcsrilu02_analysis(handle, m, nnz, descr_M,
    d_csrVal, d_csrRowPtr, d_csrColInd, info_M, policy_M, pBuffer);

// step 4: perform analysis of incomplete Cholesky on M
//         perform analysis of triangular solve on L
//         perform analysis of triangular solve on U

// pBuffer returned by cudaMalloc is automatically aligned to 128 bytes.
cudaMalloc((void**)&pBuffer, pBufferSize);

cusparseDestroyMatDescr(descr_M);
cusparseDestroyMatDescr(descr_L);
cusparseDestroyMatDescr(descr_U);
cusparseDestroyCsrilu02Info(info_M);
cusparseDestroyCsrsv2Info(info_L);
cusparseDestroyCsrsv2Info(info_U);
cusparseDestroy(handle);
```
### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>number of rows and columns of matrix A.</td>
</tr>
<tr>
<td><code>nnz</code></td>
<td>number of nonzeros of matrix A.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix A. 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>.</td>
</tr>
<tr>
<td><code>csrValA_valM</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> nonzero elements of matrix A.</td>
</tr>
<tr>
<td><code>csrRowPtrA</code></td>
<td>integer array of <code>m + 1</code> elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><code>csrColIndA</code></td>
<td>integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td><code>info</code></td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
<tr>
<td><code>policy</code></td>
<td>the supported policies are <code>CUSPARSE_SOLVE_POLICY_NO_LEVEL</code> and <code>CUSPARSE_SOLVE_POLICY_USE_LEVEL</code>.</td>
</tr>
</tbody>
</table>
| `pBuffer`     | buffer allocated by the user; the size is returned by `csrilu02_bufferSize()`.

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>csrValA_valM</code></td>
<td><code>&lt;type&gt;</code> matrix containing the incomplete-LU lower and upper triangular factors.</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed (<code>m, nnz&lt;=0</code>), base index is not 0 or 1.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.2.7. cusparseXcsrilu02_zeroPivot()

cusparseStatus_t

cusparseXcsrilu02_zeroPivot(cusparseHandle_t handle,
csrilu02Info_t info,
int *position);

If the returned error code is CUSPARSE_STATUS_ZERO_PIVOT, position=j means A(j,j) has either a structural zero or a numerical zero; otherwise, position=-1.

The position can be 0-based or 1-based, the same as the matrix.

Function cusparseXcsrilu02_zeroPivot() is a blocking call. It calls cudaDeviceSynchronize() to make sure all previous kernels are done.

The position can be in the host memory or device memory. The user can set proper mode with cusparseSetPointerMode().

Input

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>info contains structural zero or numerical zero if the user already called csrilu02_analysis() or csrilu02().</td>
</tr>
</tbody>
</table>

Output

| position | if no structural or numerical zero, position is -1; otherwise if A(j,j) is missing or U(j,j) is zero, position=j. |

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>info is not valid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
10.2.8. cusparse<t>bssrilu02_numericBoost()  

```c
cusparseStatus_t cusparseSbssrilu02_numericBoost(cusparseHandle_t handle, 
                                                 bssrilu02Info_t info, 
                                                 int enable_boost, 
                                                 double *tol, 
                                                 float *boost_val);

cusparseStatus_t cusparseDbssrilu02_numericBoost(cusparseHandle_t handle, 
                                                 bssrilu02Info_t info, 
                                                 int enable_boost, 
                                                 double *tol, 
                                                 double *boost_val);

cusparseStatus_t cusparseCbssrilu02_numericBoost(cusparseHandle_t handle, 
                                                 bssrilu02Info_t info, 
                                                 int enable_boost, 
                                                 double *tol, 
                                                 cuComplex *boost_val);

cusparseStatus_t cusparseZbssrilu02_numericBoost(cusparseHandle_t handle, 
                                                 bssrilu02Info_t info, 
                                                 int enable_boost, 
                                                 double *tol, 
                                                 cuDoubleComplex *boost_val);
```

The user can use a boost value to replace a numerical value in incomplete LU factorization. Parameter `tol` is used to determine a numerical zero, and `boost_val` is used to replace a numerical zero. The behavior is as follows:

```c
if tol >= fabs(A(j,j)), then reset each diagonal element of block A(j,j) by boost_val.
```

To enable a boost value, the user sets parameter `enable_boost` to 1 before calling `bssrilu02()`. To disable the boost value, the user can call `bssrilu02_numericBoost()` with parameter `enable_boost=0`.

If `enable_boost=0`, `tol` and `boost_val` are ignored.

Both `tol` and `boost_val` can be in host memory or device memory. The user can set the proper mode with `cusparseSetPointerMode()`.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>info</code></td>
<td>structure initialized using <code>cusparseCreateBssrilu02Info()</code>.</td>
</tr>
<tr>
<td><code>enable_boost</code></td>
<td>disable boost by setting <code>enable_boost=0</code>. Otherwise, boost is enabled.</td>
</tr>
<tr>
<td><code>tol</code></td>
<td>tolerance to determine a numerical zero.</td>
</tr>
<tr>
<td>boost_val</td>
<td>boost value to replace a numerical zero.</td>
</tr>
<tr>
<td>-----------------</td>
<td>------------------------------------------</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>info or pointer mode is not valid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
10.2.9. `cusparse<t>bsrilu02_bufferSize()`

This function returns the size of the buffer used in computing the incomplete-LU factorization with 0 fill-in and no pivoting.
\( A \approx LU \)

\( A \) is an \((\text{mb} \times \text{blockDim}) \times (\text{mb} \times \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \( \text{bsrValA} \), \( \text{bsrRowPtrA} \), and \( \text{bsrColIndA} \).

The buffer size depends on the dimensions of \( \text{mb}, \text{blockDim}, \) and the number of nonzero blocks of the matrix \( \text{nnzb} \). If the user changes the matrix, it is necessary to call \( \text{bsrilu02_bufferSize()} \) again to have the correct buffer size; otherwise, a segmentation fault may occur.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>handle</strong></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><strong>dirA</strong></td>
<td>storage format of blocks, either ( \text{CUSPARSE_DIRECTION_ROW} ) or ( \text{CUSPARSE_DIRECTION_COLUMN} ).</td>
</tr>
<tr>
<td><strong>mb</strong></td>
<td>number of block rows and columns of matrix ( A ).</td>
</tr>
<tr>
<td><strong>nnzb</strong></td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><strong>descrA</strong></td>
<td>the descriptor of matrix ( A ). The supported matrix type is ( \text{CUSPARSE_MATRIX_TYPE_GENERAL} ). Also, the supported index bases are ( \text{CUSPARSE_INDEX_BASE_ZERO} ) and ( \text{CUSPARSE_INDEX_BASE_ONE} ).</td>
</tr>
<tr>
<td><strong>bsrValA</strong></td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnzb} = (\text{bsrRowPtrA(mb)} - \text{bsrRowPtrA(0)}) ) nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><strong>bsrRowPtrA</strong></td>
<td>integer array of ( \text{mb} + 1 ) elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td><strong>bsrColIndA</strong></td>
<td>integer array of ( \text{nnzb} = (\text{bsrRowPtrA(mb)} - \text{bsrRowPtrA(0)}) ) column indices of the nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><strong>blockDim</strong></td>
<td>block dimension of sparse matrix ( A ), larger than zero.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>info</strong></td>
<td>record internal states based on different algorithms.</td>
</tr>
<tr>
<td><strong>pBufferSizeInBytes</strong></td>
<td>number of bytes of the buffer used in ( \text{bsrilu02_analysis()} ) and ( \text{bsrilu02()} ).</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \text{CUSPARSE_STATUS_SUCCESS} )</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>( \text{CUSPARSE_STATUS_NOT_INITIALIZED} )</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>( \text{CUSPARSE_STATUS_ALLOC_FAILED} )</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>( \text{CUSPARSE_STATUS_INVALID_VALUE} )</td>
<td>invalid parameters were passed (( \text{mb}, \text{nnzb}&lt;=0 )), base index is not 0 or 1.</td>
</tr>
<tr>
<td>cuSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td>------------------------------------</td>
<td>----------------------------------------------------------</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.2.10. `cusparse<t>bsrilu02_analysis()`

cusparseStatus_t
cusparseSbsrilu02_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
float *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseDbsrilu02_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
double *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseCbsrilu02_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
cuComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseZbsrilu02_analysis(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrA,
cuDoubleComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);
This function performs the analysis phase of the incomplete-LU factorization with 0 fill-in and no pivoting

\[ A \approx LU \]

\( A \) is an \((\text{mb} \times \text{blockDim}) \times (\text{mb} \times \text{blockDim})\) sparse matrix that is defined in BSR storage format by the three arrays \( \text{bsrValA}, \text{bsrRowPtrA}, \) and \( \text{bsrColIndA} \). The block in BSR format is of size \( \text{blockDim} \times \text{blockDim} \), stored as column-major or row-major as determined by parameter \( \text{dirA} \), which is either \text{CUSPARSE_DIRECTION_COLUMN} or \text{CUSPARSE_DIRECTION_ROW}. The matrix type must be \text{CUSPARSE_MATRIX_TYPE_GENERAL}, and the fill mode and diagonal type are ignored.

This function requires a buffer size returned by \( \text{bsrilu02_bufferSize()} \). The address of \( \text{pBuffer} \) must be multiple of 128 bytes. If it is not, \text{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Function \( \text{bsrilu02_analysis()} \) reports a structural zero and computes level information stored in the opaque structure \( \text{info} \). The level information can extract more parallelism during incomplete LU factorization. However \( \text{bsrilu02()} \) can be done without level information. To disable level information, the user needs to specify the parameter \( \text{policy} \) of \( \text{bsrilu02[[_analysis| ]} \) as \text{CUSPARSE_SOLVE_POLICY_NO_LEVEL}.

Function \( \text{bsrilu02_analysis()} \) always reports the first structural zero, even with parameter \( \text{policy} \) is \text{CUSPARSE_SOLVE_POLICY_NO_LEVEL}. The user must call \( \text{cusparseXbsrilu02_zeroPivot()} \) to know where the structural zero is.

It is the user’s choice whether to call \( \text{bsrilu02()} \) if \( \text{bsrilu02_analysis()} \) reports a structural zero. In this case, the user can still call \( \text{bsrilu02()} \), which will return a numerical zero at the same position as the structural zero. However the result is meaningless.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>dirA</td>
<td>storage format of blocks, either \text{CUSPARSE_DIRECTION_ROW} or \text{CUSPARSE_DIRECTION_COLUMN}.</td>
</tr>
<tr>
<td>mb</td>
<td>number of block rows and block columns of matrix ( A ).</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix ( A ). The supported matrix type is \text{CUSPARSE_MATRIX_TYPE_GENERAL}. Also, the supported index bases are \text{CUSPARSE_INDEX_BASE_ZERO} and \text{CUSPARSE_INDEX_BASE_ONE}.</td>
</tr>
<tr>
<td>bsrValA</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnzb} = \text{bsrRowPtrA}(\text{mb}) - \text{bsrRowPtrA}(0) ) nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>integer array of ( \text{mb} + 1 ) elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
</tbody>
</table>
bsrColIndA | integer array of \( \text{nnzb} = \text{bsrRowPtrA}(mb) - \text{bsrRowPtrA}(0) \) column indices of the nonzero blocks of matrix \( A \).

blockDim | block dimension of sparse matrix \( A \), larger than zero.

info | structure initialized using `cusparseCreateBsrilu02Info()`. The `policy` is the supported policies are `CUSPARSE_SOLVE_POLICY_NO_LEVEL` and `CUSPARSE_SOLVE_POLICY_USE_LEVEL`.

pBuffer | buffer allocated by the user, the size is returned by `bsrilu02_bufferSize()`.

Output

info | structure filled with information collected during the analysis phase (that should be passed to the solve phase unchanged).

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed ((mb, \text{nnzb} &lt;= 0)); the base index is not 0 or 1.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.2.11. cusparse<t>bsrilu02()

cusparseStatus_t
cusparseSbsrilu02(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrY,
float *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseDbsrilu02(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrY,
double *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseCbsrilu02(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrY,
cuComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);

cusparseStatus_t
cusparseZbsrilu02(cusparseHandle_t handle,
cusparseDirection_t dirA,
int mb,
int nnzb,
const cusparseMatDescr_t descrY,
cuDoubleComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int blockDim,
bsrilu02Info_t info,
cusparseSolvePolicy_t policy,
void *pBuffer);
This function performs the solve phase of the incomplete-LU factorization with 0 fill-in and no pivoting

\[ A \approx LU \]

\( A \) is an \((mb \times blockDim) \times (mb \times blockDim)\) sparse matrix that is defined in BSR storage format by the three arrays \( bsrValA, bsrRowPtrA, \) and \( bsrColIndA \). The block in BSR format is of size \( blockDim \times blockDim \), stored as column-major or row-major determined by parameter \( dirA \), which is either \texttt{CUSPARSE_DIRECTION_COLUMN} or \texttt{CUSPARSE_DIRECTION_ROW}. The matrix type must be \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, and the fill mode and diagonal type are ignored. Function \texttt{bsrilu02()} supports an arbitrary \( blockDim \).

This function requires a buffer size returned by \texttt{bsrilu02_bufferSize()}.

The address of \( pBuffer \) must be a multiple of 128 bytes. If it is not, \texttt{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Although \texttt{bsrilu02()} can be used without level information, the user must be aware of consistency. If \texttt{bsrilu02_analysis()} is called with policy \texttt{CUSPARSE_SOLVE_POLICY_USE_LEVEL}, \texttt{bsrilu02()} can be run with or without levels. On the other hand, if \texttt{bsrilu02_analysis()} is called with \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}, \texttt{bsrilu02()} can only accept \texttt{CUSPARSE_SOLVE_POLICY_NO_LEVEL}; otherwise, \texttt{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Function \texttt{bsrilu02()} has the same behavior as \texttt{csrilu02()}. That is, 
\[ \text{bsr2csr(bsrilu02(A))} = \text{csrilu02(bsr2csr(A))} \]. The numerical zero of \texttt{csrilu02()} means there exists some zero \( U(j,j) \). The numerical zero of \texttt{bsrilu02()} means there exists some block \( U(j,j) \) that is not invertible.

Function \texttt{bsrilu02} reports the first numerical zero, including a structural zero. The user must call \texttt{cusparseXbsrilu02_zeroPivot()} to know where the numerical zero is.
For example, suppose $A$ is a real $m$-by-$m$ matrix where $m=mb*\text{blockDim}$. The following code solves precondition system $M*y = x$, where $M$ is the product of LU factors $L$ and $U$.  

```c
// Suppose that A is m x m sparse matrix represented by BSR format,
// the number of block rows/columns is mb, and
// the number of nonzero blocks is nnzb.
// Assumption:
// - handle is already created by cusparseCreate(),
// - (d_bsrRowPtr, d_bsrColInd, d_bsrVal) is BSR of A on device memory,
// - d_x is right hand side vector on device memory.
// - d_y is solution vector on device memory.
// - d_z is intermediate result on device memory.
// - d_x, d_y and d_z are of size m,

const double alpha = 1.0;
const cusparseDirection_t dir = CUSPARSE_DIRECTION_COLUMN;
const cusparseOperation_t trans_U = CUSPARSE_OPERATION_NON_TRANSPOSE;
const cusparseOperation_t trans_L = CUSPARSE_OPERATION_NON_TRANSPOSE;
const cusparseSolvePolicy_t policy_U = CUSPARSE_SOLVE_POLICY_USE_LEVEL;
const cusparseSolvePolicy_t policy_L = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const cusparseSolvePolicy_t policy_M = CUSPARSE_SOLVE_POLICY_NO_LEVEL;

int numerical_zero;
int structural_zero;

int pBufferSize;
int pBufferSize_U;
int pBufferSize_L;
int pBufferSize_M;
int pBufferSize_M;
int pBufferSize_L;
int pBufferSize_U;
void *pBuffer = 0;
```

### Assumption:

- Matrix $M$ has base-1
- Matrix $L$ has base-1
- Matrix $L$ is lower triangular
- Matrix $L$ has unit diagonal
- Matrix $U$ is base-1
- Matrix $U$ is upper triangular
- Matrix $U$ has non-unit diagonal

```c
cusparseSetMatDiagType(&descr_M, CUSPARSE_DIAG_TYPE_NON_UNIT);
cusparseSetMatFillMode(&descr_M, CUSPARSE_FILL_MODE_UPPER);
cusparseSetMatType(&descr_M, CUSPARSE_MATRIX_TYPE_GENERAL);
cusparseSetMatIndexBase(&descr_M, CUSPARSE_INDEX_BASE_ONE);
```

### Step 2:

- Create a empty info structure
- We need one info for bsrilu02 and two info's for bsrsv2

```c
cusparseCreateBsrilu02Info(&info_M);
cusparseCreateBsrsv2Info(&info_L);
cusparseCreateBsrsv2Info(&info_U);
```

### Step 3:

- Query how much memory used in bsrilu02 and bsrsv2, and allocate the buffer

```c
cusparseDbsrilu02_bufferSize(handle, dir, mb, nnzb, &pBufferSize_M);
cusparseDbsrsv2_bufferSize(handle, dir, trans_L, mb, nnzb, &pBufferSize_L);
cusparseDbsrsv2_bufferSize(handle, dir, trans_U, mb, nnzb, &pBufferSize_U);
```
**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>dirA</code></td>
<td>storage format of blocks: either <code>CUSPARSE_DIRECTION_ROW</code> or <code>CUSPARSE_DIRECTION_COLUMN</code>.</td>
</tr>
<tr>
<td><code>mb</code></td>
<td>number of block rows and block columns of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>nnzb</code></td>
<td>number of nonzero blocks of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix <code>A</code>. 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>.</td>
</tr>
<tr>
<td><code>bsrValA</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnzb</code> nonzero blocks of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>bsrRowPtrA</code></td>
<td>integer array of <code>mb + 1</code> elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td><code>bsrColIndA</code></td>
<td>integer array of <code>nnzb</code> column indices of the nonzero blocks of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>blockDim</code></td>
<td>block dimension of sparse matrix <code>A</code>; must be larger than zero.</td>
</tr>
<tr>
<td><code>info</code></td>
<td>structure with information collected during the analysis phase (that should have been passed to the solve phase unchanged).</td>
</tr>
<tr>
<td><code>policy</code></td>
<td>the supported policies are <code>CUSPARSE_SOLVE_POLICY_NO_LEVEL</code> and <code>CUSPARSE_SOLVE_POLICY_USE_LEVEL</code>.</td>
</tr>
<tr>
<td><code>pBuffer</code></td>
<td>buffer allocated by the user; the size is returned by <code>bsrilu02_bufferSize()</code>.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>bsrValA</code></td>
<td><code>&lt;type&gt;</code> matrix containing the incomplete-LU lower and upper triangular factors.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed (<code>mb,nnzb&lt;=0</code>); the base index is not 0 or 1.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device only supports compute capability 2.0 and above.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
10.2.12. cusparseXbsrilu02_zeroPivot()

cusparseStatus_t

cusparseXbsrilu02_zeroPivot(cusparseHandle_t handle,
                        bsrilu02Info_t info,
                        int *position);

If the returned error code is CUSPARSE_STATUS_ZERO_PIVOT, position=j means
A(j,j) has either a structural zero or a numerical zero (the block is not invertible).
Otherwise position=-1.

The position can be 0-based or 1-based, the same as the matrix.

Function cusparseXbsrilu02_zeroPivot() is a blocking call. It calls
cudaDeviceSynchronize() to make sure all previous kernels are done.

The position can be in the host memory or device memory. The user can set proper the
mode with cusparseSetPointerMode().

Input

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>info</td>
<td>info contains structural zero or numerical zero if the user already called bsrilu02_analysis() or bsrilu02().</td>
</tr>
</tbody>
</table>

Output

| position | if no structural or numerical zero, position is -1; otherwise if A(j,j) is missing or U(j,j) is not invertible, position=j. |

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>info is not valid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>

10.3. tridiagonal solve

Different algorithms for tridiagonal solve are discussed in this section.
10.3.1. cusparse<t>gtsv()

This function computes the solution of a tridiagonal linear system with multiple right-hand sides:

\[ A \cdot Y = X \]

The coefficient matrix \( A \) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower (\( dl \)), main (\( d \)), and upper (\( du \)) matrix diagonals; the right-hand sides are stored in the dense matrix \( X \). Notice that solution \( Y \) overwrites right-hand-side matrix \( X \) on exit.

Assuming \( A \) is of size \( m \) and base-1, \( dl \), \( d \) and \( du \) are defined by the following formula:

\( dl(i) := A(i, i-1) \) for \( i=1,2,\ldots,m \)

The first element of \( dl \) is out-of-bound (\( dl(1) := A(1,0) \)), so \( dl(1) = 0 \).

\( d(i) = A(i,i) \) for \( i=1,2,\ldots,m \)

\( du(i) = A(i,i+1) \) for \( i=1,2,\ldots,m \)
The last element of du is out-of-bound \((du(m) := A(m,m+1))\), so \(du(m) = 0\).

The routine does perform pivoting, which usually results in more accurate and more stable results than \(\text{cusparse}<t>\text{gtsv_nopivot()}\) at the expense of some execution time.

This routine requires significant amount of temporary extra storage \((\min(m,8) \times (3+n) \times \text{sizeof(<type>))}\)). The temporary storage is managed by \text{cudaMalloc} and \text{cudaFree} which stop concurrency. The user can call \(\text{cusparse}<t>\text{gtsv2()}\) which has no explicit \text{cudaMalloc} and \text{cudaFree}.

**Input**

| **handle** | handle to the cuSPARSE library context. |
| **m**      | the size of the linear system (must be \(\geq 3\)). |
| **n**      | number of right-hand sides, columns of matrix \(B\). |
| **dl**     | <type> dense array containing the lower diagonal of the tri-diagonal linear system. The first element of each lower diagonal must be zero. |
| **d**      | <type> dense array containing the main diagonal of the tri-diagonal linear system. |
| **du**     | <type> dense array containing the upper diagonal of the tri-diagonal linear system. The last element of each upper diagonal must be zero. |
| **B**      | <type> dense right-hand-side array of dimensions \((1\text{dB}, n)\). |
| **ldb**    | leading dimension of \(B\) (that is \(\geq \max(1, m)\)). |

**Output**

| **B** | <type> dense solution array of dimensions \((1\text{dB}, n)\). |

**Status Returned**

| **CUSPARSE_STATUS_SUCCESS** | the operation completed successfully. |
| **CUSPARSE_STATUS_NOT_INITIALIZED** | the library was not initialized. |
| **CUSPARSE_STATUS_ALLOC_FAILED** | the resources could not be allocated. |
| **CUSPARSE_STATUS_INVALID_VALUE** | invalid parameters were passed \((m<3, n<0)\). |
| **CUSPARSE_STATUS_ARCH_MISMATCH** | the device does not support double precision. |
| **CUSPARSE_STATUS_EXECUTION_FAILED** | the function failed to launch on the GPU. |
| **CUSPARSE_STATUS_INTERNAL_ERROR** | an internal operation failed. |
| **CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED** | the matrix type is not supported. |
10.3.2. cusparse<t>gtsv_nopivot()

cusparseStatus_t

```c
cusparseGtsv_nopivot(cusparseHandle_t handle, int m, int n,
    const float *dl, const float *d,
    const float *du, float *B, int ldb)
```  

cusparseStatus_t

```c
cusparseDgtsv_nopivot(cusparseHandle_t handle, int m, int n,
    const double *dl, const double *d,
    const double *du, double *B, int ldb)
```  

cusparseStatus_t

```c
cusparseCgtsv_nopivot(cusparseHandle_t handle, int m, int n,
    const cuComplex *dl, const cuComplex *d,
    const cuComplex *du, cuComplex *B, int ldb)
```  

cusparseStatus_t

```c
cusparseZgtsv_nopivot(cusparseHandle_t handle, int m, int n,
    const cuDoubleComplex *dl, const cuDoubleComplex *d,
    const cuDoubleComplex *du, cuDoubleComplex *B, int ldb)
```  

This function computes the solution of a tridiagonal linear system with multiple right-hand sides:

\[ A \times Y = X \]

The coefficient matrix \( A \) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower (\( dl \)), main (\( d \)), and upper (\( du \)) matrix diagonals; the right-hand sides are stored in the dense matrix \( X \). Notice that solution \( Y \) overwrites right-hand-side matrix \( X \) on exit.

The routine does not perform any pivoting and uses a combination of the Cyclic Reduction (CR) and the Parallel Cyclic Reduction (PCR) algorithms to find the solution. It achieves better performance when \( m \) is a power of 2.

This routine requires a significant amount of temporary extra storage  
(\( m \times (3+n) \times \text{sizeof(<type>)} \)). The temporary storage is managed by \texttt{cudaMalloc} and \texttt{cudaFree} which stop concurrency. The user can call \texttt{cusparse<t>gtsv2_nopivot()} which has no explicit \texttt{cudaMalloc} and \texttt{cudaFree}.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \text{handle} )</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>( m )</td>
<td>the size of the linear system (must be ( \geq 3 )).</td>
</tr>
<tr>
<td>( n )</td>
<td>number of right-hand sides, columns of matrix ( B ).</td>
</tr>
<tr>
<td>( dl )</td>
<td>(&lt;\text{type}&gt;) dense array containing the lower diagonal of the tri-diagonal linear system. The first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td>( d )</td>
<td>(&lt;\text{type}&gt;) dense array containing the main diagonal of the tri-diagonal linear system.</td>
</tr>
<tr>
<td>( du )</td>
<td>(&lt;\text{type}&gt;) dense array containing the upper diagonal of the tri-diagonal linear system. The last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td>B</td>
<td>&lt;type&gt; dense right-hand-side array of dimensions (ldb, n).</td>
</tr>
<tr>
<td>-----------------</td>
<td>----------------------------------------------------------</td>
</tr>
<tr>
<td>ldb</td>
<td>leading dimension of B. (that is ≥ max(1, m)).</td>
</tr>
</tbody>
</table>

Output

| B               | <type> dense solution array of dimensions (ldb, n).       |

Status Returned

| CUSPARSE_STATUS_SUCCESS       | the operation completed successfully.                   |
| CUSPARSE_STATUS_NOT_INITIALIZED    | the library was not initialized.                       |
| CUSPARSE_STATUS_ALLOC_FAILED    | the resources could not be allocated.                  |
| CUSPARSE_STATUS_INVALID_VALUE   | invalid parameters were passed (m<3, n<0).              |
| CUSPARSE_STATUS_ARCH_MISMATCH   | the device does not support double precision.           |
| CUSPARSE_STATUS_EXECUTION_FAILED| the function failed to launch on the GPU.              |
| CUSPARSE_STATUS_INTERNAL_ERROR  | an internal operation failed.                          |
| CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported.            |
10.3.3. cusparse<t>gtsv2_buffSizeExt()

This function returns the size of the buffer used in gtsv2 which computes the solution of a tridiagonal linear system with multiple right-hand sides.

\[ A \times X = B \]

The coefficient matrix \( A \) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower (\( dl \)), main (\( d \)), and upper (\( du \)) matrix diagonals; the right-hand sides are stored in the dense matrix \( B \). Notice that solution \( X \) overwrites right-hand-side matrix \( B \) on exit.
## Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>the size of the linear system (must be ( \geq 3 )).</td>
</tr>
<tr>
<td><code>n</code></td>
<td>number of right-hand sides, columns of matrix ( B ).</td>
</tr>
<tr>
<td><code>dl</code></td>
<td>(&lt;\text{type}) dense array containing the lower diagonal of the tri-diagonal linear system. The first element of each lower diagonal must be zero.)</td>
</tr>
<tr>
<td><code>d</code></td>
<td>(&lt;\text{type}) dense array containing the main diagonal of the tri-diagonal linear system.)</td>
</tr>
<tr>
<td><code>du</code></td>
<td>(&lt;\text{type}) dense array containing the upper diagonal of the tri-diagonal linear system. The last element of each upper diagonal must be zero.)</td>
</tr>
<tr>
<td><code>B</code></td>
<td>(&lt;\text{type}) dense right-hand-side array of dimensions ((\text{ldb, n})). )</td>
</tr>
<tr>
<td><code>ldb</code></td>
<td>leading dimension of ( B ) (that is ( \geq \max(1, m) )).</td>
</tr>
</tbody>
</table>

## Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>pBufferSizeInBytes</code></td>
<td>number of bytes of the buffer used in the <code>gtsv2</code>.</td>
</tr>
</tbody>
</table>

## Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (( m&lt;3 ), ( n&lt;0 )).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.3.4. cusparse<t>gtsv2()

This function computes the solution of a tridiagonal linear system with multiple right-hand sides:

\[ A \times X = B \]

The coefficient matrix \( A \) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower (\( dL \)), main (\( d \)), and upper (\( du \)) matrix diagonals; the right-hand sides are stored in the dense matrix \( B \). Notice that solution \( X \) overwrites right-hand-side matrix \( B \) on exit.
Assuming $A$ is of size $m$ and base-1, $dl$, $d$ and $du$ are defined by the following formula:

$dl(i) := A(i, \, i-1) \text{ for } i=1,2,\ldots,m$

The first element of $dl$ is out-of-bound ($dl(1) := A(1,0)$), so $dl(1) = 0$.

$d(i) = A(i,i) \text{ for } i=1,2,\ldots,m$

$du(i) = A(i,\, i+1) \text{ for } i=1,2,\ldots,m$

The last element of $du$ is out-of-bound ($du(m) := A(m,m+1)$), so $du(m) = 0$.

The routine does perform pivoting, which usually results in more accurate and more stable results than `cusparse<t>gtsv_nopivot()` or `cusparse<t>gtsv2_nopivot()` at the expense of some execution time.

This function requires a buffer size returned by `gtsv2_bufferSizeExt()`.

The address of pBuffer must be multiple of 128 bytes. If it is not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>$m$</td>
<td>the size of the linear system (must be $\geq 3$).</td>
</tr>
<tr>
<td>$n$</td>
<td>number of right-hand sides, columns of matrix $B$.</td>
</tr>
<tr>
<td>$dl$</td>
<td>$&lt;$type$&gt;$ dense array containing the lower diagonal of the tri-diagonal linear system. The first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td>$d$</td>
<td>$&lt;$type$&gt;$ dense array containing the main diagonal of the tri-diagonal linear system.</td>
</tr>
<tr>
<td>$du$</td>
<td>$&lt;$type$&gt;$ dense array containing the upper diagonal of the tri-diagonal linear system. The last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td>$B$</td>
<td>$&lt;$type$&gt;$ dense right-hand-side array of dimensions ($ldb$, $n$).</td>
</tr>
<tr>
<td>$ldb$</td>
<td>leading dimension of $B$ (that is $\geq \max(1, m)$).</td>
</tr>
<tr>
<td>pBuffer</td>
<td>buffer allocated by the user, the size is return by <code>gtsv2_bufferSizeExt()</code>.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>$B$</td>
<td>$&lt;$type$&gt;$ dense solution array of dimensions ($ldb$, $n$).</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ($m&lt;3$, $n&lt;0$).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
</tbody>
</table>
10.3.5. cusparse<t>gtsv2_nopivot_bufferSizeExt()

This function returns the size of the buffer used in `gtsv2_nopivot` which computes the solution of a tridiagonal linear system with multiple right-hand sides.

\[ A \times X = B \]
The coefficient matrix $A$ of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower ($dl$), main ($d$), and upper ($du$) matrix diagonals; the right-hand sides are stored in the dense matrix $B$. Notice that solution $X$ overwrites right-hand-side matrix $B$ on exit.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>the size of the linear system (must be $\geq 3$).</td>
</tr>
<tr>
<td><code>n</code></td>
<td>number of right-hand sides, columns of matrix $B$.</td>
</tr>
<tr>
<td><code>dl</code></td>
<td><code>&lt;type&gt;</code> dense array containing the lower diagonal of the tri-diagonal linear system. The first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td><code>d</code></td>
<td><code>&lt;type&gt;</code> dense array containing the main diagonal of the tri-diagonal linear system.</td>
</tr>
<tr>
<td><code>du</code></td>
<td><code>&lt;type&gt;</code> dense array containing the upper diagonal of the tri-diagonal linear system. The last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td><code>B</code></td>
<td><code>&lt;type&gt;</code> dense right-hand-side array of dimensions $(ldb, n)$.</td>
</tr>
<tr>
<td><code>ldb</code></td>
<td>leading dimension of $B$. (that is $\geq \max(1, m)$).</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>pBufferSizeInBytes</code></td>
<td>number of bytes of the buffer used in the <code>gtsv2_nopivot</code>.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed ($m&lt;3$, $n&lt;0$).</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_EXECUTION_FAILED</code></td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
10.3.6. cusparse<t>gtsv2_nopivot()

This function computes the solution of a tridiagonal linear system with multiple right-hand sides:

\[ A \times X = B \]

The coefficient matrix \( A \) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower \( (dl) \), main \( (d) \), and upper \( (du) \) matrix diagonals; the right-hand sides are stored in the dense matrix \( B \). Notice that solution \( X \) overwrites right-hand-side matrix \( B \) on exit.
The routine does not perform any pivoting and uses a combination of the Cyclic Reduction (CR) and the Parallel Cyclic Reduction (PCR) algorithms to find the solution. It achieves better performance when $m$ is a power of 2.

This function requires a buffer size returned by `gtsv2_nopivot_bufferSizeExt()`. The address of `pBuffer` must be multiple of 128 bytes. If it is not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>$m$</td>
<td>the size of the linear system (must be ≥ 3).</td>
</tr>
<tr>
<td>$n$</td>
<td>number of right-hand sides, columns of matrix $B$.</td>
</tr>
<tr>
<td>$dl$</td>
<td>&lt;type&gt; dense array containing the lower diagonal of the tri-diagonal linear system. The first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td>$d$</td>
<td>&lt;type&gt; dense array containing the main diagonal of the tri-diagonal linear system.</td>
</tr>
<tr>
<td>$du$</td>
<td>&lt;type&gt; dense array containing the upper diagonal of the tri-diagonal linear system. The last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td>$B$</td>
<td>&lt;type&gt; dense right-hand-side array of dimensions $(ldb, n)$.</td>
</tr>
<tr>
<td>$ldb$</td>
<td>leading dimension of $B$. (that is ≥ max(1, m)).</td>
</tr>
<tr>
<td>$pBuffer$</td>
<td>buffer allocated by the user, the size is return by <code>gtsv2_nopivot_bufferSizeExt</code>.</td>
</tr>
</tbody>
</table>

**Output**

| $B$ | <type> dense solution array of dimensions $(ldb, n)$. |

**Status Returned**

<table>
<thead>
<tr>
<th><code>CUSPARSE_STATUS_SUCCESS</code></th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed ($m&lt;3$, $n&lt;0$).</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_EXECUTION_FAILED</code></td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

### 10.4. batched tridiagonal solve

Different algorithms for batched tridiagonal solve are discussed in this section.
10.4.1. cusparse<t>gtsvStridedBatch()

This function computes the solution of multiple tridiagonal linear systems for \(i=0, \ldots, \text{batchCount} \):

\[
A^0 \cdot y^0 = x^0
\]

The coefficient matrix \(A\) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower (dl), main (d), and upper (du) matrix diagonals; the right-hand sides are stored in the dense matrix \(X\). Notice that solution \(Y\) overwrites right-hand-side matrix \(X\) on exit. The different matrices are assumed to be of the same size and are stored with a fixed \text{batchStride} in memory.

The routine does not perform any pivoting and uses a combination of the Cyclic Reduction (CR) and the Parallel Cyclic Reduction (PCR) algorithms to find the solution. It achieves better performance when \(m\) is a power of 2.

This routine requires a significant amount of temporary extra storage \((\text{batchCount} \times (4 \times m + 2048) \times \text{sizeof(<type>)))\). The temporary storage is managed by \text{cudaMalloc} and \text{cudaFree} which stop concurrency. The user can call \text{cusparse<t>gtsv2StridedBatch()} which has no explicit \text{cudaMalloc} and \text{cudaFree}.

**Input**

<p>| handle | handle to the cuSPARSE library context. |</p>
<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>m</td>
<td>the size of the linear system (must be $\geq 3$).</td>
</tr>
<tr>
<td>dl</td>
<td>$&lt;$type$&gt;$ dense array containing the lower diagonal of the tri-diagonal linear system. The lower diagonal $dL[i]$ that corresponds to the $i^{th}$ linear system starts at location $dl + batchStride \times i$ in memory. Also, the first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td>d</td>
<td>$&lt;$type$&gt;$ dense array containing the main diagonal of the tri-diagonal linear system. The main diagonal $d[i]$ that corresponds to the $i^{th}$ linear system starts at location $d + batchStride \times i$ in memory.</td>
</tr>
<tr>
<td>du</td>
<td>$&lt;$type$&gt;$ dense array containing the upper diagonal of the tri-diagonal linear system. The upper diagonal $dU[i]$ that corresponds to the $i^{th}$ linear system starts at location $du + batchStride \times i$ in memory. Also, the last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td>x</td>
<td>$&lt;$type$&gt;$ dense array that contains the right-hand-side of the tri-diagonal linear system. The right-hand-side $x[i]$ that corresponds to the $i^{th}$ linear system starts at location $x + batchStride \times i$ in memory.</td>
</tr>
<tr>
<td>batchCount</td>
<td>number of systems to solve.</td>
</tr>
<tr>
<td>batchStride</td>
<td>stride (number of elements) that separates the vectors of every system (must be at least $m$).</td>
</tr>
</tbody>
</table>

**Output**

| x          | $<$type$>$ dense array that contains the solution of the tri-diagonal linear system. The solution $x[i]$ that corresponds to the $i^{th}$ linear system starts at location $x + batchStride \times i$ in memory. |

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ($m &lt; 3$, $batchCount \leq 0$, $batchStride &lt; m$).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
10.4.2. \texttt{cusparse\textlangle t\textrangle gtsv2StridedBatch\_bufferSizeExt()}

\begin{verbatim}
cusparseStatus_t cusparseSgtsv2StridedBatch\_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    const float *dl,
    const float *d,
    const float *du,
    const float *x,
    int batchCount,
    int batchStride,
    size_t *bufferSizeInBytes)
cusparseStatus_t cusparseDgtsv2StridedBatch\_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    const double *dl,
    const double *d,
    const double *du,
    const double *x,
    int batchCount,
    int batchStride,
    size_t *bufferSizeInBytes)
cusparseStatus_t cusparseCgtsv2StridedBatch\_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    const cuComplex *dl,
    const cuComplex *d,
    const cuComplex *du,
    const cuComplex *x,
    int batchCount,
    int batchStride,
    size_t *bufferSizeInBytes)
cusparseStatus_t cusparseZgtsv2StridedBatch\_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    const cuDoubleComplex *dl,
    const cuDoubleComplex *d,
    const cuDoubleComplex *du,
    const cuDoubleComplex *x,
    int batchCount,
    int batchStride,
    size_t *bufferSizeInBytes)
\end{verbatim}

This function returns the size of the buffer used in \texttt{gtsv2StridedBatch} which computes the solution of multiple tridiagonal linear systems for \( i=0, \ldots, \text{batchCount} \):

\[ A^{(i)} \cdot Y^{(i)} = X^{(i)} \]

The coefficient matrix \( A \) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower (\( d_l \)), main (\( d \)), and upper (\( d_u \)) matrix diagonals; the right-hand sides are stored in the dense matrix \( X \). Notice that solution \( Y \) overwrites right-
hand-side matrix $X$ on exit. The different matrices are assumed to be of the same size and are stored with a fixed $\text{batchStride}$ in memory.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>n</td>
<td>the size of the linear system (must be $\geq 3$).</td>
</tr>
<tr>
<td>dl</td>
<td>$&lt;$\text{type}$&gt;$ dense array containing the lower diagonal of the tri-diagonal linear system. The lower diagonal $d_l^0$ that corresponds to the $i^{th}$ linear system starts at location $dl+\text{batchStride}\times i$ in memory. Also, the first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td>d</td>
<td>$&lt;$\text{type}$&gt;$ dense array containing the main diagonal of the tri-diagonal linear system. The main diagonal $d_m^0$ that corresponds to the $i^{th}$ linear system starts at location $d+\text{batchStride}\times i$ in memory.</td>
</tr>
<tr>
<td>du</td>
<td>$&lt;$\text{type}$&gt;$ dense array containing the upper diagonal of the tri-diagonal linear system. The upper diagonal $d_u^0$ that corresponds to the $i^{th}$ linear system starts at location $du+\text{batchStride}\times i$ in memory. Also, the last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td>x</td>
<td>$&lt;$\text{type}$&gt;$ dense array that contains the right-hand-side of the tri-diagonal linear system. The right-hand-side $x^0$ that corresponds to the $i^{th}$ linear system starts at location $x+\text{batchStride}\times i$ in memory.</td>
</tr>
<tr>
<td>batchCount</td>
<td>number of systems to solve.</td>
</tr>
<tr>
<td>batchStride</td>
<td>stride (number of elements) that separates the vectors of every system (must be at least $m$).</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>pBufferSizeInBytes</td>
<td>number of bytes of the buffer used in the $\text{gtsv2StridedBatch}$.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ($m&lt;3$, $\text{batchCount}\leq 0$, $\text{batchStride}&lt;m$).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
10.4.3. `cusparse<t>gtsv2StridedBatch()`

```c

cusparseStatus_t cusparseGtsv2StridedBatch(
    cusparseHandle_t handle,
    int m,
    const float *dl,
    const float *d,
    const float *du,
    float *x,
    int batchCount,
    int batchStride,
    void *pBuffer)

cusparseStatus_t cusparseDgtsv2StridedBatch(
    cusparseHandle_t handle,
    int m,
    const double *dl,
    const double *d,
    const double *du,
    double *x,
    int batchCount,
    int batchStride,
    void *pBuffer)

cusparseStatus_t cusparseCgtsv2StridedBatch(
    cusparseHandle_t handle,
    int m,
    const cuComplex *dl,
    const cuComplex *d,
    const cuComplex *du,
    cuComplex *x,
    int batchCount,
    int batchStride,
    void *pBuffer)

cusparseStatus_t cusparseZgtsv2StridedBatch(
    cusparseHandle_t handle,
    int m,
    const cuDoubleComplex *dl,
    const cuDoubleComplex *d,
    const cuDoubleComplex *du,
    cuDoubleComplex *x,
    int batchCount,
    int batchStride,
    void *pBuffer)
```

This function computes the solution of multiple tridiagonal linear systems for \( i=0, \ldots, \text{batchCount} \):

\[
A^{(i)} \cdot y^{(i)} = x^{(i)}
\]

The coefficient matrix \( A \) of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower (\( d_l \)), main (\( d \)), and upper (\( d_u \)) matrix diagonals; the right-hand sides are stored in the dense matrix \( X \). Notice that solution \( Y \) overwrites right-
hand-side matrix $\mathbf{X}$ on exit. The different matrices are assumed to be of the same size and are stored with a fixed batchStride in memory.

The routine does not perform any pivoting and uses a combination of the Cyclic Reduction (CR) and the Parallel Cyclic Reduction (PCR) algorithms to find the solution. It achieves better performance when $m$ is a power of 2.

This function requires a buffer size returned by gtsv2StridedBatch_bufferSizeExt(). The address of pBuffer must be multiple of 128 bytes. If it is not, CUSPARSE_STATUS_INVALID_VALUE is returned.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>n</td>
<td>the size of the linear system (must be ≥ 3).</td>
</tr>
<tr>
<td>dl</td>
<td>$&lt;\text{type}&gt;$ dense array containing the lower diagonal of the tri-diagonal linear system. The lower diagonal $d_l^{(i)}$ that corresponds to the $i^{th}$ linear system starts at location $dl+\text{batchStride}i$ in memory. Also, the first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td>d</td>
<td>$&lt;\text{type}&gt;$ dense array containing the main diagonal of the tri-diagonal linear system. The main diagonal $d^{(i)}$ that corresponds to the $i^{th}$ linear system starts at location $d+\text{batchStride}i$ in memory.</td>
</tr>
<tr>
<td>du</td>
<td>$&lt;\text{type}&gt;$ dense array containing the upper diagonal of the tri-diagonal linear system. The upper diagonal $d_u^{(i)}$ that corresponds to the $i^{th}$ linear system starts at location $du+\text{batchStride}i$ in memory. Also, the last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td>x</td>
<td>$&lt;\text{type}&gt;$ dense array that contains the right-hand-side of the tri-diagonal linear system. The right-hand-side $x^{(i)}$ that corresponds to the $i^{th}$ linear system starts at location $x+\text{batchStride}i$ in memory.</td>
</tr>
<tr>
<td>batchCount</td>
<td>number of systems to solve.</td>
</tr>
<tr>
<td>batchStride</td>
<td>stride (number of elements) that separates the vectors of every system (must be at least $n$).</td>
</tr>
<tr>
<td>pBuffer</td>
<td>buffer allocated by the user, the size is return by gtsv2StridedBatch_bufferSizeExt.</td>
</tr>
</tbody>
</table>

**Output**

| x | $<\text{type}>$ dense array that contains the solution of the tri-diagonal linear system. The solution $x^{(i)}$ that corresponds to the $i^{th}$ linear system starts at location $x+\text{batchStride}i$ in memory. |

**Status Returned**

<p>| CUSPARSE_STATUS_SUCCESS | the operation completed successfully. |</p>
<table>
<thead>
<tr>
<th>cuSPARSE_STATUS</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m&lt;3, batchCount&lt;0, batchStride&lt;m).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
10.4.4. cusparse<t>gtsvInterleavedBatch()

cusparseStatus_t
cusparseSgtsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const float *dl,
  const float *d,
  const float *du,
  const float *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseDgtsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const double *dl,
  const double *d,
  const double *du,
  const double *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseCgtsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const cuComplex *dl,
  const cuComplex *d,
  const cuComplex *du,
  const cuComplex *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseZgtsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const cuDoubleComplex *dl,
  const cuDoubleComplex *d,
  const cuDoubleComplex *du,
  const cuDoubleComplex *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseSgtsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  float *dl,
  float *d,
  float *du,
  float *x,
  int batchCount,
  void *pBuffer)

cusparseStatus_t
cusparseDgtsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  double *dl,
  double *d,
  double *du,
  double *x,
  int batchCount,
  void *pBuffer)

cusparseStatus_t
cusparseCgtsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  cuComplex *dl,
  cuComplex *d,
  cuComplex *du,
  cuComplex *x,
  int batchCount,
  void *pBuffer)

cusparseStatus_t
cusparseZgtsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  cuDoubleComplex *dl,
  cuDoubleComplex *d,
  cuDoubleComplex *du,
  cuDoubleComplex *x,
  int batchCount,
  void *pBuffer)
This function computes the solution of multiple tridiagonal linear systems for $i=0,\ldots,\text{batchCount}$:

$$A^{(i)}x^{(i)}=b^{(i)}$$

The coefficient matrix $A$ of each of these tri-diagonal linear system is defined with three vectors corresponding to its lower ($dl$), main ($d$), and upper ($du$) matrix diagonals; the right-hand sides are stored in the dense matrix $B$. Notice that solution $X$ overwrites right-hand-side matrix $B$ on exit.

Assuming $A$ is of size $m$ and base-1, $dl$, $d$ and $du$ are defined by the following formula:

$dl(i) := A(i, i-1)$ for $i=1,2,\ldots,m$

The first element of $dl$ is out-of-bound ($dl(1) := A(1,0)$), so $dl(1) = 0$.

$d(i) = A(i,i)$ for $i=1,2,\ldots,m$

$du(i) = A(i,i+1)$ for $i=1,2,\ldots,m$

The last element of $du$ is out-of-bound ($du(m) := A(m,m+1)$), so $du(m) = 0$.

The data layout is different from $\text{gtsvStridedBatch}$ which aggregates all matrices one after another. Instead, $\text{gtsvInterleavedBatch}$ gathers different matrices of the same element in a continuous manner. If $dl$ is regarded as a 2-D array of size $m$-by-$\text{batchCount}$, $dl(:,j)$ to store $j$-th matrix. $\text{gtsvStridedBatch}$ uses column-major while $\text{gtsvInterleavedBatch}$ uses row-major.

The routine provides three different algorithms, selected by parameter $\text{algo}$. The first algorithm is $\text{cuThomas}$ provided by Barcelona Supercomputing Center. The second algorithm is LU with partial pivoting and last algorithm is QR. From stability perspective, $\text{cuThomas}$ is not numerically stable because it does not have pivoting. LU with partial pivoting and QR are stable. From performance perspective, LU with partial pivoting and QR is about 10% to 20% slower than $\text{cuThomas}$.

This function requires a buffer size returned by $\text{gtsvInterleavedBatch_bufferSizeExt()}$. The address of $pBuffer$ must be multiple of 128 bytes. If it is not, $\text{CUSPARSE_STATUS_INVALID_VALUE}$ is returned.

Appendix F shows an example of $\text{gtsvInterleavedBatch}$. If the user prepares aggregate format, one can use $\text{cublasXgeam}$ to get interleaved format. However such transformation takes time comparable to solver itself. To reach best performance, the user must prepare interleaved format explicitly.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>algo</td>
<td>algo = 0: $\text{cuThomas}$ (unstable algorithm); algo = 1: LU with pivoting (stable algorithm); algo = 2: QR (stable algorithm)</td>
</tr>
<tr>
<td>m</td>
<td>the size of the linear system.</td>
</tr>
<tr>
<td>dl</td>
<td>&lt;type&gt; dense array containing the lower diagonal of the tri-diagonal linear system. The first element of each lower diagonal must be zero.</td>
</tr>
<tr>
<td>Variable</td>
<td>Description</td>
</tr>
<tr>
<td>-----------</td>
<td>--------------------------------------------------------</td>
</tr>
<tr>
<td>d</td>
<td>&lt;type&gt; dense array containing the main diagonal of the tri-diagonal linear system.</td>
</tr>
<tr>
<td>du</td>
<td>&lt;type&gt; dense array containing the upper diagonal of the tri-diagonal linear system. The last element of each upper diagonal must be zero.</td>
</tr>
<tr>
<td>x</td>
<td>&lt;type&gt; dense right-hand-side array of dimensions (batchCount, n).</td>
</tr>
<tr>
<td>pBuffer</td>
<td>buffer allocated by the user, the size is return by gtsvInterleavedBatch_bufferSizeExt.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>x</td>
<td>&lt;type&gt; dense solution array of dimensions (batchCount, n).</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m&lt;0, batchcount&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

### 10.5. batched pentadiagonal solve

Different algorithms for batched pentadiagonal solve are discussed in this section.
10.5.1. cusparse<t>gpsvInterleavedBatch()

cusparseStatus_t
cusparseSgpsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const float *ds,
  const float *dl,
  const float *d,
  const float *du,
  const float *dw,
  const float *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseDgpsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const double *ds,
  const double *dl,
  const double *d,
  const double *du,
  const double *dw,
  const double *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseCgpsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const cuComplex *ds,
  const cuComplex *dl,
  const cuComplex *d,
  const cuComplex *du,
  const cuComplex *dw,
  const cuComplex *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseZgpsvInterleavedBatch_bufferSizeExt(
  cusparseHandle_t handle,
  int algo,
  int m,
  const cuDoubleComplex *ds,
  const cuDoubleComplex *dl,
  const cuDoubleComplex *d,
  const cuDoubleComplex *du,
  const cuDoubleComplex *dw,
  const cuDoubleComplex *x,
  int batchCount,
  size_t *pBufferSizeInBytes)

cusparseStatus_t
cusparseSgpsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  float *ds,
  float *dl,
  float *d,
  float *du,
  float *dw,
  float *x,
  int batchCount,
  void *pBuffer)

cusparseStatus_t
cusparseDgpsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  double *ds,
  double *dl,
  double *d,
  double *du,
  double *dw,
  double *x,
  int batchCount,
  void *pBuffer)

cusparseStatus_t
cusparseCgpsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  cuComplex *ds,
  cuComplex *dl,
  cuComplex *d,
  cuComplex *du,
  cuComplex *dw,
  cuComplex *x,
  int batchCount,
  void *pBuffer)

cusparseStatus_t
cusparseZgpsvInterleavedBatch(
  cusparseHandle_t handle,
  int algo,
  int m,
  cuDoubleComplex *ds,
  cuDoubleComplex *dl,
  cuDoubleComplex *d,
  cuDoubleComplex *du,
  cuDoubleComplex *dw,
  cuDoubleComplex *x,
  int batchCount,
  void *pBuffer)
This function computes the solution of multiple penta-diagonal linear systems for \( i=0, \ldots, \text{batchCount} \):

\[
A^{(i)} x^{(i)} = b^{(i)}
\]

The coefficient matrix \( A \) of each of these penta-diagonal linear system is defined with five vectors corresponding to its lower (\( ds, dl \)), main (\( d \)), and upper (\( du, dw \)) matrix diagonals; the right-hand sides are stored in the dense matrix \( B \). Notice that solution \( X \) overwrites right-hand-side matrix \( B \) on exit.

Assuming \( A \) is of size \( m \) and base-1, \( ds, dl, d, du \) and \( dw \) are defined by the following formula:

\[
ds(i) := A(i, i-2) \quad \text{for} \quad i=1,2,\ldots,m
\]

The first two elements of \( ds \) is out-of-bound (\( ds(1) := A(1,-1), ds(2) := A(2,0) \)), so \( ds(1) = 0 \) and \( ds(2) = 0 \).

\[
dl(i) := A(i, i-1) \quad \text{for} \quad i=1,2,\ldots,m
\]

The first element of \( dl \) is out-of-bound (\( dl(1) := A(1,0) \)), so \( dl(1) = 0 \).

\[
d(i) = A(i,i) \quad \text{for} \quad i=1,2,\ldots,m
\]

\[
du(i) = A(i,i+1) \quad \text{for} \quad i=1,2,\ldots,m
\]

The last element of \( du \) is out-of-bound (\( du(m) := A(m,m+1) \)), so \( du(m) = 0 \).

\[
dw(i) = A(i,i+2) \quad \text{for} \quad i=1,2,\ldots,m
\]

The last two elements of \( dw \) is out-of-bound (\( dw(m-1) := A(m-1,m+1), dw(m) := A(m,m+2) \)), so \( dw(m-1) = 0 \) and \( dw(m) = 0 \).

The data layout is the same as \texttt{gtsvStridedBatch}.

The routine is numerically stable because it uses QR to solve the linear system.

This function requires a buffer size returned by \texttt{gpsvInterleavedBatch_bufferSizeExt()} . The address of \texttt{pBuffer} must be multiple of 128 bytes. If it is not, \texttt{CUSPARSE_STATUS_INVALID_VALUE} is returned.

Appendix G shows an example of \texttt{gpsvInterleavedBatch} . If the user prepares aggregate format, one can use \texttt{cublasXgeam} to get interleaved format. However such transformation takes time comparable to solver itself. To reach best performance, the user must prepare interleaved format explicitly.

**Input**

<table>
<thead>
<tr>
<th>\textbf{handle}</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>\textbf{algo}</td>
<td>only support algo = 0 (QR)</td>
</tr>
<tr>
<td>\textbf{m}</td>
<td>the size of the linear system.</td>
</tr>
<tr>
<td>\textbf{ds}</td>
<td>&lt;type&gt; dense array containing the lower diagonal (distance 2 to the diagonal) of the penta-diagonal linear system. The first two elements must be zero.</td>
</tr>
</tbody>
</table>
### dl

| dl | <type> dense array containing the lower diagonal (distance 1 to the diagonal) of the penta-diagonal linear system. The first element must be zero. |

### d

| d | <type> dense array containing the main diagonal of the penta-diagonal linear system. |

### du

| du | <type> dense array containing the upper diagonal (distance 1 to the diagonal) of the penta-diagonal linear system. The last element must be zero. |

### dw

| dw | <type> dense array containing the upper diagonal (distance 2 to the diagonal) of the penta-diagonal linear system. The last two elements must be zero. |

### x

| x | <type> dense right-hand-side array of dimensions (batchCount, n). |

### pBuffer

| pBuffer | buffer allocated by the user, the size is return by gpsvInterleavedBatch_bufferSizeExt. |

### Output

| x | <type> dense solution array of dimensions (batchCount, n). |

### Status Returned

<table>
<thead>
<tr>
<th>Status</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m&lt;0, batchcount&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
This chapter describes the reordering routines used to manipulate sparse matrices.

### 11.1. cusparse<t>csrColor()

This function performs the coloring of the adjacency graph associated with the matrix A stored in CSR format. The coloring is an assignment of colors (integer numbers) to nodes, such that neighboring nodes have distinct colors. An approximate coloring algorithm is used in this routine, and is stopped when a certain percentage of nodes has been colored. The rest of the nodes are assigned distinct colors (an increasing sequence of integers numbers, starting from the last integer used previously). The last two auxiliary routines can be used to extract the resulting number of colors, their assignment
and the associated reordering. The reordering is such that nodes that have been assigned 
the same color are reordered to be next to each other.

The matrix A passed to this routine, must be stored as a general matrix and have a 
symmetric sparsity pattern. If the matrix is nonsymmetric the user should pass A+A^T 
as a parameter to this routine.

**Input**

| handle | handle to the cuSPARSE library context. |
| m | number of rows of matrix A. |
| nnz | number of nonzero elements of matrix A. |
| descrA | the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE. |
| csrValA | <type> array of nnz (= csrRowPtrA(m) - csrRowPtrA(0)) nonzero elements of matrix A. |
| csrRowPtrA | integer array of m+1 elements that contains the start of every row and the end of the last row plus one. |
| csrColIndA | integer array of nnz (= csrRowPtrA(m) - csrRowPtrA(0)) column indices of the nonzero elements of matrix A. |
| fractionToColor | fraction of nodes to be colored, which should be in the interval [0.0,1.0], for example 0.8 implies that 80 percent of nodes will be colored. |
| info | structure with information to be passed to the coloring. |

**Output**

| ncolors | The number of distinct colors used (at most the size of the matrix, but likely much smaller). |
| coloring | The resulting coloring permutation |
| reordering | The resulting reordering permutation (untouched if NULL) |

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully. |
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized. |
| CUSPARSE_STATUS_ALLOC FAILED | the resources could not be allocated. |
| CUSPARSE_STATUS_INVALID_VALUE | invalid parameters were passed (m,nnz<0). |
| CUSPARSE_STATUS_ARCH_MISMATCH | the device does not support double precision (compute capability (c.c.) >= 1.3 required). |
| CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed. |
CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED the matrix type is not supported.
Chapter 12.
CUSPARSE FORMAT CONVERSION
REFERENCE

This chapter describes the conversion routines between different sparse and dense storage formats.

cooSort, csrSort, cscSort, csrU2csr and csr2csc_indexOnly are sorting routines without malloc inside, the following table estimates the buffer size

<table>
<thead>
<tr>
<th>routine</th>
<th>buffer size</th>
<th>maximum problem size if buffer is limited by 2GB</th>
</tr>
</thead>
<tbody>
<tr>
<td>cooSort</td>
<td>&gt; 16*n bytes</td>
<td>125M</td>
</tr>
<tr>
<td>csrSort or cscSort</td>
<td>&gt; 20*n bytes</td>
<td>100M</td>
</tr>
<tr>
<td>csrU2csr</td>
<td>'d' &gt; 28<em>n bytes ; 'z' &gt; 36</em>n bytes</td>
<td>71M for 'd' and 55M for 'z'</td>
</tr>
<tr>
<td>csr2csc_indexOnl</td>
<td>&gt; 16*n bytes</td>
<td>125M</td>
</tr>
</tbody>
</table>
12.1. cusparse<

t>bsr2csr()

cusparseStatus_t
cusparseSbsr2csr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int mb,
    int nb,
    const cusparseMatDescr_t descrA,
    const float *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    float *csrValC,
    int *csrRowPtrC,
    int *csrColIndC)
cusparseStatus_t
cusparseDbsr2csr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int mb,
    int nb,
    const cusparseMatDescr_t descrA,
    const double *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    double *csrValC,
    int *csrRowPtrC,
    int *csrColIndC)
cusparseStatus_t
cusparseCbsr2csr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int mb,
    int nb,
    const cusparseMatDescr_t descrA,
    const cuComplex *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    cuComplex *csrValC,
    int *csrRowPtrC,
    int *csrColIndC)
cusparseStatus_t
cusparseZbsr2csr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int mb,
    int nb,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *bsrValA,
    const int *bsrRowPtrA,
    const int *bsrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    cuDoubleComplex *csrValC,
    int *csrRowPtrC,
    int *csrColIndC)
This function converts a sparse matrix in BSR format that is defined by the three arrays `bsrValA`, `bsrRowPtrA`, and `bsrColIndA`) into a sparse matrix in CSR format that is defined by arrays `csrValC`, `csrRowPtrC`, and `csrColIndC`.

Let \( m (=mb \times blockDim) \) be the number of rows of \( A \) and \( n (=nb \times blockDim) \) be number of columns of \( A \), then \( A \) and \( C \) are \( m \times n \) sparse matrices. The BSR format of \( A \) contains \( nnzb (=bsrRowPtrA[mb] - bsrRowPtrA[0]) \) nonzero blocks, whereas the sparse matrix \( A \) contains \( nnz (=nnzb \times blockDim \times blockDim) \) elements. The user must allocate enough space for arrays `csrRowPtrC`, `csrColIndC`, and `csrValC`. The requirements are as follows:

- `csrRowPtrC` of \( m+1 \) elements
- `csrValC` of \( nnz \) elements
- `csrColIndC` of \( nnz \) elements

The general procedure is as follows:

```c
// Given BSR format (bsrRowPtrA, bsrcolIndA, bsrValA) and blocks of BSR format are stored in column-major order.
cusparseDirection_t dir = CUSPARSE_DIRECTION_COLUMN;
int m = mb * blockDim;
int nnzb = bsrRowPtrA[mb] - bsrRowPtrA[0]; // number of blocks
int nnz = nnzb * blockDim * blockDim; // number of elements
cudaMalloc((void**)&csrRowPtrC, sizeof(int)*(m+1));
cudaMalloc((void**)&csrColIndC, sizeof(int)*nnz);
cudaMalloc((void**)&csrValC, sizeof(float)*nnz);
cusparseSbsr2csr(handle, dir, mb, nb,
    descrA,
    bsrValA, bsrRowPtrA, bsrColIndA,
    blockDim,
    descrC,
    csrValC, csrRowPtrC, csrColIndC);
```

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>dir</code></td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td><code>mb</code></td>
<td>number of block rows of sparse matrix ( A ).</td>
</tr>
<tr>
<td><code>nb</code></td>
<td>number of block columns of sparse matrix ( A ).</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix ( A ).</td>
</tr>
<tr>
<td><code>bsrValA</code></td>
<td>( \text{&lt;type&gt;} ) array of ( nnzb \times blockDim \times blockDim ) nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td><code>bsrRowPtrA</code></td>
<td>integer array of ( mb+1 ) elements that contains the start of every block row and the end of the last block row plus one of matrix ( A ).</td>
</tr>
<tr>
<td><code>bsrColIndA</code></td>
<td>integer array of ( nnzb ) column indices of the nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><code>blockDim</code></td>
<td>block dimension of sparse matrix ( A ).</td>
</tr>
<tr>
<td><code>descrC</code></td>
<td>the descriptor of matrix ( C ).</td>
</tr>
</tbody>
</table>

### Output
### csrValC

<type> array of \( \text{nnz} = (\text{csrRowPtrC}[m] - \text{csrRowPtrC}[0]) \) nonzero elements of matrix \( C \).

### csrRowPtrC

integer array of \( m+1 \) elements that contains the start of every row and the end of the last row plus one of matrix \( C \).

### csrColIndC

integer array of \( \text{nnz} \) column indices of the nonzero elements of matrix \( C \).

### Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (( mb, nb&lt;0 ), IndexBase of descrA, descrC is not base-0 or base-1, dir is not row-major or column-major, or blockDim&lt;1).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
12.2. cusparse<t>gebsr2gebsc_bufferSize()

This function returns size of buffer used in computing \texttt{gebsr2gebsc()}.

\textbf{Input}

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>\texttt{handle}</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>\texttt{mb}</td>
<td>number of block rows of sparse matrix A.</td>
</tr>
<tr>
<td>Variable</td>
<td>Description</td>
</tr>
<tr>
<td>------------</td>
<td>------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>nb</td>
<td>number of block columns of sparse matrix ( A ).</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of nonzero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>bsrVal</td>
<td>&lt;type&gt; array of ( nnzb \times \text{rowBlockDim} \times \text{colBlockDim} ) non-zero elements of matrix ( A ).</td>
</tr>
<tr>
<td>bsrRowPtr</td>
<td>integer array of ( mb+1 ) elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td>bsrColInd</td>
<td>integer array of ( nnzb ) column indices of the non-zero blocks of matrix ( A ).</td>
</tr>
<tr>
<td>rowBlockDim</td>
<td>number of rows within a block of ( A ).</td>
</tr>
<tr>
<td>colBlockDim</td>
<td>number of columns within a block of ( A ).</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>pBufferSize</td>
<td>host pointer containing number of bytes of the buffer used in <code>gebsr2gebsc()</code></td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (( mb, nb, nnzb&lt;0 ), or ( \text{rowBlockDim, colBlockDim}&lt;1 )).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
12.3. cusparse<t>gebsr2gebsc()

cusparseStatus_t
 cusparseSgebsr2gebsc(cusparseHandle_t handle,
                  int mb,
                  int nb,
                  int nnzb,
                  const float *bsrVal,
                  const int *bsrRowPtr,
                  const int *bsrColInd,
                  int rowBlockDim,
                  int colBlockDim,
                  float *bscVal,
                  int *bscRowInd,
                  int *bscColPtr,
                  cusparseAction_t copyValues,
                  cusparseIndexBase_t baseIdx,
                  void *pBuffer)

cusparseStatus_t
 cusparseDgebsr2gebsc(cusparseHandle_t handle,
                  int mb,
                  int nb,
                  int nnzb,
                  const double *bsrVal,
                  const int *bsrRowPtr,
                  const int *bsrColInd,
                  int rowBlockDim,
                  int colBlockDim,
                  double *bscVal,
                  int *bscRowInd,
                  int *bscColPtr,
                  cusparseAction_t copyValues,
                  cusparseIndexBase_t baseIdx,
                  void *pBuffer)

cusparseStatus_t
 cusparseCgebsr2gebsc(cusparseHandle_t handle,
                  int mb,
                  int nb,
                  int nnzb,
                  const cuComplex *bsrVal,
                  const int *bsrRowPtr,
                  const int *bsrColInd,
                  int rowBlockDim,
                  int colBlockDim,
                  cuComplex *bscVal,
                  int *bscRowInd,
                  int *bscColPtr,
                  cusparseAction_t copyValues,
                  cusparseIndexBase_t baseIdx,
                  void *pBuffer)

cusparseStatus_t
 cusparseZgebsr2gebsc(cusparseHandle_t handle,
                  int mb,
                  int nb,
                  int nnzb,
                  const cuDoubleComplex *bsrVal,
                  const int *bsrRowPtr,
                  const int *bsrColInd,
                  int rowBlockDim,
                  int colBlockDim,
                  cuDoubleComplex *bscVal,
                  int *bscRowInd,
This function can be seen as the same as `csr2csc()` when each block of size `rowBlockDim*colBlockDim` is regarded as a scalar.

This sparsity pattern of the result matrix can also be seen as the transpose of the original sparse matrix, but the memory layout of a block does not change.

The user must call `gebsr2gebsc_bufferSize()` to determine the size of the buffer required by `gebsr2gebsc()`, allocate the buffer, and pass the buffer pointer to `gebsr2gebsc()`.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>mb</code></td>
<td>number of block rows of sparse matrix $A$.</td>
</tr>
<tr>
<td><code>nb</code></td>
<td>number of block columns of sparse matrix $A$.</td>
</tr>
<tr>
<td><code>nnzb</code></td>
<td>number of nonzero blocks of matrix $A$.</td>
</tr>
<tr>
<td><code>bsrVal</code></td>
<td><code>&lt;type&gt;</code> array of $nnzb \times rowBlockDim \times colBlockDim$ nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td><code>bsrRowPtr</code></td>
<td>integer array of $mb+1$ elements that contains the start of every block row and the end of the last block row plus one.</td>
</tr>
<tr>
<td><code>bsrColInd</code></td>
<td>integer array of $nnzb$ column indices of the non-zero blocks of matrix $A$.</td>
</tr>
<tr>
<td><code>rowBlockDim</code></td>
<td>number of rows within a block of $A$.</td>
</tr>
<tr>
<td><code>colBlockDim</code></td>
<td>number of columns within a block of $A$.</td>
</tr>
<tr>
<td><code>copyValues</code></td>
<td>CUSPARSE_ACTION_SYMBOLIC or CUSPARSE_ACTION_NUMERIC.</td>
</tr>
<tr>
<td><code>baseIdx</code></td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>
| `pBuffer`         | buffer allocated by the user; the size is return by `gebsr2gebsc_bufferSize()`.

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>bscVal</code></td>
<td><code>&lt;type&gt;</code> array of $nnzb \times rowBlockDim \times colBlockDim$ non-zero elements of matrix $A$. It is only filled-in if <code>copyValues</code> is set to CUSPARSE_ACTION_NUMERIC.</td>
</tr>
<tr>
<td><code>bscRowInd</code></td>
<td>integer array of $nnzb$ row indices of the non-zero blocks of matrix $A$.</td>
</tr>
<tr>
<td><code>bscColPtr</code></td>
<td>integer array of $nb+1$ elements that contains the start of every block column and the end of the last block column plus one.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>Status Code</td>
<td>Description</td>
</tr>
<tr>
<td>-------------------------------------------------</td>
<td>-----------------------------------------------------------------------------</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (mb, nb, nnzb&lt;0, baseIdx is not base-0 or base-1, or rowBlockDim, colBlockDim&lt;1).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
12.4. `cusparse<tte>gebsr2gebsr_bufferSize()`

cusparse<tte>gebsr2gebsr_bufferSize(cusparseHandle_t handle,  
cusparseDirection_t dir,  
int mb,  
int nb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const float *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int rowBlockDimA,  
int colBlockDimA,  
int rowBlockDimC,  
int colBlockDimC,  
int *pBufferSize)

cusparse<tte>gebsr2gebsr_bufferSize(cusparseHandle_t handle,  
cusparseDirection_t dir,  
int mb,  
int nb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const double *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int rowBlockDimA,  
int colBlockDimA,  
int rowBlockDimC,  
int colBlockDimC,  
int *pBufferSize)

cusparse<tte>gebsr2gebsr_bufferSize(cusparseHandle_t handle,  
cusparseDirection_t dir,  
int mb,  
int nb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const cuComplex *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int rowBlockDimA,  
int colBlockDimA,  
int rowBlockDimC,  
int colBlockDimC,  
int *pBufferSize)

cusparse<tte>gebsr2gebsr_bufferSize(cusparseHandle_t handle,  
cusparseDirection_t dir,  
int mb,  
int nb,  
int nnzb,  
const cusparseMatDescr_t descrA,  
const cuDoubleComplex *bsrValA,  
const int *bsrRowPtrA,  
const int *bsrColIndA,  
int rowBlockDimA,  
int colBlockDimA,  
int rowBlockDimC,  
int colBlockDimC,  
int *pBufferSize)
This function returns size of the buffer used in computing `gebsr2gebsrNnz()` and `gebsr2gebsr()`.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>dir</td>
<td>storage format of blocks, either <code>CUSPARSE_DIRECTION_ROW</code> or <code>CUSPARSE_DIRECTION_COLUMN</code>.</td>
</tr>
<tr>
<td>mb</td>
<td>number of block rows of sparse matrix A.</td>
</tr>
<tr>
<td>nb</td>
<td>number of block columns of sparse matrix A.</td>
</tr>
<tr>
<td>nnzb</td>
<td>number of nonzero blocks of matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. 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>.</td>
</tr>
<tr>
<td>bsrValA</td>
<td><code>&lt;type&gt;</code> array of <code>nnzb*rowBlockDimA*colBlockDimA</code> non-zero elements of matrix A.</td>
</tr>
<tr>
<td>bsrRowPtrA</td>
<td>integer array of <code>mb+1</code> elements that contains the start of every block row and the end of the last block row plus one of matrix A.</td>
</tr>
<tr>
<td>bsrColIndA</td>
<td>integer array of <code>nnzb</code> column indices of the nonzero blocks of matrix A.</td>
</tr>
<tr>
<td>rowBlockDimA</td>
<td>number of rows within a block of A.</td>
</tr>
<tr>
<td>colBlockDimA</td>
<td>number of columns within a block of A.</td>
</tr>
<tr>
<td>rowBlockDimC</td>
<td>number of rows within a block of C.</td>
</tr>
<tr>
<td>colBlockDimC</td>
<td>number of columns within a block of C.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>pBufferSize</td>
<td>host pointer containing number of bytes of the buffer used in <code>gebsr2gebsr()</code>.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed (<code>mb</code>, <code>nb</code>, <code>nnzb&lt;0</code>; or <code>rowBlockDimA</code>, <code>colBlockDimA</code>, <code>rowBlockDimC</code>, <code>colBlockDimC&lt;1</code>).</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INTERNAL_ERROR</code></td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
12.5. `cusparse< t >gebsr2gebsr()`

cusparseStatus_t
cusparseXgebsr2gebsrNnz(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
int nnzb,
const cusparseMatDescr_t descrA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int rowBlockDimA,
int colBlockDimA,
const cusparseMatDescr_t descrC,
int *bsrRowPtrC,
int rowBlockDimC,
int colBlockDimC,
int *nnzTotalDevHostPtr,
void *pBuffer)

cusparseStatus_t
cusparseSgebsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
int nnzb,
const cusparseMatDescr_t descrA,
const float *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int rowBlockDimA,
int colBlockDimA,
const cusparseMatDescr_t descrC,
float *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDimC,
int colBlockDimC,
void *pBuffer)

cusparseStatus_t
cusparseDgebsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
int nnzb,
const cusparseMatDescr_t descrA,
const double *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int rowBlockDimA,
int colBlockDimA,
const cusparseMatDescr_t descrC,
double *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDimC,
int colBlockDimC,
void *pBuffer)

cusparseStatus_t
cusparseCgebsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
int nnzb,
const cusparseMatDescr_t descrA,
const cuComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int rowBlockDimA,
int colBlockDimA,
const cusparseMatDescr_t descrC,
cuComplex *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDimC,
int colBlockDimC,
void *pBuffer)

cusparseStatus_t
cusparseZgebsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
int nnzb,
const cusparseMatDescr_t descrA,
const cuDoubleComplex *bsrValA,
const int *bsrRowPtrA,
const int *bsrColIndA,
int rowBlockDimA,
int colBlockDimA,
const cusparseMatDescr_t descrC,
cuDoubleComplex *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDimC,
int colBlockDimC,
void *pBuffer)
This function converts a sparse matrix in general BSR format that is defined by the three arrays \texttt{bsrValA}, \texttt{bsrRowPtrA}, and \texttt{bsrColIndA} into a sparse matrix in another general BSR format that is defined by arrays \texttt{bsrValC}, \texttt{bsrRowPtrC}, and \texttt{bsrColIndC}.

If \texttt{rowBlockDimA}=1 and \texttt{colBlockDimA}=1, \texttt{cusparse[S|D|C|Z]gebsr2gebsr()} is the same as \texttt{cusparse[S|D|C|Z]csr2gebsr()}.

If \texttt{rowBlockDimC}=1 and \texttt{colBlockDimC}=1, \texttt{cusparse[S|D|C|Z]gebsr2gebsr()} is the same as \texttt{cusparse[S|D|C|Z]gebsr2csr()}.

\( A \) is an \( m \times n \) sparse matrix where \( m(=mb \times \text{rowBlockDim}) \) is the number of rows of \( A \), and \( n(=nb \times \text{colBlockDim}) \) is the number of columns of \( A \). The general BSR format of \( A \) contains \( \text{nnzb}(=\text{bsrRowPtrA}[mb] - \text{bsrRowPtrA}[0]) \) nonzero blocks. The matrix \( C \) is also general BSR format with a different block size, \( \text{rowBlockDimC} \times \text{colBlockDimC} \).

If \( m \) is not a multiple of \( \text{rowBlockDimC} \), or \( n \) is not a multiple of \( \text{colBlockDimC} \), zeros are filled in. The number of block rows of \( C \) is \( mc(=(m+\text{rowBlockDimC}-1)/\text{rowBlockDimC}) \). The number of block rows of \( C \) is \( nc(=(n+\text{colBlockDimC}-1)/\text{colBlockDimC}) \). The number of nonzero blocks of \( C \) is \( \text{nnzc} \).

The implementation adopts a two-step approach to do the conversion. First, the user allocates \texttt{bsrRowPtrC} of \( mc+1 \) elements and uses function \texttt{cusparseXgebsr2gebsrNnz()} to determine the number of nonzero block columns per block row of matrix \( C \). Second, the user gathers \( \text{nnzc} \) (number of nonzero block columns of matrix \( C \)) from either \( \text{nnzc} = \text{nnzTotalDevHostPtr} \) or \( \text{nnzc} = \text{bsrRowPtrC}[mc] - \text{bsrRowPtrC}[0] \) and allocates \texttt{bsrValC} of \( \text{nnzc} \times \text{rowBlockDimC} \times \text{colBlockDimC} \) elements and \texttt{bsrColIndC} of \( \text{nnzc} \) integers. Finally the function \texttt{cusparse[S|D|C|Z]gebsr2gebsr()} is called to complete the conversion.

The user must call \texttt{gebsr2gebsr_bufferSize()} to know the size of the buffer required by \texttt{gebsr2gebsr()}, allocate the buffer, and pass the buffer pointer to \texttt{gebsr2gebsr()}. 
The general procedure is as follows:

```c
// Given general BSR format (bsrRowPtrA, bsrColIndA, bsrValA) and
// blocks of BSR format are stored in column-major order.
cusparseDirection_t dir = CUSPARSE_DIRECTION_COLUMN;
int base, nnzc;
int m = mb*rowBlockDimA;
int n = nb*colBlockDimA;
int mc = (m+rowBlockDimC-1)/rowBlockDimC;
int nc = (n+colBlockDimC-1)/colBlockDimC;
int bufferSize;
void *pBuffer;
cusparseSgebsr2gebsr_bufferSize(handle, dir, mb, nb, nnzb,
    descrA, bsrValA, bsrRowPtrA, bsrColIndA,
    rowBlockDimA, colBlockDimA,
    rowBlockDimC, colBlockDimC,
    &bufferSize);
cudaMalloc((void**)&pBuffer, bufferSize);
// nnzTotalDevHostPtr points to host memory
int *nnzTotalDevHostPtr = &nnzc;
cusparseXgebsr2gebsrNnz(handle, dir, mb, nb, nnzb,
    descrA, bsrRowPtrA, bsrColIndA,
    rowBlockDimA, colBlockDimA,
    descrC, bsrRowPtrC,
    rowBlockDimC, colBlockDimC,
    nnzTotalDevHostPtr, pBuffer);
if (NULL != nnzTotalDevHostPtr){
    nnzc = *nnzTotalDevHostPtr;
} else{
    cudaMemcpy(&nnzc, bsrRowPtrC+mc, sizeof(int), cudaMemcpyDeviceToHost);
    cudaMemcpy(&base, bsrRowPtrC, sizeof(int), cudaMemcpyDeviceToHost);
    nnzc -= base;
}
cudaMalloc((void**)&bsrColIndC, sizeof(int)*nnzc);
cudaMalloc((void**)&bsrValC, sizeof(float)*(rowBlockDimC*colBlockDimC)*nnzc);
cusparseSgebsr2gebsr(handle, dir, mb, nb, nnzb,
    descrA, bsrValA, bsrRowPtrA, bsrColIndA,
    rowBlockDimA, colBlockDimA,
    descrC, bsrValC, bsrRowPtrC, bsrColIndC,
    rowBlockDimC, colBlockDimC,
    pBuffer);
```

**Input**

<table>
<thead>
<tr>
<th></th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>handle</strong></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><strong>dir</strong></td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td><strong>mb</strong></td>
<td>number of block rows of sparse matrix A.</td>
</tr>
<tr>
<td><strong>nb</strong></td>
<td>number of block columns of sparse matrix A.</td>
</tr>
<tr>
<td><strong>nnzb</strong></td>
<td>number of nonzero blocks of matrix A.</td>
</tr>
<tr>
<td><strong>descrA</strong></td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td><strong>bsrValA</strong></td>
<td>array of nnzb<em>rowBlockDimA</em>colBlockDimA non-zero elements of matrix A.</td>
</tr>
</tbody>
</table>
bsrRowPtrA | integer array of mb+1 elements that contains the start of every block row and the end of the last block row plus one of matrix A.

bsrColIndA | integer array of nnzb column indices of the non-zero blocks of matrix A.

rowBlockDimA | number of rows within a block of A.

colBlockDimA | number of columns within a block of A.

descrC | the descriptor of matrix C. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.

rowBlockDimC | number of rows within a block of C.

colBlockDimC | number of columns within a block of C.

 pBuffer | buffer allocated by the user; the size is return by gebsr2gebsr_bufferSize().

Output

bsrValC | <type> array of nnzc*rowBlockDimC*colBlockDimC non-zero elements of matrix C.

bsrRowPtrC | integer array of mc+1 elements that contains the start of every block row and the end of the last block row plus one of matrix C.

bsrColIndC | integer array of nnzc block column indices of the nonzero blocks of matrix C.

nnzTotalDevHostPtr | total number of nonzero blocks of C. *nnzTotalDevHostPtr is the same as bsrRowPtrC[mc]−bsrRowPtrC[0].

Status Returned

CUSPARSE_STATUS_SUCCESS | the operation completed successfully.

CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized.

CUSPARSE_STATUS_ALLOC_FAILED | the resources could not be allocated.

CUSPARSE_STATUS_INVALID_VALUE | invalid parameters were passed (mb, nb, nnzb<0, baseIdx is not base-0 or base-1; or rowBlockDimA, colBlockDimA, rowBlockDimC, colBlockDimC<1).

CUSPARSE_STATUS_ARCH_MISMATCH | the device does not support double precision.

CUSPARSE_STATUS_EXECUTION_FAILED | the function failed to launch on the GPU.

CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported.

CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed.
12.6. \( \text{cusparse} < t > \text{gebsr2csr}() \)

```c
cusparseStatus_t
cusparseSgebsr2csr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
const cusparseMatDescr_t descrA,
const float *bsrValA,
const int * bsrRowPtrA,
const int * bsrColIndA,
int rowBlockDim,
int colBlockDim,
const cusparseMatDescr_t descrC,
float *csrValC,
int *csrRowPtrC,
int *csrColIndC )

cusparseStatus_t
cusparseDgebsr2csr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
const cusparseMatDescr_t descrA,
const double *bsrValA,
const int * bsrRowPtrA,
const int * bsrColIndA,
int rowBlockDim,
int colBlockDim,
const cusparseMatDescr_t descrC,
double *csrValC,
int *csrRowPtrC,
int *csrColIndC )

cusparseStatus_t
cusparseCgebsr2csr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
const cusparseMatDescr_t descrA,
cuComplex * bsrValA,
const int * bsrRowPtrA,
const int * bsrColIndA,
int rowBlockDim,
int colBlockDim,
const cusparseMatDescr_t descrC,
cuComplex *csrValC,
int *csrRowPtrC,
int *csrColIndC )

cusparseStatus_t
cusparseZgebsr2csr(cusparseHandle_t handle,
cusparseDirection_t dir,
int mb,
int nb,
const cusparseMatDescr_t descrA,
cuDoubleComplex *bsrValA,
const int * bsrRowPtrA,
const int * bsrColIndA,
int rowBlockDim,
int colBlockDim,
const cusparseMatDescr_t descrC,
cuDoubleComplex *csrValC,
int *csrRowPtrC,
int *csrColIndC )
```
This function converts a sparse matrix in general BSR format that is defined by the three arrays `bsrValA`, `bsrRowPtrA`, and `bsrColIndA` into a sparse matrix in CSR format that is defined by arrays `csrValC`, `csrRowPtrC`, and `csrColIndC`.

Let \( m(=mb \times \text{rowBlockDim}) \) be number of rows of \( A \) and \( n(=nb \times \text{colBlockDim}) \) be number of columns of \( A \), then \( A \) and \( C \) are \( m \times n \) sparse matrices. The general BSR format of \( A \) contains \( \text{nnzb} = \text{bsrRowPtrA}[mb] - \text{bsrRowPtrA}[0] \) non-zero blocks, whereas sparse matrix \( A \) contains \( \text{nnz} = \text{nnzb} \times \text{rowBlockDim} \times \text{colBlockDim} \) elements. The user must allocate enough space for arrays `csrRowPtrC`, `csrColIndC`, and `csrValC`. The requirements are as follows:

- `csrRowPtrC` of \( m+1 \) elements
- `csrValC` of \( \text{nnz} \) elements
- `csrColIndC` of \( \text{nnz} \) elements

The general procedure is as follows:

```c
// Given general BSR format (bsrRowPtrA, bsrColIndA, bsrValA) and
// blocks of BSR format are stored in column-major order.
int m = mb*rowBlockDim;
int n = nb*colBlockDim;
int nnzb = bsrRowPtrA[mb] - bsrRowPtrA[0]; // number of blocks
int nnz  = nnzb * rowBlockDim * colBlockDim; // number of elements
cudaMalloc((void**)&csrRowPtrC, sizeof(int)*(m+1));
cudaMalloc((void**)&csrColIndC, sizeof(int)*nnz);
cudaMalloc((void**)&csrValC, sizeof(float)*nnz);
cusparseSgebsr2csr(handle, dir, mb, nb,
        descrA, bsrValA, bsrRowPtrA, bsrColIndA,
        rowBlockDim, colBlockDim,
        descrC, csrValC, csrRowPtrC, csrColIndC);
```

### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>dir</code></td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td><code>mb</code></td>
<td>number of block rows of sparse matrix ( A ).</td>
</tr>
<tr>
<td><code>nb</code></td>
<td>number of block columns of sparse matrix ( A ).</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix ( A ). The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td><code>bsrValA</code></td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnzb} \times \text{rowBlockDim} \times \text{colBlockDim} ) non-zero elements of matrix ( A ).</td>
</tr>
<tr>
<td><code>bsrRowPtrA</code></td>
<td>integer array of ( m+1 ) elements that contains the start of every block row and the end of the last block row plus one of matrix ( A ).</td>
</tr>
<tr>
<td><code>bsrColIndA</code></td>
<td>integer array of ( \text{nnzb} ) column indices of the non-zero blocks of matrix ( A ).</td>
</tr>
<tr>
<td><strong>rowBlockDim</strong></td>
<td>number of rows within a block of A.</td>
</tr>
<tr>
<td>-----------------</td>
<td>-------------------------------------</td>
</tr>
<tr>
<td><strong>colBlockDim</strong></td>
<td>number of columns within a block of A.</td>
</tr>
<tr>
<td><strong>descrC</strong></td>
<td>the descriptor of matrix C. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th><strong>csrValC</strong></th>
<th>&lt;type&gt; array of ( \text{nnz} ) non-zero elements of matrix C.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>csrRowPtrC</strong></td>
<td>integer array of ( m+1 ) elements that contains the start of every row and the end of the last row plus one of matrix C.</td>
</tr>
<tr>
<td><strong>csrColIndC</strong></td>
<td>integer array of ( \text{nnz} ) column indices of the non-zero elements of matrix C.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th><strong>CUSPARSE_STATUS_SUCCESS</strong></th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>CUSPARSE_STATUS_NOT_INITIALIZED</strong></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_ALLOC_FAILED</strong></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_INVALID_VALUE</strong></td>
<td>invalid parameters were passed (( m, nb &lt; 0 ) is not base-0 or base-1, or ( \text{rowBlockDim}, \text{colBlockDim} &lt; 1 )).</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_ARCH_MISMATCH</strong></td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_EXECUTION_FAILED</strong></td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</strong></td>
<td>the matrix type is not supported.</td>
</tr>
<tr>
<td><strong>CUSPARSE_STATUS_INTERNAL_ERROR</strong></td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
12.7. cusparse<t>csr2gebsr_bufferSize()

This function returns the size of the buffer used in computing \texttt{csr2gebsrNnz} and \texttt{csr2gebsr}.
### Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>dir</td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of sparse matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of sparse matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>csrValA</td>
<td>&lt;type&gt; array of ( nnz ) nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of ( m+1 ) elements that contains the start of every row and the end of the last row plus one of matrix A.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of ( nnz ) column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>descrC</td>
<td>the descriptor of matrix C. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>rowBlockDim</td>
<td>number of rows within a block of C.</td>
</tr>
<tr>
<td>colBlockDim</td>
<td>number of columns within a block of C.</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>pBufferSize</td>
<td>host pointer containing number of bytes of the buffer used in csr2gebsrNnz() and csr2gebsr().</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n&lt;0, or rowBlockDim, colBlockDim&lt;1).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
12.8. `cusparse<t>csr2gebsr()`

cusparsestatus_t

cusparseXcsr2gebsrNnz(cusparseHandle_t handle,
cusparseDirection_t dir,
int m,
int n,
const cusparseMatDescr_t descrA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparsesparseMatDescr_t descrC,
int *bsrRowPtrC,
int rowBlockDim,
int colBlockDim,
int *nnzTotalDevHostPtr,
void *pBuffer )

cusparsestatus_t

cusparseScsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int m,
int n,
const cusparseMatDescr_t descrA,
const float *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparsesparseMatDescr_t descrC,
float *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDim,
int colBlockDim,
void *pBuffer)

cusparsestatus_t

cusparseDcsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int m,
int n,
const cusparseMatDescr_t descrA,
const double *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparsesparseMatDescr_t descrC,
double *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDim,
int colBlockDim,
void *pBuffer)

cusparsestatus_t

cusparseCsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int m,
int n,
const cusparseMatDescr_t descrA,
const cuComplex *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparsesparseMatDescr_t descrC,
cuComplex *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDim,
void *pBuffer)

cusparsestatus_t

cusparseZcsr2gebsr(cusparseHandle_t handle,
cusparseDirection_t dir,
int m,
int n,
const cusparseMatDescr_t descrA,
const cuDoubleComplex *csrValA,
const int *csrRowPtrA,
const int *csrColIndA,
cusparsesparseMatDescr_t descrC,
cuDoubleComplex *bsrValC,
int *bsrRowPtrC,
int *bsrColIndC,
int rowBlockDim,
void *pBuffer)
This function converts a sparse matrix \( A \) in CSR format (that is defined by arrays \( \text{csrValA} \), \( \text{csrRowPtrA} \), and \( \text{csrColIndA} \)) into a sparse matrix \( C \) in general BSR format (that is defined by the three arrays \( \text{bsrValC} \), \( \text{bsrRowPtrC} \), and \( \text{bsrColIndC} \)).

The matrix \( A \) is a \( m \times n \) sparse matrix and matrix \( C \) is a \((mb \times \text{rowBlockDim}) \times (nb \times \text{colBlockDim})\) sparse matrix, where \( mb = (m + \text{rowBlockDim}-1)/\text{rowBlockDim} \) is the number of block rows of \( C \), and \( nb = (n + \text{colBlockDim}-1)/\text{colBlockDim} \) is the number of block columns of \( C \).

The block of \( C \) is of size \( \text{rowBlockDim} \times \text{colBlockDim} \). If \( m \) is not multiple of \( \text{rowBlockDim} \) or \( n \) is not multiple of \( \text{colBlockDim} \), zeros are filled in.

The implementation adopts a two-step approach to do the conversion. First, the user allocates \( \text{bsrRowPtrC} \) of \( mb+1 \) elements and uses function \( \text{cusparseXcsr2gebsrNnz()} \) to determine the number of nonzero block columns per block row. Second, the user gathers \( nnzb \) (number of nonzero block columns of matrix \( C \)) from either \( (nnzb=\text{nnzTotalDevHostPtr}) \) or \( (nnzb=\text{bsrRowPtrC}[mb]-\text{bsrRowPtrC}[0]) \) and allocates \( \text{bsrValC} \) of \( nnzb \times \text{rowBlockDim} \times \text{colBlockDim} \) elements and \( \text{bsrColIndC} \) of \( nnzb \) integers. Finally function \( \text{cusparse[S|D|C|Z]csr2gebsr()} \) is called to complete the conversion.

The user must obtain the size of the buffer required by \( \text{csr2gebsr()} \) by calling \( \text{csr2gebsr_bufferSize()} \), allocate the buffer, and pass the buffer pointer to \( \text{csr2gebsr()} \).
The general procedure is as follows:

```c
// Given CSR format (csrRowPtrA, csrColIndA, csrValA) and
// blocks of BSR format are stored in column-major order.
cusparseDirection_t dir = CUSPARSE_DIRECTION_COLUMN;
int base, nnzb;
int mb = (m + rowBlockDim - 1)/rowBlockDim;
int nb = (n + colBlockDim - 1)/colBlockDim;
int bufferSize;
void *pBuffer;
cusparseScsr2gebsr_bufferSize(handle, dir, m, n,
descrA, csrValA, csrRowPtrA, csrColIndA,
rowBlockDim, colBlockDim,
&bufferSize);
cudaMalloc((void**)&pBuffer, bufferSize);
cudaMalloc((void**)&bsrRowPtrC, sizeof(int) *(mb+1));
// nnzTotalDevHostPtr points to host memory
int *nnzTotalDevHostPtr = &nnzb;
cusparseXcsr2gebsrNnz(handle, dir, m, n,
descrA, csrRowPtrA, csrColIndA,
descrC, bsrRowPtrC, rowBlockDim, colBlockDim,
nnzTotalDevHostPtr,
pBuffer);
if (NULL != nnzTotalDevHostPtr){
  nnzb = *nnzTotalDevHostPtr;
}else{
  cudaMemcpy(&nnzb, bsrRowPtrC+mb, sizeof(int), cudaMemcpyDeviceToHost);
  cudaMemcpy(&base, bsrRowPtrC, sizeof(int), cudaMemcpyDeviceToHost);
  nnzb -= base;
}
cudaMalloc((void**)&bsrColIndC, sizeof(int)*nnzb);
cudaMalloc((void**)&bsrValC, sizeof(float)*(rowBlockDim*colBlockDim)*nnzb);
cusparseScsr2gebsr(handle, dir, m, n,
descrA, csrValA, csrRowPtrA, csrColIndA,
descrC, bsrValC, bsrRowPtrC, bsrColIndC,
rowBlockDim, colBlockDim,
pBuffer);
```

## Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>handle</strong></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><strong>dir</strong></td>
<td>storage format of blocks, either CUSPARSE_DIRECTION_ROW or CUSPARSE_DIRECTION_COLUMN.</td>
</tr>
<tr>
<td><strong>m</strong></td>
<td>number of rows of sparse matrix A.</td>
</tr>
<tr>
<td><strong>n</strong></td>
<td>number of columns of sparse matrix A.</td>
</tr>
<tr>
<td><strong>descrA</strong></td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td><strong>csrValA</strong></td>
<td>&lt;type&gt; array of nnz nonzero elements of matrix A.</td>
</tr>
<tr>
<td><strong>csrRowPtrA</strong></td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one of matrix A.</td>
</tr>
<tr>
<td><strong>csrColIndA</strong></td>
<td>integer array of nnz column indices of the nonzero elements of matrix A.</td>
</tr>
</tbody>
</table>
The descriptor of matrix C. The supported matrix type is `CUSPARSE_MATRIX_TYPE_GENERAL`. Also, the supported index bases are `CUSPARSE_INDEX_BASE_ZERO` and `CUSPARSE_INDEX_BASE_ONE`.

The number of rows within a block of C.

The number of columns within a block of C.

Buffer allocated by the user, the size is return by `csr2gebsr_bufferSize()`.

The number of nonzero elements of matrix C.

Integer array of `mb+1` elements containing the start of every block row and the end of the last block row plus one of matrix C.

Integer array of `nnzb` column indices of the nonzero blocks of matrix C.

Total number of nonzero blocks of matrix C. Pointer `nnzTotalDevHostPtr` can point to a device memory or host memory.

CUSPARSE_STATUS_SUCCESS: the operation completed successfully.

CUSPARSE_STATUS_NOT_INITIALIZED: the library was not initialized.

CUSPARSE_STATUS_ALLOC_FAILED: the resources could not be allocated.

CUSPARSE_STATUS_INVALID_VALUE: invalid parameters were passed (m,n<0, baseIdx is not base-0 or base-1, or rowBlockDim, colBlockDim<1).

CUSPARSE_STATUS_ARCH_MISMATCH: the device does not support double precision.

CUSPARSE_STATUS_EXECUTION_FAILED: the function failed to launch on the GPU.

CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED: the matrix type is not supported.

CUSPARSE_STATUS_INTERNAL_ERROR: an internal operation failed.

12.9. `cusparse<t>coo2csr()`

This function converts the array containing the uncompressed row indices (corresponding to COO format) into an array of compressed row pointers (corresponding to CSR format).
It can also be used to convert the array containing the uncompressed column indices (corresponding to COO format) into an array of column pointers (corresponding to CSC format).

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>cooRowInd</td>
<td>integer array of ( \text{nnz} ) uncompressed row indices.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of non-zeros of the sparse matrix (that is also the length of array cooRowInd).</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>idxBase</td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

**Output**

| csrRowPtr               | integer array of \( m+1 \) elements that contains the start of every row and the end of the last row plus one. |

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>( \text{idxBase} ) is neither CUSPARSE_INDEX_BASE_ZERO nor CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
</tbody>
</table>
12.10. cusparse<t>csc2dense()

This function converts the sparse matrix in CSC format that is defined by the three arrays `cscValA`, `cscColPtrA`, and `cscRowIndA` into the matrix `A` in dense format. The dense matrix `A` is filled in with the values of the sparse matrix and with zeros elsewhere.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>number of rows of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>n</code></td>
<td>number of columns of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix <code>A</code>. 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>.</td>
</tr>
<tr>
<td><code>cscValA</code></td>
<td><code>&lt;type&gt;</code> array of <code>nnz (= cscColPtrA(m) - cscColPtrA(0))</code> nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>cscRowIndA</code></td>
<td>integer array of <code>nnz (= cscColPtrA(m) - cscColPtrA(0))</code> row indices of the nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td><code>cscColPtrA</code></td>
<td>integer array of <code>n+1</code> elements that contains the start of every row and the end of the last column plus one.</td>
</tr>
<tr>
<td><code>lda</code></td>
<td>leading dimension of dense array <code>A</code>.</td>
</tr>
</tbody>
</table>
Output

| A | array of dimensions (lda, n) that is filled in with the values of the sparse matrix. |

Status Returned

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

12.11. cusparse<t>csc2hyb()

cusparseStatus_t

cusparseScsc2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const float *cscValA,
    const int *cscRowIndA, const int *cscColPtrA,
    cusparseHybMat_t hybA, int userEllWidth,
    cusparseHybPartition_t partitionType)

cusparseStatus_t

cusparseDcsc2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const double *cscValA,
    const int *cscRowIndA, const int *cscColPtrA,
    cusparseHybMat_t hybA, int userEllWidth,
    cusparseHybPartition_t partitionType)

cusparseStatus_t

cusparseCcsc2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const cuComplex *cscValA,
    const int *cscRowIndA, const int *cscColPtrA,
    cusparseHybMat_t hybA, int userEllWidth,
    cusparseHybPartition_t partitionType)

cusparseStatus_t

cusparseZcsc2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *cscValA,
    const int *cscRowIndA, const int *cscColPtrA,
    cusparseHybMat_t hybA, int userEllWidth,
    cusparseHybPartition_t partitionType)

This function converts a sparse matrix in CSC format into a sparse matrix in HYB format. It assumes that the hybA parameter has been initialized with the cusparseCreateHybMat() routine before calling this function.

This function requires some amount of temporary storage and a significant amount of storage for the matrix in HYB format. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

Input
<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>cscValA</td>
<td>&lt;type&gt; array of nnz (= cscColPtrA(m) - cscColPtrA(0)) nonzero elements of matrix A.</td>
</tr>
<tr>
<td>cscRowIndA</td>
<td>integer array of nnz (= cscColPtrA(m) - cscColPtrA(0)) column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>cscColPtrA</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>userEllWidth</td>
<td>width of the regular (ELL) part of the matrix in HYB format, which should be less than the maximum number of nonzeros per row and is only required if partitionType == CUSPARSE_HYB_PARTITION_USER.</td>
</tr>
<tr>
<td>partitionType</td>
<td>partitioning method to be used in the conversion (please refer to cusparseHybPartition_t for details).</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>hybA</td>
<td>the matrix A in HYB storage format.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.12. cusparse<t>csr2bsr()

cusparseStatus_t
cusparseXcsr2bsrNnz(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int m,
    int n,
    const cusparseMatDescr_t descrA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    int *bsrRowPtrC,
    int *nnzTotalDevHostPtr)

cusparseStatus_t
cusparseScsr2bsr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int m,
    int n,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    float *bsrValC,
    int *bsrRowPtrC,
    int *bsrColIndC)

cusparseStatus_t
cusparseDcsr2bsr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int m,
    int n,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    double *bsrValC,
    int *bsrRowPtrC,
    int *bsrColIndC)

cusparseStatus_t
cusparseCcsr2bsr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int m,
    int n,
    const cusparseMatDescr_t descrA,
    const cuComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    cuComplex *bsrValC,
    int *bsrRowPtrC,
    int *bsrColIndC)

cusparseStatus_t
cusparseZcsr2bsr(cusparseHandle_t handle,
    cusparseDirection_t dir,
    int m,
    int n,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    int blockDim,
    const cusparseMatDescr_t descrC,
    cuDoubleComplex *bsrValC,
    int *bsrRowPtrC,
    int *bsrColIndC)
This function converts a sparse matrix in CSR format that is defined by the three arrays `csrValA`, `csrRowPtrA`, and `csrColIndA` into a sparse matrix in BSR format that is defined by arrays `bsrValC`, `bsrRowPtrC`, and `bsrColIndC`.

A is an \(m \times n\) sparse matrix. The BSR format of A has \(mb\) block rows, \(nb\) block columns, and \(nnzb\) nonzero blocks, where \(mb=((m+\text{blockDim}-1)/\text{blockDim})\) and \(nb=(n+\text{blockDim}-1)/\text{blockDim}\).

If \(m\) or \(n\) is not multiple of \(\text{blockDim}\), zeros are filled in.

The conversion in cuSPARSE entails a two-step approach. First, the user allocates `bsrRowPtrC` of \(mb+1\) elements and uses function `cusparseXcsr2bsrNnz()` to determine the number of nonzero block columns per block row. Second, the user gathers `nnzb` (number of non-zero block columns of matrix C) from either 
\[(\text{nnzb}^*\text{nnzTotalDevHostPtr})\] or \[(\text{nnzb}=\text{bsrRowPtrC}[mb]-\text{bsrRowPtrC}[0])\] and allocates `bsrValC` of `nnzb`\*`blockDim`\*`blockDim` elements and `bsrColIndC` of `nnzb` elements. Finally function `cusparse[S|D|C|Z]csr2bsr90` is called to complete the conversion.

The general procedure is as follows:

```c
// Given CSR format (csrRowPtrA, csrColIndA, csrValA) and blocks of BSR format are stored in column-major order.
cusparseDirection_t dir = CUSPARSE_DIRECTION_COLUMN;
int base, nnzb;
int mb = (m + blockDim-1)/blockDim;
cudaMalloc((void**)&bsrRowPtrC, sizeof(int) *(mb+1));
// nnzTotalDevHostPtr points to host memory
int *nnzTotalDevHostPtr = &nnzb;
cusparseXcsr2bsrNnz(handle, dir, m, n,
  descrA, csrRowPtrA, csrColIndA, blockDim,
  descrC, bsrRowPtrC, 
  nnzTotalDevHostPtr);
if (NULL != nnzTotalDevHostPtr){
  nnzb = *nnzTotalDevHostPtr;
} else{
  cudaMemcpy(&nnzb, bsrRowPtrC+mb, sizeof(int), cudaMemcpyDeviceToHost);
  cudaMemcpy(&base, bsrRowPtrC, sizeof(int), cudaMemcpyDeviceToHost);
  nnzb -= base;
}
cudaMalloc((void**)&bsrColIndC, sizeof(int)*nnzb);
cudaMalloc((void**)&bsrValC, sizeof(float)*(blockDim*blockDim)*nnzb);
cusparseScsr2bsr(handle, dir, m, n,
  descrA, csrValA, csrRowPtrA, csrColIndA, blockDim,
  descrC, bsrValC, bsrRowPtrC, bsrColIndC);
```

If `blockDim` is large (typically, a block cannot fit into shared memory), `cusparse[S|D|C|Z]csr2bsr()` allocates a temporary integer array of size `mb`\*`blockDim` integers. If device memory is not available, `CUSPARSE_STATUS_ALLOC_FAILED` is returned.

**Input**

| `handle` | handle to the cuSPARSE library context. |
| `dir` | storage format of blocks, either `CUSPARSE_DIRECTION_ROW` or `CUSPARSE_DIRECTION_COLUMN`. |
### m
The number of rows of sparse matrix A.

### n
The number of columns of sparse matrix A.

### descrA
The descriptor of matrix A.

### csrValA
A `<type>` array of `nnz (=csrRowPtrA[m] - csrRowPtrA[0])` non-zero elements of matrix A.

### csrRowPtrA
An integer array of `m+1` elements that contains the start of every row and the end of the last row plus one.

### csrColIndA
An integer array of `nnz` column indices of the non-zero elements of matrix A.

### blockDim
The block dimension of sparse matrix A. The range of `blockDim` is between 1 and `min(m,n)`.

### descrC
The descriptor of matrix C.

#### Output

### bsrValC
A `<type>` array of `nnzb*blockDim*blockDim` non-zero elements of matrix C.

### bsrRowPtrC
An integer array of `mb+1` elements that contains the start of every block row and the end of the last block row plus one of matrix C.

### bsrColIndC
An integer array of `nnzb` column indices of the non-zero blocks of matrix C.

### nnzTotalDevHostPtr
The total number of non-zero elements in device or host memory. It is equal to `(bsrRowPtrC[mb] - bsrRowPtrC[0])`.

#### Status Returned

**CUSPARSE_STATUS_SUCCESS**
The operation completed successfully.

**CUSPARSE_STATUS_NOT_INITIALIZED**
The library was not initialized.

**CUSPARSE_STATUS_ALLOC_FAILED**
The resources could not be allocated.

**CUSPARSE_STATUS_INVALID_VALUE**
Invalid parameters were passed (m, n<0). `IndexBase` field of descrA, descrC is not base-0 or base-1, `dir` is not row-major or column-major, or `blockDim` is not between 1 and `min(m,n)`.

**CUSPARSE_STATUS_ARCH_MISMATCH**
The device does not support double precision.

**CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED**
The matrix type is not supported.

**CUSPARSE_STATUS_INTERNAL_ERROR**
An internal operation failed.

---

### 12.13. cusparse<t>csr2coo()

```c
cusparseStatus_t
cusparseXcsr2coo(cusparseHandle_t handle, const int *csrRowPtr,
                int nnz, int m, int *cooRowInd,
                cusparseIndexBase_t idxBase)
```
This function converts the array containing the compressed row pointers (corresponding to CSR format) into an array of uncompressed row indices (corresponding to COO format).

It can also be used to convert the array containing the compressed column indices (corresponding to CSC format) into an array of uncompressed column indices (corresponding to COO format).

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th><strong>handle</strong></th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>csrRowPtr</strong></td>
<td>integer array of (m+1) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><strong>nnz</strong></td>
<td>number of nonzeros of the sparse matrix (that is also the length of array cooRowInd).</td>
</tr>
<tr>
<td><strong>m</strong></td>
<td>number of rows of matrix (A).</td>
</tr>
<tr>
<td><strong>idxBase</strong></td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

**Output**

| **cooRowInd** | integer array of \(nnz\) uncompressed row indices. |

**Status Returned**

| **CUSPARSE_STATUS_SUCCESS** | the operation completed successfully. |
| **CUSPARSE_STATUS_NOT_INITIALIZED** | the library was not initialized. |
| **CUSPARSE_STATUS_INVALID_VALUE** | idxBase is neither CUSPARSE_INDEX_BASE_ZERO nor CUSPARSE_INDEX_BASE_ONE. |
| **CUSPARSE_STATUS_EXECUTION_FAILED** | the function failed to launch on the GPU. |
This function converts a sparse matrix in CSR format (that is defined by the three arrays \texttt{csrVal}, \texttt{csrRowPtr}, and \texttt{csrColInd}) into a sparse matrix in CSC format (that is defined by arrays \texttt{cscVal}, \texttt{cscRowInd}, and \texttt{cscColPtr}). The resulting matrix can also be seen as the transpose of the original sparse matrix. Notice that this routine can also be used to convert a matrix in CSC format into a matrix in CSR format.

This function requires a significant amount of extra storage that is proportional to the matrix size. It is executed asynchronously with respect to the host, and it may return control to the application on the host before the result is ready.

\textbf{Input}

\begin{tabular}{|l|l|}
\hline
\textbf{handle} & handle to the cuSPARSE library context. \\
\textbf{m} & number of rows of matrix $A$. \\
\textbf{n} & number of columns of matrix $A$. \\
\textbf{nnz} & number of nonzero elements of matrix $A$. \\
\textbf{csrVal} & <type> array of \texttt{nnz} (= \texttt{csrRowPtr}(m) – \texttt{csrRowPtr}(0)) nonzero elements of matrix $A$. \\
\textbf{csrRowPtr} & integer array of \texttt{m}+1 elements that contains the start of every row and the end of the last row plus one. \\
\hline
\end{tabular}
<table>
<thead>
<tr>
<th>csrColInd</th>
<th>integer array of nnz (= csrRowPtr(m) - csrRowPtr(0)) column indices of the nonzero elements of matrix A.</th>
</tr>
</thead>
<tbody>
<tr>
<td>copyValues</td>
<td>CUSPARSE_ACTION_SYMBOLIC or CUSPARSE_ACTION_NUMERIC.</td>
</tr>
<tr>
<td>idxBase</td>
<td>CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>cscVal</th>
<th>&lt;type&gt; array of nnz (= cscColPtr(n) - cscColPtr(0)) nonzero elements of matrix A. It is only filled in if copyValues is set to CUSPARSE_ACTION_NUMERIC.</th>
</tr>
</thead>
<tbody>
<tr>
<td>cscRowInd</td>
<td>integer array of nnz (= cscColPtr(n) - cscColPtr(0)) column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>cscColPtr</td>
<td>integer array of n+1 elements that contains the start of every column and the end of the last column plus one.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m,n,nnz&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>

**12.15. cusparseCsr2cscEx()**

```c

int *cscSortedRowInd,
int *cscSortedColPtr,
cusparseAction_t copyValues,
cusparseIndexBase_t idxBase,
cudaDataType executiontype);

This function is an extended version of cusparse<t>csr2csc(). For detailed description of the functionality, see cusparse<t>csr2csc().
```
This function does not support half-precision execution type, but it supports half-precision IO with single precision execution.

**Input specifically required by cusparseCsr2cscEx**

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrSortedValAtype</td>
<td>Data type of csrSortedValA.</td>
</tr>
<tr>
<td>cscSortedValAtype</td>
<td>Data type of cscSortedValA.</td>
</tr>
<tr>
<td>executiontype</td>
<td>Data type used for computation.</td>
</tr>
</tbody>
</table>

### 12.16. cusparse<t>csr2dense()

This function converts the sparse matrix in CSR format (that is defined by the three arrays `csrValA`, `csrRowPtrA`, and `csrColIndA`) into the matrix `A` in dense format. The dense matrix `A` is filled in with the values of the sparse matrix and with zeros elsewhere.
This function requires no extra storage. It is executed asynchronously with respect to the host, and it may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>csrValA</td>
<td>&lt;type&gt; array of nnz (= csrRowPtrA(m) - csrRowPtrA(0)) nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of nnz (= csrRowPtrA(m) - csrRowPtrA(0)) column indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td>lda</td>
<td>leading dimension of array matrix A.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>array of dimensions (lda,n) that is filled in with the values of the sparse matrix.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m,n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.17. `cusparse<t>csr2csr_compress()`

cusparseStatus_t cusparseScsr2csr_compress(cusparseHandle_t handle, 
  int m, 
  int n, 
  const cusparseMatDescr_t descrA, 
  const float *csrValA, 
  const int *csrColIndA, 
  const int *csrRowPtrA, 
  int nnzA, 
  const int *nnzPerRow, 
  float *csrValC, 
  int *csrColIndC, 
  int *csrRowPtrC, 
  float tol);

cusparseStatus_t cusparseDcsr2csr_compress(cusparseHandle_t handle, 
  int m, 
  int n, 
  const cusparseMatDescr_t descrA, 
  const double *csrValA, 
  const int *csrColIndA, 
  const int *csrRowPtrA, 
  int nnzA, 
  const int *nnzPerRow, 
  double *csrValC, 
  int *csrColIndC, 
  int *csrRowPtrC, 
  double tol);

cusparseStatus_t cusparseCcsr2csr_compress(cusparseHandle_t handle, 
  int m, 
  int n, 
  const cusparseMatDescr_t descrA, 
  const cuComplex *csrValA, 
  const int *csrColIndA, 
  const int *csrRowPtrA, 
  int nnzA, 
  const int *nnzPerRow, 
  cuComplex *csrValC, 
  int *csrColIndC, 
  int *csrRowPtrC, 
  cuComplex tol);

cusparseStatus_t cusparseZcsr2csr_compress(cusparseHandle_t handle, 
  int m, 
  int n, 
  const cusparseMatDescr_t descrA, 
  const cuDoubleComplex *csrValA, 
  const int *csrColIndA, 
  const int *csrRowPtrA, 
  int nnzA, 
  const int *nnzPerRow, 
  cuDoubleComplex *csrValC, 
  int *csrColIndC, 
  int *csrRowPtrC, 
  cuDoubleComplex tol);
This function compresses the sparse matrix in CSR format into compressed CSR format. Given a sparse matrix $A$ and a non-negative value threshold (in the case of complex values, only the magnitude of the real part is used in the check), the function returns a sparse matrix $C$, defined by

$$C(i,j) = A(i,j) \text{ if } |A(i,j)| > \text{threshold}$$

The implementation adopts a two-step approach to do the conversion. First, the user allocates $\text{csrRowPtrC}$ of $m+1$ elements and uses function $\text{cusparse<t>nnz_compress()}$ to determine $\text{nnzPerRow}$ (the number of nonzeros columns per row) and $\text{nnzC}$ (the total number of nonzeros). Second, the user allocates $\text{csrValC}$ of $\text{nnzC}$ elements and $\text{csrColIndC}$ of $\text{nnzC}$ integers. Finally, function $\text{cusparse<t>csr2csr_compress()}$ is called to complete the conversion.

**Input**

<table>
<thead>
<tr>
<th><strong>handle</strong></th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>m</strong></td>
<td>number of rows of matrix $A$.</td>
</tr>
<tr>
<td><strong>n</strong></td>
<td>number of columns of matrix $A$.</td>
</tr>
<tr>
<td><strong>descrA</strong></td>
<td>the descriptor of matrix $A$. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td><strong>csrValA</strong></td>
<td>$&lt;$type$&gt;$ array of $\text{nnz} (= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0))$ elements of matrix $A$.</td>
</tr>
<tr>
<td><strong>csrColIndA</strong></td>
<td>integer array of $\text{nnz} (= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0))$ column indices of the elements of matrix $A$.</td>
</tr>
<tr>
<td><strong>csrRowPtrA</strong></td>
<td>integer array of $m+1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td><strong>nnzA</strong></td>
<td>number of nonzero elements in matrix $A$.</td>
</tr>
<tr>
<td><strong>nnzPerRow</strong></td>
<td>this array contains the number of elements kept in the compressed matrix, by row.</td>
</tr>
<tr>
<td><strong>tol</strong></td>
<td>on input, this contains the non-negative tolerance value used for compression. Any values in matrix $A$ less than or equal to this value will be dropped during compression.</td>
</tr>
</tbody>
</table>

**Output**

| **csrValC** | on output, this array contains the typed values of elements kept in the compressed matrix. Size = $\text{nnzC}$. |
| **csrColIndC** | on output, this integer array contains the column indices of elements kept in the compressed matrix. Size = $\text{nnzC}$. |


| csrRowPtrC | on output, this integer array contains the row pointers for elements kept in the compressed matrix. Size = m+1 |

### Status Returned

- **CUSPARSE_STATUS_SUCCESS** - the operation completed successfully.
- **CUSPARSE_STATUS_NOT_INITIALIZED** - the library was not initialized.
- **CUSPARSE_STATUS_INVALID_VALUE** - invalid parameters were passed (m,n<0).
- **CUSPARSE_STATUS_ARCH_MISMATCH** - the device does not support double precision.
- **CUSPARSE_STATUS_EXECUTION_FAILED** - the function failed to launch on the GPU.
- **CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED** - the matrix type is not supported.

The following is a sample code to show how to use this API.

```c
#include <stdio.h>
#include <sys/time.h>
#include <cusparse.h>

#define ERR_NE(X,Y) do { if ((X) != (Y)) { 
    fprintf(stderr,"Error in %s at %s:%d
\n",__func__,__FILE__,__LINE__); 
    exit(-1);}} while(0)
#define CUDA_CALL(X) ERR_NE((X),cudaSuccess)
#define CUSPARSE_CALL(X) ERR_NE((X),CUSPARSE_STATUS_SUCCESS)

int main(){
    int m = 6, n = 5;
    cusparseHandle_t handle;
    CUSPARSE_CALL( cusparseCreate(&handle) );
    cusparseMatDescr_t descrX;
    CUSPARSE_CALL(cusparseCreateMatDescr(&descrX));
    // Initialize sparse matrix
    float *X;
    CUDA_CALL(cudamALLOCN MANAGED ( &X, sizeof(float) * m * n ));
    memset( X, 0, sizeof(float) * m * n );
    X[0 + 0*m] = 1.0;  X[0 + 1*m] = 3.0;
    X[1 + 1*m] = -4.0; X[1 + 2*m] = 5.0;
    X[2 + 0*m] = 2.0;  X[2 + 3*m] = 7.0;  X[2 + 4*m] = 8.0;
    X[3 + 2*m] = 6.0;  X[3 + 4*m] = 9.0;
    X[4 + 3*m] = 3.5;  X[4 + 4*m] = 5.5;
    X[5 + 0*m] = 6.5;  X[5 + 2*m] = -9.9;
    // Initialize total_nnz, and nnzPerRowX for cusparsedense2csr()
    int total_nnz = 13;
    int *nnzPerRowX;
    CUDA_CALL( cudamALLOCN MANAGED( &nnzPerRowX, sizeof(int) * m ));
    nnzPerRowX[0] = 2;  nnzPerRowX[1] = 2;  nnzPerRowX[2] = 3;
    float *csrValX;
    int *csrRowPtrX;
    int *csrColIndX;
    CUDA_CALL( cudamALLOCN MANAGED( &csrValX, sizeof(float) * total_nnz ) );
    CUDA_CALL( cudamALLOCN MANAGED( &csrRowPtrX, sizeof(int) * (m+1) ) );
    CUDA_CALL( cudamALLOCN MANAGED( &csrColIndX, sizeof(int) * total_nnz ) );
}
```
Before calling this API, call two APIs to prepare the input.

```c
/** Call cusparseSdense2csr to generate CSR format as the inputs for
cusparseScsr2csr_compress **/
cusparseSdense2csr( handle, m, n, descrX, X,
m, nnzPerRowX, csrValX,
csrRowPtrX, csrColIndX ) ;

float tol = 3.5;
int *nnzPerRowY;
int *testNNZTotal;
cudaMallocManaged( &nnzPerRowY, sizeof(int) * m );
cudaMallocManaged( &testNNZTotal, sizeof(int) );
memset( nnzPerRowY, 0, sizeof(int) * m );
// cusparseSnnz_compress generates nnzPerRowY and testNNZTotal
CUSPARSE_CALL( cusparseSnnz_compress( handle, m, descrX, csrValX,
csrRowPtrX, nnzPerRowY,
testNNZTotal, tol ) );

float *csrValY;
int *csrRowPtrY;
int *csrColIndY;
cudaMallocManaged( &csrValY, sizeof(float) * (*testNNZTotal) );
cudaMallocManaged( &csrRowPtrY, sizeof(int) * (m+1) );
cudaMallocManaged( &csrColIndY, sizeof(int) * (*testNNZTotal) );

CUSPARSE_CALL( cusparseScsr2csr_compress( handle, m, n, descrX, csrValX,
csrColIndX, csrRowPtrX,
total_nnz, nnzPerRowY,
csrValY, csrColIndY,
csrRowPtrY, tol ) );

/* Expect results
nnzPerRowY:  0 2 2 2 1 2
csrValY:     -4 5 7 8 6 9 5.5 6.5 -9.9
csrColIndY:  1 2 3 4 2 4 4 0 2
csrRowPtrY:  0 0 2 4 6 7 9
*/
cudaFree(X);
cusparseDestroy(handle);
cudaFree(nnzPerRowX);
cudaFree(csrValX);
cudaFree(csrRowPtrX);
cudaFree(csrColIndX);
cudaFree(csrValY);
cudaFree(nnzPerRowY);
cudaFree(testNNZTotal);
cudaFree(csrRowPtrY);
cudaFree(csrColIndY);
return 0;
})
```
12.18. cusparse<t>csr2hyb()

cusparseStatus_t
 cusparseScsr2hyb(cusparseHandle_t handle, int m, int n,
                      const cusparseMatDescr_t descrA,
                      const float *csrValA,
                      const int *csrRowPtrA, const int *csrColIndA,
                      cusparseHybMat_t hybA, int userEllWidth,
                      cusparseHybPartition_t partitionType)

cusparseStatus_t
 cusparseDcsr2hyb(cusparseHandle_t handle, int m, int n,
                      const cusparseMatDescr_t descrA,
                      const double *csrValA,
                      const int *csrRowPtrA, const int *csrColIndA,
                      cusparseHybMat_t hybA, int userEllWidth,
                      cusparseHybPartition_t partitionType)

cusparseStatus_t
 cusparseCcsr2hyb(cusparseHandle_t handle, int m, int n,
                      const cusparseMatDescr_t descrA,
                      const cuComplex *csrValA,
                      const int *csrRowPtrA, const int *csrColIndA,
                      cusparseHybMat_t hybA, int userEllWidth,
                      cusparseHybPartition_t partitionType)

cusparseStatus_t
 cusparseZcsr2hyb(cusparseHandle_t handle, int m, int n,
                      const cusparseMatDescr_t descrA,
                      const cuDoubleComplex *csrValA,
                      const int *csrRowPtrA, const int *csrColIndA,
                      cusparseHybMat_t hybA, int userEllWidth,
                      cusparseHybPartition_t partitionType)

This function converts a sparse matrix in CSR format into a sparse matrix in
HYB format. It assumes that the hybA parameter has been initialized with
cusparseCreateHybMat() routine before calling this function.

This function requires some amount of temporary storage and a significant amount
of storage for the matrix in HYB format. It is executed asynchronously with respect to the
host and may return control to the application on the host before the result is ready.

Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>csrValA</td>
<td>&lt;type&gt; array of nnz (= csrRowPtrA(m) - csrRowPtrA(0)) nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowPtrA</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>Variable</td>
<td>Description</td>
</tr>
<tr>
<td>-------------------</td>
<td>-----------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>csrColIndA</td>
<td>integer array of ( \text{nnz}(= \text{csrRowPtrA}(m) - \text{csrRowPtrA}(0)) ) column indices of the nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>userEllWidth</td>
<td>width of the regular (ELL) part of the matrix in HYB format, which should be less than maximum number of nonzeros per row and is only required if ( \text{partitionType} == \text{CUSPARSE_HYB_PARTITION_USER} ).</td>
</tr>
<tr>
<td>partitionType</td>
<td>partitioning method to be used in the conversion (please refer to \text{cusparseHybPartition_t} for details).</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>hybA</td>
<td>the matrix ( A ) in HYB storage format.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ((m,n&lt;0)).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.19. cusparse<t>dense2csc()

cusparseStatus_t
cusparseSdense2csc(cusparseHandle_t handle, int m, int n,
const cusparseMatDescr_t descrA,
const float *A,
int lda, const int *nnzPerCol,
float *cscValA,
int *cscRowIndA, int *cscColPtrA)
cusparseStatus_t
cusparseDdense2csc(cusparseHandle_t handle, int m, int n,
const cusparseMatDescr_t descrA,
const double *A,
int lda, const int *nnzPerCol,
double *cscValA,
int *cscRowIndA, int *cscColPtrA)
cusparseStatus_t
cusparseCdense2csc(cusparseHandle_t handle, int m, int n,
const cusparseMatDescr_t descrA,
const cuComplex *A,
int lda, const int *nnzPerCol,
cuComplex *cscValA,
int *cscRowIndA, int *cscColPtrA)
cusparseStatus_t
cusparseZdense2csc(cusparseHandle_t handle, int m, int n,
const cusparseMatDescr_t descrA,
const cuDoubleComplex *A,
int lda, const int *nnzPerCol,
cuDoubleComplex *cscValA,
int *cscRowIndA, int *cscColPtrA)

This function converts the matrix A in dense format into a sparse matrix in CSC format. All the parameters are assumed to have been pre-allocated by the user, and the arrays are filled in based on nnzPerCol, which can be precomputed with cusparse<t>nnz().

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix A.</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>A</td>
<td>array of dimensions (lda, n).</td>
</tr>
<tr>
<td>lda</td>
<td>leading dimension of dense array A.</td>
</tr>
<tr>
<td>nnzPerCol</td>
<td>array of size n containing the number of nonzero elements per column.</td>
</tr>
</tbody>
</table>
Output

<table>
<thead>
<tr>
<th></th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>cscValA</code></td>
<td>&lt;type&gt; array of nnz (= cscRowPtrA(m) - cscRowPtrA(0)) nonzero elements of matrix A. It is only filled if copyValues is set to CUSPARSE_ACTION_NUMERIC.</td>
</tr>
<tr>
<td><code>cscRowIndA</code></td>
<td>integer array of nnz (= cscRowPtrA(m) - cscRowPtrA(0)) row indices of the nonzero elements of matrix A.</td>
</tr>
<tr>
<td><code>cscColPtrA</code></td>
<td>integer array of n+1 elements that contains the start of every column and the end of the last column plus one.</td>
</tr>
</tbody>
</table>

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

12.20. cusparse<t>dense2csr()  

```c
 cusparseStatus_t cusparseSdense2csr(cusparseHandle_t handle, int m, int n,  
                                   const cusparseMatDescr_t descrA,  
                                   const float *A,  
                                   int lda, const int *nnzPerRow,  
                                   float *csrValA,  
                                   int *csrRowPtrA, int *csrColIndA)

cusparseStatus_t cusparseDdense2csr(cusparseHandle_t handle, int m, int n,  
                                   const cusparseMatDescr_t descrA,  
                                   const double *A,  
                                   int lda, const int *nnzPerRow,  
                                   double *csrValA,  
                                   int *csrRowPtrA, int *csrColIndA)

cusparseStatus_t cusparseCdense2csr(cusparseHandle_t handle, int m, int n,  
                                   const cusparseMatDescr_t descrA,  
                                   const cuComplex *A,  
                                   int lda, const int *nnzPerRow,  
                                   cuComplex *csrValA,  
                                   int *csrRowPtrA, int *csrColIndA)

cusparseStatus_t cusparseZdense2csr(cusparseHandle_t handle, int m, int n,  
                                   const cusparseMatDescr_t descrA,  
                                   const cuDoubleComplex *A,  
                                   int lda, const int *nnzPerRow,  
                                   cuDoubleComplex *csrValA,  
                                   int *csrRowPtrA, int *csrColIndA)
```
This function converts the matrix $A$ in dense format into a sparse matrix in CSR format. All the parameters are assumed to have been pre-allocated by the user and the arrays are filled in based on `nnzPerRow`, which can be pre-computed with `cusparse<t>nnz()`.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>m</code></td>
<td>number of rows of matrix $A$.</td>
</tr>
<tr>
<td><code>n</code></td>
<td>number of columns of matrix $A$.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix $A$. 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>.</td>
</tr>
<tr>
<td><code>A</code></td>
<td>array of dimensions $(lda, n)$.</td>
</tr>
<tr>
<td><code>lda</code></td>
<td>leading dimension of dense array $A$.</td>
</tr>
<tr>
<td><code>nnzPerRow</code></td>
<td>array of size $n$ containing the number of non-zero elements per row.</td>
</tr>
</tbody>
</table>

Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>csrValA</code></td>
<td><code>&lt;type&gt;</code> array of $nnz (= csrRowPtrA(m) - csrRowPtrA(0))$ nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td><code>csrRowPtrA</code></td>
<td>integer array of $m+1$ elements that contains the start of every column and the end of the last column plus one.</td>
</tr>
<tr>
<td><code>csrColIndA</code></td>
<td>integer array of $nnz (= csrRowPtrA(m) - csrRowPtrA(0))$ column indices of the non-zero elements of matrix $A$.</td>
</tr>
</tbody>
</table>

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_NOT_INITIALIZED</code></td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ALLOC_FAILED</code></td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_INVALID_VALUE</code></td>
<td>invalid parameters were passed ($m$, $n&lt;0$).</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_ARCH_MISMATCH</code></td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_EXECUTION_FAILED</code></td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td><code>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</code></td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.21. \texttt{cusparse<t>dense2hyb()}

\begin{verbatim}
cusparseStatus_t

cusparseSdense2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const float           *A,
    int lda, const int *nnzPerRow, cusparseHybMat_t hybA,
    int userEllWidth,
    cusparseHybPartition_t partitionType)

cusparseStatus_t

cusparseDdense2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const double          *A,
    int lda, const int *nnzPerRow, cusparseHybMat_t
    hybA,
    int userEllWidth,
    cusparseHybPartition_t partitionType)

cusparseStatus_t

cusparseCdense2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const cuComplex       *A,
    int lda, const int *nnzPerRow, cusparseHybMat_t
    hybA,
    int userEllWidth,
    cusparseHybPartition_t partitionType)

cusparseStatus_t

cusparseZdense2hyb(cusparseHandle_t handle, int m, int n,
    const cusparseMatDescr_t descrA,
    const cuDoubleComplex *A,
    int lda, const int *nnzPerRow, cusparseHybMat_t
    hybA,
    int userEllWidth,
    cusparseHybPartition_t partitionType)
\end{verbatim}

This function converts matrix \(A\) in dense format into a sparse matrix in HYB format. It assumes that the routine \texttt{cusparseCreateHybMat()} was used to initialize the opaque structure \(hybA\) and that the array \(nnzPerRow\) was pre-computed with \texttt{cusparse<t>nnz()}. This function requires some amount of temporary storage and a significant amount of storage for the matrix in HYB format. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

\textbf{Input}

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>number of rows of matrix (A).</td>
</tr>
<tr>
<td>n</td>
<td>number of columns of matrix (A).</td>
</tr>
<tr>
<td>descrA</td>
<td>the descriptor of matrix (A). The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}.</td>
</tr>
<tr>
<td>A</td>
<td>array of dimensions ((lda, n)).</td>
</tr>
<tr>
<td>lda</td>
<td>leading dimension of dense array (A).</td>
</tr>
<tr>
<td>nnzPerRow</td>
<td>array of size (m) containing the number of nonzero elements per row.</td>
</tr>
</tbody>
</table>
userEllWidth | width of the regular (ELL) part of the matrix in HYB format, which should be less than maximum number of nonzeros per row and is only required if `partitionType == CUSPARSE_HYB_PARTITION_USER`.

| partitionType | partitioning method to be used in the conversion (please refer to `cusparseHybPartition_t` for details).

**Output**

| hybA | the matrix A in HYB storage format.

**Status Returned**

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully.
|-------------------------|------------------------------------------|
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized.
| CUSPARSE_STATUS_ALLOC_FAILED | the resources could not be allocated.
| CUSPARSE_STATUS_INVALID_VALUE | invalid parameters were passed (m, n<0).
| CUSPARSE_STATUS_ARCH_MISMATCH | the device does not support double precision.
| CUSPARSE_STATUS_EXECUTION_FAILED | the function failed to launch on the GPU.
| CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed.
| CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported.

### 12.22. cusparse<t>hyb2csc()

cusparseStatus_t cusparseShyb2csc(cusparseHandle_t handle,  
const cusparseMatDescr_t descrA,  
const cusparseHybMat_t hybA,  
float       *cscValA, int *cscRowIndA, int *cscColPtrA)
cusparseStatus_t cusparseDhyb2csc(cusparseHandle_t handle,  
const cusparseMatDescr_t descrA,  
const cusparseHybMat_t hybA,  
double      *cscValA, int *cscRowIndA, int *cscColPtrA)
cusparseStatus_t cusparseChyb2csc(cusparseHandle_t handle,  
const cusparseMatDescr_t descrA,  
const cusparseHybMat_t hybA,  
cuComplex   *cscValA, int *cscRowIndA, int *cscColPtrA)
cusparseStatus_t cusparseZhyb2csc(cusparseHandle_t handle,  
const cusparseMatDescr_t descrA,  
const cusparseHybMat_t hybA,  
cuDoubleComplex *cscValA, int *cscRowIndA, int

This function converts a sparse matrix in HYB format into a sparse matrix in CSC format.
This function requires some amount of temporary storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A in Hyb format. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL.</td>
</tr>
<tr>
<td>hybA</td>
<td>the matrix A in HYB storage format.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>cscValA</th>
<th>&lt;type&gt; array of nnz (= cscColPtrA(m) - cscColPtrA(0)) nonzero elements of matrix A.</th>
</tr>
</thead>
<tbody>
<tr>
<td>cscRowIndA</td>
<td>integer array of nnz (= cscColPtrA(m) - cscColPtrA(0)) column indices of the non-zero elements of matrix A.</td>
</tr>
<tr>
<td>cscColPtrA</td>
<td>integer array of m+1 elements that contains the start of every column and the end of the last row plus one.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>CUSPARSE_STATUS_SUCCESS</th>
<th>the operation completed successfully.</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.23. `cusparse<t>hyb2csr()`

```c
 cusparseStatus_t
cusparseShyb2csr(cusparseHandle_t handle,
    const cusparseMatDescr_t descrA,
    const cusparseHybMat_t hybA,
    float       *csrValA, int *csrRowPtrA, int *csrColIndA)

cusparseStatus_t
cusparseDhyb2csr(cusparseHandle_t handle,
    const cusparseMatDescr_t descrA,
    const cusparseHybMat_t hybA,
    double      *csrValA, int *csrRowPtrA, int *csrColIndA)

cusparseStatus_t
cusparseChyb2csr(cusparseHandle_t handle,
    const cusparseMatDescr_t descrA,
    const cusparseHybMat_t hybA,
    cuComplex   *csrValA, int *csrRowPtrA, int *csrColIndA)

cusparseStatus_t
cusparseZhyb2csr(cusparseHandle_t handle,
    const cusparseMatDescr_t descrA,
    const cusparseHybMat_t hybA,
    cuDoubleComplex *csrValA, int *csrRowPtrA, int
    *csrColIndA)
```

This function converts a sparse matrix in HYB format into a sparse matrix in CSR format.

This function requires some amount of temporary storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>handle</code></td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td><code>descrA</code></td>
<td>the descriptor of matrix A in Hyb format. The supported matrix type is <code>CUSPARSE_MATRIX_TYPE_GENERAL</code>.</td>
</tr>
<tr>
<td><code>hybA</code></td>
<td>the matrix A in HYB storage format.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>csrValA</code></td>
<td>&lt;type&gt; array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> nonzero elements of matrix A.</td>
</tr>
<tr>
<td><code>csrRowPtrA</code></td>
<td>integer array of <code>m+1</code> elements that contains the start of every column and the end of the last row plus one.</td>
</tr>
<tr>
<td><code>csrColIndA</code></td>
<td>integer array of <code>nnz (= csrRowPtrA(m) - csrRowPtrA(0))</code> column indices of the nonzero elements of matrix A.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>CUSPARSE_STATUS_SUCCESS</code></td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>-------------------------------</td>
<td>---------------------------------</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

### 12.24. cusparse<t>hyb2dense()

```c
 cusparseStatus_t
cusparseShyb2dense(cusparseHandle_t handle,
                   const cusparseMatDescr_t descrA,
                   const cusparseHybMat_t hybA,
                   float *A,
                   int lda)                             

cusparseStatus_t
cusparseDhyb2dense(cusparseHandle_t handle,
                   const cusparseMatDescr_t descrA,
                   const cusparseHybMat_t hybA,
                   double *A,
                   int lda)                              

cusparseStatus_t
cusparseChyb2dense(cusparseHandle_t handle,
                   const cusparseMatDescr_t descrA,
                   const cusparseHybMat_t hybA,
                   cuComplex *A,
                   int lda)                              

cusparseStatus_t
cusparseZhyb2dense(cusparseHandle_t handle,
                   const cusparseMatDescr_t descrA,
                   const cusparseHybMat_t hybA,
                   cuDoubleComplex *A,
                   int lda)                             
```

This function converts a sparse matrix in HYB format (contained in the opaque structure) into matrix A in dense format. The dense matrix A is filled in with the values of the sparse matrix and with zeros elsewhere.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>handle</th>
<th>handle to the cuSPARSE library context.</th>
</tr>
</thead>
<tbody>
<tr>
<td>descrA</td>
<td>the descriptor of matrix A in Hyb format. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL.</td>
</tr>
<tr>
<td>hybA</td>
<td>the matrix A in HYB storage format.</td>
</tr>
<tr>
<td>lda</td>
<td>leading dimension of dense array A.</td>
</tr>
</tbody>
</table>
Output

| A | array of dimensions (lda, n) that is filled in with the values of the sparse matrix. |

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>the internally stored hyb format parameters are invalid.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

12.25. cusparse<t>nnz()

cusparseStatus_t 
cusparseSnnz(cusparseHandle_t handle, cusparseDirection_t dirA, int m, int n, const cusparseMatDescr_t descrA, const float *A, int lda, int *nnzPerRowColumn, int *nnzTotalDevHostPtr) 
cusparseStatus_t 
cusparseDnnz(cusparseHandle_t handle, cusparseDirection_t dirA, int m, int n, const cusparseMatDescr_t descrA, const double *A, int lda, int *nnzPerRowColumn, int *nnzTotalDevHostPtr) 
cusparseStatus_t 
cusparseCnnz(cusparseHandle_t handle, cusparseDirection_t dirA, int m, int n, const cusparseMatDescr_t descrA, const cuComplex *A, int lda, int *nnzPerRowColumn, int *nnzTotalDevHostPtr) 
cusparseStatus_t 
cusparseZnnz(cusparseHandle_t handle, cusparseDirection_t dirA, int m, int n, const cusparseMatDescr_t descrA, const cuDoubleComplex *A, int lda, int *nnzPerRowColumn, int *nnzTotalDevHostPtr) 

This function computes the number of nonzero elements per row or column and the total number of nonzero elements in a dense matrix.

This function requires no extra storage. It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

Input

| handle | handle to the cuSPARSE library context. |
| dirA | direction that specifies whether to count nonzero elements by CUSPARSE_DIRECTION_ROW or by CUSPARSE_DIRECTION_COLUMN. |
| m | number of rows of matrix A. |
| n | number of columns of matrix A. |
The descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>descrA</td>
<td>array of dimensions (lda, n).</td>
</tr>
<tr>
<td>lda</td>
<td>leading dimension of dense array A.</td>
</tr>
</tbody>
</table>

Array of dimensions (lda, n).

Array of size m or n containing the number of nonzero elements per row or column, respectively.

Total number of nonzero elements in device or host memory.

Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ALLOC_FAILED</td>
<td>the resources could not be allocated.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

12.26. cusparseCreateIdentityPermutation()

cusparseCreateIdentityPermutation(cusparseHandle_t handle,
int n,
int *p);

This function creates an identity map. The output parameter p represents such map by p = 0:1:(n-1).

This function is typically used with coosort, csrsort, cscsort, csr2csc_indexOnly.

Input

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
</tr>
<tr>
<td>n</td>
<td>host</td>
</tr>
</tbody>
</table>

Output

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
</table>
### Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (n&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>

### 12.27. cusparseXcoosort()

```c

cusparseStatus_t
cusparseXcoosort_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnz,
    const int *cooRows,
    const int *cooCols,
    size_t *pBufferSizeInBytes);

cusparseStatus_t
cusparseXcoosortByRow(cusparseHandle_t handle,
    int m,
    int n,
    int nnz,
    int *cooRows,
    int *cooCols,
    int *P,
    void *pBuffer);

cusparseStatus_t
cusparseXcoosortByColumn(cusparseHandle_t handle,
    int m,
    int n,
    int nnz,
    int *cooRows,
    int *cooCols,
    int *P,
    void *pBuffer);
```

This function sorts COO format. The sorting is in-place. Also the user can sort by row or sort by column.

**A** is an \( m \times n \) sparse matrix that is defined in COO storage format by the three arrays `cooVals`, `cooRows`, and `cooCols`.

There is no assumption for the base index of the matrix. `coosort` uses stable sort on signed integer, so the value of `cooRows` or `cooCols` can be negative.
This function `coosort()` requires buffer size returned by `coosort_bufferSizeExt()`. The address of `pBuffer` must be multiple of 128 bytes. If not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

The parameter `P` is both input and output. If the user wants to compute sorted `cooVal`, `P` must be set as `0:1:(nnz-1)` before `coosort()`, and after `coosort()`, new sorted value array satisfies `cooVal_sorted = cooVal(P)`.

Remark: the dimension `m` and `n` are not used. If the user does not know the value of `m` or `n`, just passes a value positive. This usually happens if the user only reads a COO array first and needs to decide the dimension `m` or `n` later.

Appendix D provides a simple example of `coosort()`.

**Input**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>host</td>
<td>number of rows of matrix <code>A</code>.</td>
</tr>
<tr>
<td>n</td>
<td>host</td>
<td>number of columns of matrix <code>A</code>.</td>
</tr>
<tr>
<td>nnz</td>
<td>host</td>
<td>number of nonzero elements of matrix <code>A</code>.</td>
</tr>
<tr>
<td>cooRows</td>
<td>device</td>
<td>integer array of <code>nnz</code> unsorted row indices of <code>A</code>.</td>
</tr>
<tr>
<td>cooCols</td>
<td>device</td>
<td>integer array of <code>nnz</code> unsorted column indices of <code>A</code>.</td>
</tr>
<tr>
<td>P</td>
<td>device</td>
<td>integer array of <code>nnz</code> unsorted map indices. To construct <code>cooVal</code>, the user has to set <code>P=0:1:(nnz-1)</code>.</td>
</tr>
<tr>
<td>pBuffer</td>
<td>device</td>
<td>buffer allocated by the user; the size is returned by <code>coosort_bufferSizeExt()</code>.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>cooRows</td>
<td>device</td>
<td>integer array of <code>nnz</code> sorted row indices of <code>A</code>.</td>
</tr>
<tr>
<td>cooCols</td>
<td>device</td>
<td>integer array of <code>nnz</code> sorted column indices of <code>A</code>.</td>
</tr>
<tr>
<td>P</td>
<td>device</td>
<td>integer array of <code>nnz</code> sorted map indices.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>host</td>
<td>number of bytes of the buffer.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (<code>n&lt;0</code>).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
12.28. cusparseXcsrsort()

```c
 cusparseStatus_t
 cusparseXcsrsort_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnz,
    const int *csrRowPtr,
    const int *csrColInd,
    size_t *pBufferSizeInBytes);
```

```c
 cusparseStatus_t
 cusparseXcsrsort(cusparseHandle_t handle,
    int m,
    int n,
    int nnz,
    const cusparseMatDescr_t descrA,
    const int *csrRowPtr,
    int *csrColInd,
    int *P,
    void *pBuffer);
```

This function sorts CSR format. The stable sorting is in-place.

The matrix type is regarded as `CUSPARSE_MATRIX_TYPE_GENERAL` implicitly. In other words, any symmetric property is ignored.

This function `csrsort()` requires buffer size returned by `csrsort_bufferSizeExt()`. The address of `pBuffer` must be multiple of 128 bytes. If not, `CUSPARSE_STATUS_INVALID_VALUE` is returned.

The parameter `P` is both input and output. If the user wants to compute sorted `csrVal`, `P` must be set as 0:1:(nnz-1) before `csrsort()`, and after `csrsort()`, new sorted value array satisfies `csrVal_sorted = csrVal(P)`. 
The general procedure is as follows:

```c
// A is a 3x3 sparse matrix, base-0
//   | 1 2 3 |
// A = | 4 5 6 |
//   | 7 8 9 |
const int m = 3;
const int n = 3;
const int nnz = 9;
csrRowPtr[m+1] = { 0, 3, 6, 9}; // on device
csrColInd[nnz] = { 2, 1, 0, 0, 2, 1, 2, 0}; // on device
csrVal[nnz] = { 3, 2, 1, 4, 6, 5, 8, 9, 7}; // on device
size_t pBufferSizeInBytes = 0;
void *pBuffer = NULL;
int *P = NULL;
// step 1: allocate buffer
cusparseXcsrsort_bufferSizeExt(handle, m, n, nnz, csrRowPtr, csrColInd, &pBufferSizeInBytes);
cudaMalloc( &pBuffer, sizeof(char)* pBufferSizeInBytes);
// step 2: setup permutation vector P to identity
cudaMalloc( (void**)&P, sizeof(int)*nnz);
cusparseCreateIdentityPermutation(handle, nnz, P);
// step 3: sort CSR format
cusparseXcsrsort(handle, m, n, nnz, descrA, csrRowPtr, csrColInd, P, pBuffer);
// step 4: gather sorted csrVal
cusparseDgthr(handle, nnz, csrVal, csrVal_sorted, P, CUSPARSE_INDEX_BASE_ZERO);
```

### Input

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>host</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>host</td>
<td>number of columns of matrix A.</td>
</tr>
<tr>
<td>nnz</td>
<td>host</td>
<td>number of nonzero elements of matrix A.</td>
</tr>
<tr>
<td>csrRowsPtr</td>
<td>device</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColInd</td>
<td>device</td>
<td>integer array of nnz unsorted column indices of A.</td>
</tr>
<tr>
<td>P</td>
<td>device</td>
<td>integer array of nnz unsorted map indices. To construct csrVal, the user has to set P=0:1:(nnz-1).</td>
</tr>
<tr>
<td>pBuffer</td>
<td>device</td>
<td>buffer allocated by the user; the size is returned by csrsort_bufferSizeExt().</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrColInd</td>
<td>device</td>
<td>integer array of nnz sorted column indices of A.</td>
</tr>
<tr>
<td>P</td>
<td>device</td>
<td>integer array of nnz sorted map indices.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>host</td>
<td>number of bytes of the buffer.</td>
</tr>
</tbody>
</table>

### Status Returned
<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n, nnz&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>

12.29. cusparseXcscsort()

cusparseStatus_t

cusparseXcscsort_bufferSizeExt(
  _cusparseHandle_t handle,
  int m,
  int n,
  int nnz,
  const int *cscColPtr,
  const int *cscRowInd,
  size_t *pBufferSizeInBytes);

cusparseStatus_t

cusparseXcscsort(cusparseHandle_t handle,
  int m,
  int n,
  int nnz,
  const cusparseMatDescr_t descrA,
  const int *cscColPtr,
  int *cscRowInd,
  int *P,
  void *pBuffer);

This function sorts CSC format. The stable sorting is in-place.

The matrix type is regarded as CUSPARSE_MATRIX_TYPE_GENERAL implicitly. In other words, any symmetric property is ignored.

This function cscsort() requires buffer size returned by cscsort_bufferSizeExt(). The address of pBuffer must be multiple of 128 bytes. If not, CUSPARSE_STATUS_INVALID_VALUE is returned.

The parameter P is both input and output. If the user wants to compute sorted cscVal, P must be set as 0:1:(nnz-1) before cscsort(), and after cscsort(), new sorted value array satisfies cscVal_sorted = cscVal(P).
The general procedure is as follows:

```c
// A is a 3x3 sparse matrix, base-0
//     | 1 2 |
// A = | 4 0 |
//     | 0 8 |
const int m = 3;
const int n = 2;
const int nnz = 4;
cscColPtr[n+1] = { 0, 2, 4}; // on device
cscRowInd[nnz] = { 1, 0, 2, 0}; // on device
cscVal[nnz] = { 4.0, 1.0, 8.0, 2.0 }; // on device
size_t pBufferSizeInBytes = 0;
void *pBuffer = NULL;
int *P = NULL;
// step 1: allocate buffer
cusparseXcscsort_bufferSizeExt(handle, m, n, nnz, cscColPtr, cscRowInd,
&pBufferSizeInBytes);
cudaMalloc( &pBuffer, sizeof(char)* pBufferSizeInBytes);

// step 2: setup permutation vector P to identity
cudaMalloc( (void**)&P, sizeof(int)*nnz);
cusparseCreateIdentityPermutation(handle, nnz, P);

// step 3: sort CSC format
cusparseXcscsort(handle, m, n, nnz, descrA, cscColPtr, cscRowInd, P, pBuffer);

// step 4: gather sorted cscVal
cusparseDgthr(handle, nnz, cscVal, cscVal_sorted, P, CUSPARSE_INDEX_BASE_ZERO);
```

**Input**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>host</td>
<td>number of rows of matrix A.</td>
</tr>
<tr>
<td>n</td>
<td>host</td>
<td>number of columns of matrix A.</td>
</tr>
<tr>
<td>nnz</td>
<td>host</td>
<td>number of nonzero elements of matrix A.</td>
</tr>
<tr>
<td>cscColPtr</td>
<td>device</td>
<td>integer array of n+1 elements that contains the start of every column and the end of the last column plus one.</td>
</tr>
<tr>
<td>cscRowInd</td>
<td>device</td>
<td>integer array of nnz unsorted row indices of A.</td>
</tr>
<tr>
<td>P</td>
<td>device</td>
<td>integer array of nnz unsorted map indices. To construct cscVal, the user has to set P=0:1:(nnz-1).</td>
</tr>
<tr>
<td>pBuffer</td>
<td>device</td>
<td>buffer allocated by the user; the size is returned by cscsort_bufferSizeExt().</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>cscRowInd</td>
<td>device</td>
<td>integer array of nnz sorted row indices of A.</td>
</tr>
<tr>
<td>P</td>
<td>device</td>
<td>integer array of nnz sorted map indices.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>host</td>
<td>number of bytes of the buffer.</td>
</tr>
</tbody>
</table>

**Status Returned**
<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n, nnz&lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.30. cusparseXcsru2csr()
This function transfers unsorted CSR format to CSR format, and vice versa. The operation is in-place.

This function is a wrapper of \texttt{csrsort} and \texttt{gthr}. The usecase is the following scenario.

If the user has a matrix $A$ of CSR format which is unsorted, and implements his own code (which can be CPU or GPU kernel) based on this special order (for example, diagonal first, then lower triangle, then upper triangle), and wants to convert it to CSR format when calling CUSPARSE library, and then convert it back when doing something else on his/her kernel. For example, suppose the user wants to solve a linear system $Ax=b$ by the following iterative scheme

$$x(k+1) = x(k) + L^{-1} * (b - A x(k))$$

The code heavily uses SpMv and triangular solve. Assume that the user has an in-house design of SpMV (Sparse Matrix-Vector multiplication) based on special order of $A$. However the user wants to use CUSPAPRSE library for triangular solver. Then the following code can work.

\begin{verbatim}
    do
        step 1: compute residual vector $r = b - A x(k)$ by
        step 2: $B := \text{sort}(A)$, and $L$ is lower triangular part of $B$
            (only sort $A$ once and keep the permutation vector)
            step 3: solve $z = L^{-1} * (b)$
            step 4: add correction $x(k+1) = x(k) + z$
            (use permutation vector to get back the unsorted CSR)
        until convergence
\end{verbatim}

The requirements of step 2 and step 5 are

1. In-place operation.
2. The permutation vector $P$ is hidden in an opaque structure.
3. No \texttt{cudaMalloc} inside the conversion routine. Instead, the user has to provide the buffer explicitly.
4. The conversion between unsorted CSR and sorted CSR may needs several times, but the function only generates the permutation vector $P$ once.
5. The function is based on \texttt{csrsort}, \texttt{gather} and \texttt{scatter} operations.

The operation is called \texttt{csru2csr}, which means unsorted CSR to sorted CSR. Also we provide the inverse operation, called \texttt{csr2csru}.

In order to keep the permutation vector invisible, we need an opaque structure called \texttt{csru2csrInfo}. Then two functions (\texttt{cusparseCreateCsr2csrInfo}, \texttt{cusparseDestroyCsr2csrInfo}) are used to initialize and to destroy the opaque structure.

\texttt{cusparse[S|D|C|Z]csru2csr_bufferSizeExt} returns the size of the buffer.

The permutation vector $P$ is also allcated inside \texttt{csru2csrInfo}. The lifetime of the permutation vector is the same as the lifetime of \texttt{csru2csrInfo}.
**cusparse[S|D|C|Z]csru2csr** performs forward transformation from unsorted CSR to sorted CSR. First call uses cxsrort to generate the permutation vector $\mathbf{P}$, and subsequent call uses $\mathbf{P}$ to do transformation.

**cusparse[S|D|C|Z]csr2csru** performs backward transformation from sorted CSR to unsorted CSR. $\mathbf{P}$ is used to get unsorted form back.

The following tables describe parameters of **csr2csru_bufferSizeExt** and **csr2csru**.

### Input

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>m</td>
<td>host</td>
<td>number of rows of matrix $\mathbf{A}$.</td>
</tr>
<tr>
<td>n</td>
<td>host</td>
<td>number of columns of matrix $\mathbf{A}$.</td>
</tr>
<tr>
<td>nnz</td>
<td>host</td>
<td>number of nonzero elements of matrix $\mathbf{A}$.</td>
</tr>
<tr>
<td>descrA</td>
<td>host</td>
<td>the descriptor of matrix $\mathbf{A}$. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>csrVal</td>
<td>device</td>
<td>&lt;type&gt; array of nnz unsorted nonzero elements of matrix $\mathbf{A}$.</td>
</tr>
<tr>
<td>csrRowsPtr</td>
<td>device</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColInd</td>
<td>device</td>
<td>integer array of nnz unsorted column indices of $\mathbf{A}$.</td>
</tr>
<tr>
<td>info</td>
<td>host</td>
<td>opaque structure initialized using cusparseCreateCsru2csrInfo().</td>
</tr>
<tr>
<td>pBuffer</td>
<td>device</td>
<td>buffer allocated by the user; the size is returned by csru2csr_bufferSizeExt().</td>
</tr>
</tbody>
</table>

### Output

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>csrVal</td>
<td>device</td>
<td>&lt;type&gt; array of nnz sorted nonzero elements of matrix $\mathbf{A}$.</td>
</tr>
<tr>
<td>csrColInd</td>
<td>device</td>
<td>integer array of nnz sorted column indices of $\mathbf{A}$.</td>
</tr>
<tr>
<td>pBuffer</td>
<td>host</td>
<td>number of bytes of the buffer.</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed ($m,n,nnz&lt;0$).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
</tbody>
</table>
CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED the matrix type is not supported.
12.31. `cusparseXpruneDense2csr()`

cusparseStatus_t
cusparseHpruneDense2csr_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    const __half *A,
    int lda,
    const __half *threshold,
    const cusparseMatDescr_t descrC,
    const __half *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    size_t *pBufferSizeInBytes);

cusparseStatus_t
cusparseSpruneDense2csr_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    const float *A,
    int lda,
    const float *threshold,
    const cusparseMatDescr_t descrC,
    const float *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    size_t *pBufferSizeInBytes);

cusparseStatus_t
cusparseDpruneDense2csr_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    const double *A,
    int lda,
    const double *threshold,
    const cusparseMatDescr_t descrC,
    const double *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    size_t *pBufferSizeInBytes);

cusparseStatus_t
cusparseHpruneDense2csrNnz(
    cusparseHandle_t handle,
    int m,
    int n,
    const __half *A,
    int lda,
    const __half *threshold,
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr,
    void *pBuffer);

cusparseStatus_t
cusparseSpruneDense2csrNnz(
    cusparseHandle_t handle,
    int m,
    int n,
    const float *A,
    int lda,
    const float *threshold,
    const cusparseMatDescr_t descrC,
    float *csrValC,
    int *nnzTotalDevHostPtr,
    void *pBuffer);

cusparseStatus_t
cusparseDpruneDense2csrNnz(
    cusparseHandle_t handle,
    int m,
    int n,
    const double *A,
    int lda,
    const double *threshold,
    const cusparseMatDescr_t descrC,
    double *csrValC,
    int *nnzTotalDevHostPtr,
    void *pBuffer);
This function prunes a dense matrix to a sparse matrix with CSR format.

Given a dense matrix $A$ and a non-negative value $\text{threshold}$, the function returns a sparse matrix $C$, defined by

$$C(i,j) = A(i,j) \quad \text{if} \quad |A(i,j)| > \text{threshold}$$

The implementation adopts a two-step approach to do the conversion. First, the user allocates $\text{csrRowPtrC}$ of $m+1$ elements and uses function $\text{pruneDense2csrNnz()}$ to determine the number of nonzeros columns per row. Second, the user gathers $\text{nnzC}$ (number of nonzeros of matrix $C$) from either ($\text{nnzC}=\text{nnzTotalDevHostPtr}$) or ($\text{nnzC}=\text{csrRowPtrC}[m]-\text{csrRowPtrC}[0]$) and allocates $\text{csrValC}$ of $\text{nnzC}$ elements and $\text{csrColIndC}$ of $\text{nnzC}$ integers. Finally function $\text{pruneDense2csr()}$ is called to complete the conversion.

The user must obtain the size of the buffer required by $\text{pruneDense2csr()}$ by calling $\text{pruneDense2csr_bufferSizeExt()}$, allocate the buffer, and pass the buffer pointer to $\text{pruneDense2csr()}$.

Appendix E.1 provides a simple example of $\text{pruneDense2csr()}$.

**Input**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>$m$</td>
<td>host</td>
<td>number of rows of matrix $A$.</td>
</tr>
<tr>
<td>$n$</td>
<td>host</td>
<td>number of columns of matrix $A$.</td>
</tr>
<tr>
<td>$A$</td>
<td>device</td>
<td>array of dimension $(\text{lda}, n)$.</td>
</tr>
<tr>
<td>$\text{lda}$</td>
<td>device</td>
<td>leading dimension of $A$. It must be at least max(1, $m$).</td>
</tr>
<tr>
<td>$\text{threshold}$</td>
<td>host or device</td>
<td>a value to drop the entries of $A$. $\text{threshold}$ can point to a device memory or host memory.</td>
</tr>
<tr>
<td>$\text{descrC}$</td>
<td>host</td>
<td>the descriptor of matrix $C$. The supported matrix type is \text{CUSPARSE_MATRIX_TYPE_GENERAL}. Also, the supported index bases are \text{CUSPARSE_INDEX_BASE_ZERO} and \text{CUSPARSE_INDEX_BASE_ONE}.</td>
</tr>
<tr>
<td>$\text{pBuffer}$</td>
<td>device</td>
<td>buffer allocated by the user; the size is returned by $\text{pruneDense2csr_bufferSizeExt()}$.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\text{nnzTotalDevHostPtr}$</td>
<td>device or host</td>
<td>total number of nonzero of matrix $C$. $\text{nnzTotalDevHostPtr}$ can point to a device memory or host memory.</td>
</tr>
<tr>
<td>$\text{csrValC}$</td>
<td>device</td>
<td>$&lt;$type$&gt;$ array of $\text{nnzC}$ nonzero elements of matrix $C$.</td>
</tr>
<tr>
<td>$\text{csrRowsPtrC}$</td>
<td>device</td>
<td>integer array of $m+1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>$\text{csrColIndC}$</td>
<td>device</td>
<td>integer array of $\text{nnzC}$ column indices of $C$.</td>
</tr>
<tr>
<td>$\text{pBufferSizeInBytes}$</td>
<td>host</td>
<td>number of bytes of the buffer.</td>
</tr>
</tbody>
</table>
### Status Returned

<table>
<thead>
<tr>
<th>Status Returned</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n &lt; 0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.32. `cusparseXpruneCsr2csr()`

```c
cusparseStatus_t
cusparseXpruneCsr2csr_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const __half *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const __half *threshold,
    const cusparseMatDescr_t descrC,
    const __half *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    size_t *pBufferSizeInBytes);
```

```c
cusparseStatus_t
cusparseXpruneCsr2csr_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const float *threshold,
    const cusparseMatDescr_t descrC,
    const float *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    size_t *pBufferSizeInBytes);
```

```c
cusparseStatus_t
cusparseXpruneCsr2csr_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const double *threshold,
    const cusparseMatDescr_t descrC,
    const double *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    size_t *pBufferSizeInBytes);
```

```c
cusparseStatus_t
cusparseXpruneCsr2csrNnz(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const __half *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const __half *threshold,
    const cusparseMatDescr_t descrC,
    __half *csrValC,
    const int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    void *pBuffer);
```

```c
cusparseStatus_t
cusparseXpruneCsr2csrNnz(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const float *threshold,
    const cusparseMatDescr_t descrC,
    float *csrValC,
    const int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    void *pBuffer);
```

```c
cusparseStatus_t
cusparseXpruneCsr2csrNnz(
    cusparseHandle_t handle,
    int m,
    int n,
    int nnzA,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const double *threshold,
    const cusparseMatDescr_t descrC,
    double *csrValC,
    const int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    void *pBuffer);
```
This function prunes a sparse matrix to a sparse matrix with CSR format.

Given a sparse matrix \( A \) and a non-negative value \( \text{threshold} \), the function returns a sparse matrix \( C \), defined by

\[
C(i,j) = A(i,j) \quad \text{if} \quad |A(i,j)| > \text{threshold}
\]

The implementation adopts a two-step approach to do the conversion. First, the user allocates \( \text{csrRowPtrC} \) of \( m+1 \) elements and uses function \( \text{pruneCsr2csrNnz()} \) to determine the number of nonzeros columns per row. Second, the user gathers \( \text{nnzC} \) (number of nonzeros of matrix \( C \)) from either \( (\text{nnzC} = \text{nnzTotalDevHostPtr}) \) or \( (\text{nnzC} = \text{csrRowPtrC}[m] - \text{csrRowPtrC}[0]) \) and allocates \( \text{csrValC} \) of \( \text{nnzC} \) elements and \( \text{csrColIndC} \) of \( \text{nnzC} \) integers. Finally function \( \text{pruneCsr2csr()} \) is called to complete the conversion.

The user must obtain the size of the buffer required by \( \text{pruneCsr2csr()} \) by calling \( \text{pruneCsr2csr_bufferSizeExt()} \), allocate the buffer, and pass the buffer pointer to \( \text{pruneCsr2csr()} \).

Appendix E.2 provides a simple example of \( \text{pruneCsr2csr()} \).

**Input**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>( m )</td>
<td>host</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>( n )</td>
<td>host</td>
<td>number of columns of matrix ( A ).</td>
</tr>
<tr>
<td>( \text{nnzA} )</td>
<td>host</td>
<td>number of nonzeros of matrix ( A ).</td>
</tr>
<tr>
<td>descrA</td>
<td>host</td>
<td>the descriptor of matrix ( A ). The supported matrix type is ( \text{CUSPARSE_MATRIX_TYPE_GENERAL} ). Also, the supported index bases are ( \text{CUSPARSE_INDEX_BASE_ZERO} ) and ( \text{CUSPARSE_INDEX_BASE_ONE} ).</td>
</tr>
<tr>
<td>( \text{csrValA} )</td>
<td>device</td>
<td>(&lt;\text{type}&gt;) array of ( \text{nnzA} ) nonzero elements of matrix ( A ).</td>
</tr>
<tr>
<td>( \text{csrRowsPtrA} )</td>
<td>device</td>
<td>integer array of ( m+1 ) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>( \text{csrColIndA} )</td>
<td>device</td>
<td>integer array of ( \text{nnzA} ) column indices of ( A ).</td>
</tr>
<tr>
<td>( \text{threshold} )</td>
<td>host or device</td>
<td>a value to drop the entries of ( A ). ( \text{threshold} ) can point to a device memory or host memory.</td>
</tr>
<tr>
<td>descrC</td>
<td>host</td>
<td>the descriptor of matrix ( C ). The supported matrix type is ( \text{CUSPARSE_MATRIX_TYPE_GENERAL} ). Also, the supported index bases are ( \text{CUSPARSE_INDEX_BASE_ZERO} ) and ( \text{CUSPARSE_INDEX_BASE_ONE} ).</td>
</tr>
<tr>
<td>pBuffer</td>
<td>device</td>
<td>buffer allocated by the user; the size is returned by ( \text{pruneCsr2csr_bufferSizeExt()} ).</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Variable</td>
<td>Type</td>
<td>Description</td>
</tr>
<tr>
<td>------------------------</td>
<td>----------</td>
<td>------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>nnzTotalDevHostPtr</td>
<td>device or host</td>
<td>total number of nonzero of matrix C. nnzTotalDevHostPtr can point to a device memory or host memory.</td>
</tr>
<tr>
<td>csrValC</td>
<td>device</td>
<td>&lt;type&gt; array of nnzC nonzero elements of matrix C.</td>
</tr>
<tr>
<td>csrRowsPtrC</td>
<td>device</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndC</td>
<td>device</td>
<td>integer array of nnzC column indices of C.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>host</td>
<td>number of bytes of the buffer.</td>
</tr>
</tbody>
</table>

**Status Returned**

<table>
<thead>
<tr>
<th>Status</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n &lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.33. `cusparseXpruneDense2csrPercentage()`

```c
cusparseStatus_t
cusparseHpruneDense2csrByPercentage_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    const __half *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const __half *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    size_t *pBufferSizeInBytes);

cusparseStatus_t
cusparseSpruneDense2csrByPercentage_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    const float *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const float *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    size_t *pBufferSizeInBytes);

cusparseStatus_t
cusparseDpruneDense2csrByPercentage_bufferSizeExt(
    cusparseHandle_t handle,
    int m,
    int n,
    const double *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const double *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    size_t *pBufferSizeInBytes);

cusparseStatus_t
cusparseHpruneDense2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const __half *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    void *pBuffer);

cusparseStatus_t
cusparseSpruneDense2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const float *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    void *pBuffer);

cusparseStatus_t
cusparseDpruneDense2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const double *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    void *pBuffer);
```

```c
cusparseStatus_t
cusparseHpruneDense2csrByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const __half *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const __half *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    __half *csrValC);

cusparseStatus_t
cusparseSpruneDense2csrByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const float *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const float *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    float *csrValC);

cusparseStatus_t
cusparseDpruneDense2csrByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const double *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const double *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    double *csrValC);
```

```c
cusparseStatus_t
cusparseHpruneDense2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const __half *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    __half *pBuffer);

cusparseStatus_t
cusparseSpruneDense2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const float *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    float *pBuffer);

cusparseStatus_t
cusparseDpruneDense2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m,
    int n,
    const double *A,
    int lda,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    double *pBuffer);
```
This function prunes a dense matrix to a sparse matrix by percentage.

Given a dense matrix \( A \) and a non-negative value \( \text{percentage} \), the function computes sparse matrix \( C \) by the following three steps:

Step 1: sort absolute value of \( A \) in ascending order.

\[
\text{key} := \text{sort( } |A| \text{ )}
\]

Step 2: choose threshold by the parameter \( \text{percentage} \)

\[
pos = \text{ceil}(m \times n \times (\text{percentage}/100)) - 1 \\
pos = \min(pos, m \times n - 1) \\
pos = \max(pos, 0) \\
\text{threshold} = \text{key}[pos]
\]

Step 3: call \( \text{pruneDense2csr() } \) by with the parameter \( \text{threshold} \).

The implementation adopts a two-step approach to do the conversion.
First, the user allocates \( \text{csrRowPtrC} \) of \( m+1 \) elements and uses function \( \text{pruneDense2csrNnzByPercentage()} \) to determine the number of nonzeros columns per row. Second, the user gathers \( \text{nnzC} \) (number of nonzeros of matrix \( C \)) from either \( \text{(nnzC=}^*\text{nnzTotalDevHostPtr)} \) or \( \text{(nnzC=}^\text{csrRowPtrC[m]}-\text{csrRowPtrC[0]}) \) and allocates \( \text{csrValC} \) of \( \text{nnzC} \) elements and \( \text{csrColIndC} \) of \( \text{nnzC} \) integers. Finally function \( \text{pruneDense2csrByPercentage()} \) is called to complete the conversion.

The user must obtain the size of the buffer required by \( \text{pruneDense2csrByPercentage()} \) by calling \( \text{pruneDense2csrByPercentage_bufferSizeExt()} \), allocate the buffer, and pass the buffer pointer to \( \text{pruneDense2csrByPercentage()} \).

Remark 1: the value of \( \text{percentage} \) must be not greater than 100. Otherwise, \( \text{CUSPARSE_STATUS_INVALID_VALUE} \) is returned.

Remark 2: the zeros of \( A \) are not ignored. All entries are sorted, including zeros. This is different from \( \text{pruneCsr2csrByPercentage()} \)

Appendix E.3 provides a simple example of \( \text{pruneDense2csrByPercentage()} \).

**Input**

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>( m )</td>
<td>host</td>
<td>number of rows of matrix ( A ).</td>
</tr>
<tr>
<td>( n )</td>
<td>host</td>
<td>number of columns of matrix ( A ).</td>
</tr>
<tr>
<td>( A )</td>
<td>device</td>
<td>array of dimension (lda, n).</td>
</tr>
<tr>
<td>( \text{lda} )</td>
<td>device</td>
<td>leading dimension of ( A ). It must be at least max(1, m).</td>
</tr>
<tr>
<td>( \text{percentage} )</td>
<td>host</td>
<td>percentage &lt;=100 and percentage &gt;= 0</td>
</tr>
<tr>
<td>( \text{descrC} )</td>
<td>host</td>
<td>the descriptor of matrix ( C ). The supported matrix type is ( \text{CUSPARSE_MATRIX_TYPE_GENERAL} ), Also, the supported index bases are ( \text{CUSPARSE_INDEX_BASE_ZERO} ) and ( \text{CUSPARSE_INDEX_BASE_ONE} ).</td>
</tr>
</tbody>
</table>
pBuffer | device | buffer allocated by the user; the size is returned by pruneDense2csrByPercentage_bufferSizeExt().

### Output

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnzTotalDevHostPtr</td>
<td>device or host</td>
<td>total number of nonzero of matrix C.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>nnzTotalDevHostPtr can point to a device memory or host memory.</td>
</tr>
<tr>
<td>csrValC</td>
<td>device</td>
<td>&lt;type&gt; array of nnzC nonzero elements of matrix C.</td>
</tr>
<tr>
<td>csrRowsPtrC</td>
<td>device</td>
<td>integer array of m+1 elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndC</td>
<td>device</td>
<td>integer array of nnzC column indices of C.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>host</td>
<td>number of bytes of the buffer.</td>
</tr>
</tbody>
</table>

### Status Returned

<table>
<thead>
<tr>
<th>Status</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_STATUS_SUCCESS</td>
<td>the operation completed successfully.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_NOT_INITIALIZED</td>
<td>the library was not initialized.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INVALID_VALUE</td>
<td>invalid parameters were passed (m, n &lt;0).</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_ARCH_MISMATCH</td>
<td>the device does not support double precision.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_EXECUTION_FAILED</td>
<td>the function failed to launch on the GPU.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_INTERNAL_ERROR</td>
<td>an internal operation failed.</td>
</tr>
<tr>
<td>CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED</td>
<td>the matrix type is not supported.</td>
</tr>
</tbody>
</table>
12.34. `cusparseXpruneCsr2csrByPercentage()`

```c
 cusparseStatus_t
cusparseXpruneCsr2csrBufferSizeExt(
    cusparseHandle_t handle,
    int m, int n, int nnzA,
    const cusparseMatDescr_t descrA,
    const __half *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const __half *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    size_t *pBufferSizeInBytes);
```

```c
 cusparseStatus_t
cusparseXpruneCsr2csrBufferSizeExt(
    cusparseHandle_t handle,
    int m, int n, int nnzA,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const float *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    size_t *pBufferSizeInBytes);
```

```c
 cusparseStatus_t
cusparseXpruneCsr2csrBufferSizeExt(
    cusparseHandle_t handle,
    int m, int n, int nnzA,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    const double *csrValC,
    const int *csrRowPtrC,
    const int *csrColIndC,
    pruneInfo_t info,
    size_t *pBufferSizeInBytes);
```

```c
 cusparseStatus_t
cusparseXpruneCsr2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m, int n, int nnzA,
    const cusparseMatDescr_t descrA,
    const __half *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    void *pBuffer);
```

```c
 cusparseStatus_t
cusparseXpruneCsr2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m, int n, int nnzA,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    void *pBuffer);
```

```c
 cusparseStatus_t
cusparseXpruneCsr2csrNnzByPercentage(
    cusparseHandle_t handle,
    int m, int n, int nnzA,
    const cusparseMatDescr_t descrA,
    const double *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    float percentage, /* between 0 to 100 */
    const cusparseMatDescr_t descrC,
    int *csrRowPtrC,
    int *nnzTotalDevHostPtr, /* can be on host or device */
    pruneInfo_t info,
    void *pBuffer);
```
This function prunes a sparse matrix to a sparse matrix by percentage.

Given a sparse matrix $A$ and a non-negative value $\text{percentage}$, the function computes sparse matrix $C$ by the following three steps:

Step 1: sort absolute value of $A$ in ascending order.

\[
\text{key} := \text{sort} (|\text{csrValA}|) 
\]

Step 2: choose threshold by the parameter $\text{percentage}$

\[
\begin{align*}
\text{pos} &= \text{ceil}(\text{nnzA}*(\text{percentage}/100)) - 1 \\
\text{pos} &= \min(\text{pos}, \text{nnzA}-1) \\
\text{pos} &= \max(\text{pos}, 0) \\
\text{threshold} &= \text{key}[\text{pos}] 
\end{align*}
\]

Step 3: call $\text{pruneCsr2csr}()$ by with the parameter $\text{threshold}$.

The implementation adopts a two-step approach to do the conversion. First, the user allocates $\text{csrRowPtrC}$ of $m+1$ elements and uses function $\text{pruneCsr2csrNnzByPercentage}()$ to determine the number of nonzeros columns per row. Second, the user gathers $\text{nnzC}$ (number of nonzeros of matrix $C$) from either ($\text{nnzC} = \text{nnzTotalDevHostPtr}$) or ($\text{nnzC} = \text{csrRowPtrC}[m] - \text{csrRowPtrC}[0]$) and allocates $\text{csrValC}$ of $\text{nnzC}$ elements and $\text{csrColIndC}$ of $\text{nnzC}$ integers. Finally function $\text{pruneCsr2csrByPercentage}()$ is called to complete the conversion.

The user must obtain the size of the buffer required by $\text{pruneCsr2csrByPercentage}()$ by calling $\text{pruneCsr2csrByPercentage_bufferSizeExt}()$, allocate the buffer, and pass the buffer pointer to $\text{pruneCsr2csrByPercentage}()$.

Remark 1: the value of $\text{percentage}$ must be not greater than 100. Otherwise, $\text{CUSPARSE_STATUS_INVALID_VALUE}$ is returned.

Appendix E.4 provides a simple example of $\text{pruneCsr2csrByPercentage}()$.

Input

<table>
<thead>
<tr>
<th>parameter</th>
<th>device or host</th>
<th>description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>host</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>$m$</td>
<td>host</td>
<td>number of rows of matrix $A$.</td>
</tr>
<tr>
<td>$n$</td>
<td>host</td>
<td>number of columns of matrix $A$.</td>
</tr>
<tr>
<td>$\text{nnzA}$</td>
<td>host</td>
<td>number of nonzeros of matrix $A$.</td>
</tr>
<tr>
<td>$\text{descrA}$</td>
<td>host</td>
<td>the descriptor of matrix $A$. The supported matrix type is $\text{CUSPARSE_MATRIX_TYPE_GENERAL}$. Also, the supported index bases are $\text{CUSPARSE_INDEX_BASE_ZERO}$ and $\text{CUSPARSE_INDEX_BASE_ONE}$.</td>
</tr>
<tr>
<td>$\text{csrValA}$</td>
<td>device</td>
<td>$&lt;\text{type}&gt;$ array of $\text{nnzA}$ nonzero elements of matrix $A$.</td>
</tr>
<tr>
<td>$\text{csrRowsPtrA}$</td>
<td>device</td>
<td>integer array of $m+1$ elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>$\text{csrColIndA}$</td>
<td>device</td>
<td>integer array of $\text{nnzA}$ column indices of $A$.</td>
</tr>
<tr>
<td>$\text{percentage}$</td>
<td>host</td>
<td>$\text{percentage} &lt;= 100$ and $\text{percentage} &gt;= 0$</td>
</tr>
</tbody>
</table>
The descriptor of matrix C. The supported matrix type is \texttt{CUSPARSE_MATRIX_TYPE_GENERAL}, Also, the supported index bases are \texttt{CUSPARSE_INDEX_BASE_ZERO} and \texttt{CUSPARSE_INDEX_BASE_ONE}.

Buffer allocated by the user; the size is returned by \texttt{pruneCsr2csrByPercentage_bufferSizeExt}.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Device or Host</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>nnzTotalDevHostPtr</td>
<td>Device or Host</td>
<td>Total number of nonzero of matrix C. (\text{nnzTotalDevHostPtr}) can point to a device memory or host memory.</td>
</tr>
<tr>
<td>csrValC</td>
<td>Device</td>
<td>(&lt;\text{type}&gt;) array of (\text{nnzC}) nonzero elements of matrix C.</td>
</tr>
<tr>
<td>csrRowsPtrC</td>
<td>Device</td>
<td>Integer array of (m+1) elements that contains the start of every row and the end of the last row plus one.</td>
</tr>
<tr>
<td>csrColIndC</td>
<td>Device</td>
<td>Integer array of (\text{nnzC}) column indices of C.</td>
</tr>
<tr>
<td>pBufferSizeInBytes</td>
<td>Host</td>
<td>Number of bytes of the buffer.</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Status Returned</th>
</tr>
</thead>
<tbody>
<tr>
<td>\texttt{CUSPARSE_STATUS_SUCCESS}</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_NOT_INITIALIZED}</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_INVALID_VALUE}</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_ARCH_MISMATCH}</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_EXECUTION_FAILED}</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_INTERNAL_ERROR}</td>
</tr>
<tr>
<td>\texttt{CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED}</td>
</tr>
</tbody>
</table>
### 12.35. cusparse<t>nnz_compress()

This function is the step one to convert from csr format to compressed csr format.

Given a sparse matrix $A$ and a non-negative value threshold, the function returns $\text{nnzPerRow}$(the number of nonzeros columns per row) and $\text{nnzC}$(the total number of nonzeros) of a sparse matrix $C$, defined by

$$ C(i,j) = A(i,j) \text{ if } |A(i,j)| > \text{threshold} $$

A key assumption for the cuComplex and cuDoubleComplex case is that this tolerance is given as the real part. For example $\text{tol} = 1e-8 + 0*i$ and we extract cureal, that is the $x$ component of this struct.

**Input**

| handle | handle to the cuSPARSE library context. |
| m      | number of rows of matrix $A$.          |
descrA | the descriptor of matrix A. The supported matrix type is CUSPARSE_MATRIX_TYPE_GENERAL. Also, the supported index bases are CUSPARSE_INDEX_BASE_ZERO and CUSPARSE_INDEX_BASE_ONE.

csrValA | csr noncompressed values array

csrRowPtrA | the corresponding input noncompressed row pointer.

tol | non-negative tolerance to determine if a number less than or equal to it.

Output

| nnzPerRow | this array contains the number of elements whose absolute values are greater than tol per row.
| nnzC | host/device pointer of the total number of elements whose absolute values are greater than tol.

Status Returned

| CUSPARSE_STATUS_SUCCESS | the operation completed successfully.
| CUSPARSE_STATUS_NOT_INITIALIZED | the library was not initialized.
| CUSPARSE_STATUS_ALLOC_FAILED | the resources could not be allocated.
| CUSPARSE_STATUS_INVALID_VALUE | invalid parameters were passed (m, n<0).
| CUSPARSE_STATUS_ARCH_MISMATCH | the device does not support double precision.
| CUSPARSE_STATUS_EXECUTION_FAILED | the function failed to launch on the GPU.
| CUSPARSE_STATUS_INTERNAL_ERROR | an internal operation failed.
| CUSPARSE_STATUS_MATRIX_TYPE_NOT_SUPPORTED | the matrix type is not supported.
Chapter 13.
CUSPARSE GENERIC API REFERENCE

Starting with CUDA 10.1, the cuSPARSE Library provides a new set of API to compute sparse matrix-matrix multiplication (SpMM), in addition to the existing legacy API. All following functions and types operate on matrix problems of the form:

\[ C = \alpha AB + \beta C \]

where A is sparse matrix, B and C are dense matrices. They are considered a preview feature i.e. the data structures and APIs for them are subject to change and may not be compatible with future releases.

13.1. Generic Types Reference

The cuSPARSE generic type references are described in this section.

13.1.1. cudaDataType_t

The section describes types shared by multiple CUDA Libraries and defined in the header file `library_types.h`. The `cudaDataType_t` type is an enumerant to specify the data precision. It is used when the data reference does not carry the type itself (e.g. `void *`) For example, it is used in the routine `cusparseSpMM()`.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUDA_R_16F</td>
<td>the data type is 16-bit floating-point (_half)</td>
</tr>
<tr>
<td>CUDA_C_16F</td>
<td>the data type is 16-bit complex floating-point (_half2)</td>
</tr>
<tr>
<td>CUDA_R_32F</td>
<td>the data type is 32-bit floating-point (float)</td>
</tr>
<tr>
<td>CUDA_C_32F</td>
<td>the data type is 32-bit complex floating-point (cuComplex)</td>
</tr>
<tr>
<td>CUDA_R_64F</td>
<td>the data type is 64-bit floating-point (double)</td>
</tr>
<tr>
<td>CUDA_C_64F</td>
<td>the data type is 64-bit complex floating-point (cuDoubleComplex)</td>
</tr>
<tr>
<td>Value</td>
<td>Meaning</td>
</tr>
<tr>
<td>---------------</td>
<td>-------------------------------------------------------------------------</td>
</tr>
<tr>
<td>CUDA_R_8I</td>
<td>the data type is 8-bit signed integer (int8_t)</td>
</tr>
</tbody>
</table>

### 13.1.2. cusparseFormat_t

This type indicates the format of the sparse matrix. Currently, only COO format is supported.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_FORMAT_COO</td>
<td>the matrix is stored in Coordinate (COO) format.</td>
</tr>
</tbody>
</table>

### 13.1.3. cusparseOrder_t

This type indicates the memory layout of a dense matrix. Currently, only column-major layout is supported.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_ORDER_COL</td>
<td>the matrix is stored in column-major.</td>
</tr>
</tbody>
</table>

### 13.1.4. cusparseSpMMAlg_t

This type indicates the algorithm parameter used for `cusparseSpMM()` and `cusparseSpMM_bufferSize()` sparse matrix-matrix multiplication.

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_COOMM_ALG1</td>
<td>COO matrix-matrix multiplication specialized for small matrices. The algorithm provides non-deterministic results (different execution could provide slightly different output). It supports batched matrix computation</td>
</tr>
<tr>
<td>CUSPARSE_COOMM_ALG2</td>
<td>COO matrix-matrix multiplication specialized for small matrices. The algorithm provides deterministic results (generate the same bit-wise output). It supports only op(A) CUSPARSE_OPERATION_NON_TRANSPOSE</td>
</tr>
<tr>
<td>CUSPARSE_COOMM_ALG3</td>
<td>COO matrix-matrix multiplication specialized for large matrices. The algorithm provides non-deterministic results (different execution could provide slightly different output). It does not support cuDoubleComplex data type</td>
</tr>
</tbody>
</table>

### 13.1.5. cusparseIndexType_t

This type indicates the index type for representing the sparse matrix indices.
### Value

<table>
<thead>
<tr>
<th>Value</th>
<th>Meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>CUSPARSE_INDEX_16U</td>
<td>16-bit unsigned integer for both row and column COO indices [1, 65535]</td>
</tr>
<tr>
<td>CUSPARSE_INDEX_32I</td>
<td>32-bit signed integer for both row and column COO indices [1, 2147483647]</td>
</tr>
</tbody>
</table>

### 13.2. Generic Sparse API helper functions

The cuSPARSE helper functions for sparse matrix descriptor are described in this section.

#### 13.2.1. cusparseCreateCoo()

```c
cusparseStatus_t cusparseCreateCoo(cusparseSpMatDescr_t* spMatDescr,
                                   int rows,
                                   int cols,
                                   int nnz,
                                   void* cooRowInd,
                                   void* cooColInd,
                                   void* cooValues,
                                   cusparseIndexType_t cooIdxType,
                                   cusparseIndexBase_t idxBase,
                                   cudaDataType valueType)
```

This function initializes the sparse matrix descriptor `spMatDescr` in the COO format.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>rows</td>
<td>number of rows of the sparse matrix.</td>
</tr>
<tr>
<td>cols</td>
<td>number of cols of sparse matrix.</td>
</tr>
<tr>
<td>nnz</td>
<td>number of nonzero elements of the sparse matrix.</td>
</tr>
<tr>
<td>cooRowInd</td>
<td>integer array of <code>nnz</code> uncompressed row indices.</td>
</tr>
<tr>
<td>cooColInd</td>
<td>integer array of <code>nnz</code> uncompressed column indices.</td>
</tr>
<tr>
<td>cooValues</td>
<td>value array of size <code>nnz</code> of the sparse matrix.</td>
</tr>
<tr>
<td>cooIdxType</td>
<td>Index base CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>idxBase</td>
<td>Index data type CUSPARSE_INDEX_16U or CUSPARSE_INDEX_32I.</td>
</tr>
<tr>
<td>valueType</td>
<td>value type.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>spMatDescr</td>
<td>sparse matrix descriptor.</td>
</tr>
</tbody>
</table>
13.2.2. cusparseDestroySpMat()

cusparseStatus_t
 cusparseDestroySpMat(cusparseSpMatDescr_t spMatDescr)

This function releases the memory allocated for the sparse matrix descriptor spMatDescr.

Input

| spMatDescr | sparse matrix descriptor. |

13.2.3. cusparseSpMatGetNumBatches()

cusparseStatus_t
 cusparseSpMatGetNumBatches(const cusparseSpMatDescr_t spMatDescr, int* batchCount)

This function returns the number of batches of the sparse matrix descriptor spMatDescr.

Input

| spMatDescr | sparse matrix descriptor. |

Output

| batchCount | number of batches of the sparse matrix. |

13.2.4. cusparseCooGet()

cusparseStatus_t
 cusparseCooGet(const cusparseSpMatDescr_t spMatDescr, int* rows, int* cols, int* nnz, void** cooRowInd, void** cooColInd, void** cooValues, cusparseIndexType_t* idxType, cusparseIndexBase_t* idxBase, cudaDataType* valueType)

This function returns the spMatDescr fields of the sparse matrix descriptor.

Input

| spMatDescr | sparse matrix descriptor. |

Output

<p>| rows | number of rows of the sparse matrix. |
| cols | number of cols of sparse matrix. |
| nnz | number of nonzero elements of the sparse matrix. |</p>
<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>cooRowInd</td>
<td>integer array of nnz uncompressed row indices.</td>
</tr>
<tr>
<td>cooColInd</td>
<td>integer array of nnz uncompressed column indices.</td>
</tr>
<tr>
<td>cooIdxType</td>
<td>Index base CUSPARSE_INDEX_BASE_ZERO or CUSPARSE_INDEX_BASE_ONE.</td>
</tr>
<tr>
<td>idxBase</td>
<td>Index data type CUSPARSE_INDEX_16U or CUSPARSE_INDEX_32I.</td>
</tr>
</tbody>
</table>

### 13.2.5. cusparseCooGet()

```c
cusparseStatus_t
cusparseSpMatGetFormat(const cusparseSpMatDescr_t spMatDescr, cusparseFormat_t* format)
```

This function returns the format of the sparse matrix descriptor `spMatDescr`.

**Input**

- **spMatDescr**: sparse matrix descriptor.

**Output**

- **format**: storage format of the sparse matrix.

### 13.2.6. cusparseSpMatGetIndexBase()

```c
cusparseStatus_t
cusparseSpMatGetIndexBase(const cusparseSpMatDescr_t spMatDescr, cusparseIndexBase_t* idxBase)
```

This function returns the index base of the sparse matrix descriptor `spMatDescr`.

**Input**

- **spMatDescr**: sparse matrix descriptor.

**Output**

- **idxBase**: index base of the sparse matrix.

### 13.2.7. cusparseSpMatSetNumBatches()

```c
cusparseStatus_t
cusparseSpMatSetNumBatches(cusparseSpMatDescr_t spMatDescr, int batchCount)
```

This function sets the number of batches of the sparse matrix descriptor `spMatDescr`.

**Input**

- **spMatDescr**: sparse matrix descriptor.
- **batchCount**: number of batches for the sparse matrix descriptor.
13.3. Generic Dense API helper functions

The cuSPARSE helper functions for dense matrix descriptor are described in this section.

13.3.1. cusparseCreateDnMat()

```c
 cusparseStatus_t
 cusparseCreateDnMat(cusparseDnMatDescr_t* dnMatDescr,
                     size_t                rows,
                     size_t                cols,
                     int64_t               ld,
                     void*                 valuesPtr,
                     cudaDataType          type,
                     cusparseOrder_t       order)
```

This function initializes the dense matrix descriptor `dnMatDescr`.

**Input**

<table>
<thead>
<tr>
<th>rows</th>
<th>number of rows of the dense matrix.</th>
</tr>
</thead>
<tbody>
<tr>
<td>cols</td>
<td>number of cols of the dense matrix.</td>
</tr>
<tr>
<td>ld</td>
<td>leading dimension of the dense matrix.</td>
</tr>
<tr>
<td>valuesPtr</td>
<td>value array of dimensions (ld x cols).</td>
</tr>
<tr>
<td>type</td>
<td>datatype of the dense matrix.</td>
</tr>
<tr>
<td>order</td>
<td>memory layout CUSPARSE_ORDER_COL.</td>
</tr>
</tbody>
</table>

**Output**

<table>
<thead>
<tr>
<th>dnMatDescr</th>
<th>dense matrix descriptor.</th>
</tr>
</thead>
</table>

13.3.2. cusparseDestroyDnMat()

```c
 cusparseStatus_t
 cusparseDestroyDnMat(cusparseDnMatDescr_t dnMatDescr)
```

This function releases the memory allocated for the sparse matrix descriptor `dnMatDescr`.

**Input**

<table>
<thead>
<tr>
<th>dnMatDescr</th>
<th>dense matrix descriptor.</th>
</tr>
</thead>
</table>
13.3.3. cusparseDnMatGet()

cusparseStatus_t
cusparseDnMatGet(const cusparseDnMatDescr_t dnMatDescr,
        size_t*           rows,
        size_t*           cols,
        int64_t*          ld,
        void**           valuesPtr,
        cudaDataType*    type,
        cusparseOrder_t* order)

This function returns the `dnMatDescr` fields of the dense matrix descriptor.

**Input**

<table>
<thead>
<tr>
<th><code>dnMatDescr</code></th>
<th>dense matrix descriptor.</th>
</tr>
</thead>
</table>

**Output**

<table>
<thead>
<tr>
<th><code>rows</code></th>
<th>number of rows of the dense matrix.</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>cols</code></td>
<td>number of cols of dense matrix.</td>
</tr>
<tr>
<td><code>ld</code></td>
<td>leading dimension of the dense matrix.</td>
</tr>
<tr>
<td><code>valuesPtr</code></td>
<td>value array of dimensions (ld x cols).</td>
</tr>
<tr>
<td><code>type</code></td>
<td>datatype of the dense matrix.</td>
</tr>
<tr>
<td><code>order</code></td>
<td>memory layout CUSPARSE_ORDER_COL.</td>
</tr>
</tbody>
</table>

13.3.4. cusparseDnMatGetStridedBatch()

cusparseStatus_t
cusparseDnMatGetStridedBatch(const cusparseDnMatDescr_t dnMatDescr,
            int         batchCount,
            size_t      batchStride)

This function returns the number of batches of the dense matrix descriptor `dnMatDescr`.

**Input**

<table>
<thead>
<tr>
<th><code>dnMatDescr</code></th>
<th>dense matrix descriptor.</th>
</tr>
</thead>
</table>

**Output**

<table>
<thead>
<tr>
<th><code>batchCount</code></th>
<th>number of batches of the dense matrix.</th>
</tr>
</thead>
</table>

13.3.5. cusparseDnMatSetNumBatches()

cusparseStatus_t
cusparseDnMatSetStridedBatch(cusparseDnMatDescr_t dnMatDescr,
               int          batchCount,
               size_t      batchStride)

This function sets the number of batches of the dense matrix descriptor `dnMatDescr`.

**Input**
13.4. Generic SpMM API functions

The sparse matrix-matrix multiplication functions are described in this section.

13.4.1. cusparseSpMM()

This function performs one of the following sparse matrix-matrix operations:

\[ C = \alpha \text{op}(A)\text{op}(B) + \beta C \]

where \( \text{op}(A) \) is a sparse matrix with dimensions \( m \times k \) and \( B \) and \( C \) are matrices stored in column-major format with dimensions of \( \text{op}(B) \): \( k \times n \) and the dimensions of \( C \): \( m \times n \).

\[
\text{op}(A) = \begin{cases} 
A & \text{if } \text{transA} == \text{CUSPARSE_OPERATION_NON_TRANSPOSE} \\
A^T & \text{if } \text{transA} == \text{CUSPARSE_OPERATION_TRANSPOSE} \\
A^H & \text{if } \text{transA} == \text{CUSPARSE_OPERATION_CONJUGATE_TRANSPOSE} 
\end{cases}
\]

It is executed asynchronously with respect to the host and may return control to the application on the host before the result is ready.

**Input**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>handle</td>
<td>handle to the cuSPARSE library context.</td>
</tr>
<tr>
<td>transA</td>
<td>the operation of ( \text{op}(A) ).</td>
</tr>
<tr>
<td>transB</td>
<td>the operation of ( \text{op}(B) ).</td>
</tr>
<tr>
<td>alpha</td>
<td>scaling factor for ( A^*B ); of same type as computeType.</td>
</tr>
<tr>
<td>matA</td>
<td>sparse matrix descriptor of the matrix ( A ).</td>
</tr>
<tr>
<td>matB</td>
<td>dense matrix descriptor of the matrix ( B ).</td>
</tr>
<tr>
<td>beta</td>
<td>scaling factor for ( C ); of same type as computeType.</td>
</tr>
<tr>
<td>matC</td>
<td>dense matrix descriptor of the matrix ( C ).</td>
</tr>
<tr>
<td>computeType</td>
<td>computation type.</td>
</tr>
</tbody>
</table>
alg | algorithm implementation for SpMM. see cusparseSpMMAlg_t for possible values.
---|---
externalBuffer | pointer to workspace buffer.

Output

matC | updated dense matrix descriptor of the matrix C.

13.4.2. cusparseSpMM_bufferSize()

cusparseStatus_t

cusparseSpMM_bufferSize(cusparseHandle_t handle, cusparseOperation_t transA, cusparseOperation_t transB, const void* alpha, const cusparseSpMatDescr_t matA, const cusparseDnMatDescr_t matB, const void* beta, cusparseDnMatDescr_t matC, cudaDataType computeType, cusparseSpMMAlg_t alg, size_t* bufferSize)

This function returns the size of the workspace needed by cusparseSpMM(). User needs to allocate a buffer of this size and give that buffer to cusparseSpMM() as an argument. All the arguments are similar to cusparseSpMM() except the following output argument.

Output

bufferSize | number of bytes of workspace needed by cusparseSpMM().
Chapter 14.
APPENDIX A: CUSPARSE LIBRARY C++ EXAMPLE

For sample code reference please see the example code below. It shows an application written in C++ using the cuSPARSE library API. The code performs the following actions:

1. Creates a sparse test matrix in COO format.
2. Creates a sparse and dense vector.
3. Allocates GPU memory and copies the matrix and vectors into it.
4. Initializes the cuSPARSE library.
5. Creates and sets up the matrix descriptor.
6. Converts the matrix from COO to CSR format.
7. Exercises Level 1 routines.
8. Exercises Level 2 routines.
9. Exercises Level 3 routines.
10. Destroys the matrix descriptor.
11. Releases resources for the cuSPARSE library.

```c
#include <stdio.h>
#include <stdlib.h>
#include <cuda_runtime.h>
#include "cusparse.h"

#define CLEANUP(s) do { 
    printf ("%s\n", s); 
    if (yHostPtr) free(yHostPtr); 
    if (xHostPtr) free(xHostPtr); 
    if (xValHostPtr) free(xValHostPtr); 
    if (cooRowIndexHostPtr) free(cooRowIndexHostPtr); 
    if (cooColIndexHostPtr) free(cooColIndexHostPtr); 
    if (cooValHostPtr) free(cooValHostPtr); 
    if (y) cudaFree(y); 
    if (z) cudaFree(z); 
    if (xInd) cudaFree(xInd); 
    if (xVal) cudaFree(xVal); 
    if (csrRowPtr) cudaFree(csrRowPtr); 
    if (cooRowIndex) cudaFree(cooRowIndex); 
    if (cooColIndex) cudaFree(cooColIndex); 
    if (cooVal) cudaFree(cooVal); 
    if (descr) cusparseDestroyMatDescr(descr); 
    cudaDeviceReset(); 
    fflush (stdout); 
} while (0)

int main(){
    cusparseStatus_t status;
    cusparseHandle_t handle=0;
    cusparseMatDescr_t descr=0;
    int * cooRowIndexHostPtr=0;
    int * cooColIndexHostPtr=0;
    double * cooValHostPtr=0;
    int * cooRowIndex=0;
    int * cooColIndex=0;
    double * cooVal=0;
    int * xIndHostPtr=0;
    double * xValHostPtr=0;
    int * yHost=0;
    double * yHostPtr=0;
    double * zHostPtr=0;
    double * z=0;
    int n, nnz, nnz_vector;
    double dzero =0.0;
    double dtwo =2.0;
    double dthree=3.0;
    double dfive =5.0;

    printf("testing example\n");
    /* create the following sparse test matrix in COO format */
    /* |1.0  2.0  3.0|  */
    /* |4.0 |  */
    /* |5.0  6.0  7.0|  */
    /* |8.0  9.0| */
    n=4; nnz=9;
    cooRowIndexHostPtr = (int *) malloc(nnz*sizeof(cooRowIndexHostPtr[0]));
    cooColIndexHostPtr = (int *) malloc(nnz*sizeof(cooColIndexHostPtr[0]));
    cooValHostPtr = (double *)malloc(nnz*sizeof(cooValHostPtr[0]));
    if (((cooRowIndexHostPtr) || (!cooColIndexHostPtr)) || (!cooValHostPtr)) {
        CLEANUP("Host malloc failed (matrix)");
        return 1;
    }
    cooRowIndexHostPtr[0]=0; cooColIndexHostPtr[0]=0; cooValHostPtr[0]=1.0;
    cooRowIndexHostPtr[1]=0; cooColIndexHostPtr[1]=2; cooValHostPtr[1]=2.0;
```
Chapter 15.
APPENDIX B: CUSPARSE FORTRAN BINDINGS

The cuSPARSE library is implemented using the C-based CUDA toolchain, and it thus provides a C-style API that makes interfacing to applications written in C or C++ trivial. There are also many applications implemented in Fortran that would benefit from using cuSPARSE, and therefore a cuSPARSE Fortran interface has been developed.

Unfortunately, Fortran-to-C calling conventions are not standardized and differ by platform and toolchain. In particular, differences may exist in the following areas:

Symbol names (capitalization, name decoration)
Argument passing (by value or reference)
Passing of pointer arguments (size of the pointer)

To provide maximum flexibility in addressing those differences, the cuSPARSE Fortran interface is provided in the form of wrapper functions, which are written in C and are located in the file `cusparse_fortran.c`. This file also contains a few additional wrapper functions (for `cudaMalloc()`, `cudaMemset`, and so on) that can be used to allocate memory on the GPU.

The cuSPARSE Fortran wrapper code is provided as an example only and needs to be compiled into an application for it to call the cuSPARSE API functions. Providing this source code allows users to make any changes necessary for a particular platform and toolchain.

The cuSPARSE Fortran wrapper code has been used to demonstrate interoperability with the compilers g95 0.91 (on 32-bit and 64-bit Linux) and g95 0.92 (on 32-bit and 64-bit Mac OS X). In order to use other compilers, users have to make any changes to the wrapper code that may be required.

The direct wrappers, intended for production code, substitute device pointers for vector and matrix arguments in all cuSPARSE functions. To use these interfaces, existing applications need to be modified slightly to allocate and deallocate data structures in GPU memory space (using `CUDA_MALLOC()` and `CUDA_FREE()`) and to copy data between GPU and CPU memory spaces (using the `CUDA_MEMCPY()` routines). The sample wrappers provided in `cusparse_fortran.c` map device pointers to the OS-
dependent type `size_t`, which is 32 bits wide on 32-bit platforms and 64 bits wide on a 64-bit platforms.

One approach to dealing with index arithmetic on device pointers in Fortran code is to use C-style macros and to use the C preprocessor to expand them. On Linux and Mac OS X, preprocessing can be done by using the option `-cpp` with g95 or gfortran. The function `GET_SHIFTED_ADDRESS()`, provided with the cuSPARSE Fortran wrappers, can also be used, as shown in example B.

Example B shows the the C++ of example A implemented in Fortran 77 on the host. This example should be compiled with `ARCH_64` defined as 1 on a 64-bit OS system and as undefined on a 32-bit OS system. For example, on g95 or gfortran, it can be done directly on the command line using the option `-cpp -DARCH_64=1.`
15.1. Example B, Fortran Application

```fortran
program cusparse_fortran_example
implicit none
integer cuda_malloc
external cuda_free
integer cuda_memcpy_c2fort_int
integer cuda_memcpy_c2fort_real
integer cuda_memcpy_fort2c_int
integer cuda_memcpy_fort2c_real
integer cuda_memset
integer cusparse_create
external cusparse_destroy
integer cusparse_get_version
integer cusparse_create_mat_descr
external cusparse_destroy_mat_descr
integer cusparse_set_mat_type
integer cusparse_set_mat_type
integer cusparse_get_mat_type
integer cusparse_get_mat_fill_mode
integer cusparse_get_mat_diag_type
integer cusparse_set_mat_index_base
integer cusparse_get_mat_index_base
integer cusparse_xcoo2csr
integer cusparse_dsctr
integer cusparse_dcsrmv
integer cusparse_dcsrmm
external get_shifted_address

#if ARCH_64
integer*8 handle
integer*8 descrA
integer*8 cooRowIndex
integer*8 cooColIndex
integer*8 cooVal
integer*8 xInd
integer*8 xVal
integer*8 y
integer*8 z
integer*8 csrRowPtr
integer*8 ynpl
#else
integer*4 handle
integer*4 descrA
integer*4 cooRowIndex
integer*4 cooColIndex
integer*4 cooVal
integer*4 xInd
integer*4 xVal
integer*4 y
integer*4 z
integer*4 csrRowPtr
integer*4 ynpl
#endif
integer status
integer cudaStat1,cudaStat2,cudaStat3
cinteger cudaStat4,cudaStat5,cudaStat6
cinteger n, nnz, nnz_vector
cparameter (n=4, nnz=9, nnz_vector=3)
cinteger cooRowIndexHostPtr(nnz)
cinteger cooColIndexHostPtr(nnz)
creal*8 cooValHostPtr(nnz)
cinteger xIndHostPtr(nnz_vector)
creal*8 xValHostPtr(nnz_vector)
creal*8 yHostPtr(2*n)
creal*8 zHostPtr(2*(n+1))
cinteger i, j
integer version, mtype, fmode, dtype, ibase
creal*8 dzero,dtwo,dthree,dfive
creal*8 epsilon

write(*,*) "section fortran-example:"
```

16.1. power method

This chapter provides a simple example in the C programming language of eigenvalue problem

\[ A^*x = \lambda^*x \]

A is a 4x4 sparse matrix,

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

The goal is to find the largest eigen-pair by power method. The following code uses \texttt{csrmv_mp} inside the loop of power method.
Appendix C: Examples of sparse matrix vector multiplication

/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include csrmvmp_example.cpp
 *   g++ -fopenmp -o csrmvmp_example csrmvmp_example.o -L/usr/local/cuda/lib64 -lcublas -lcusparse -lcudart
 *#
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cublas_v2.h>
#include <cusparse.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
", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cublasHandle_t cublasH = NULL;
    cusparseHandle_t cusparseH = NULL;
    cudaStream_t stream = NULL;
    cusparseMatDescr_t descrA = NULL;

    cublasStatus_t cublasStat = CUBLAS_STATUS_SUCCESS;
    cusparseStatus_t cusparseStat = CUSPARSE_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    cudaError_t cudaStat2 = cudaSuccess;
    cudaError_t cudaStat3 = cudaSuccess;
    cudaError_t cudaStat4 = cudaSuccess;
    cudaError_t cudaStat5 = cudaSuccess;
    const int n = 4;
    const int nnzA = 9;

    * A = 1 0 2 3
      0 4 0 0
      5 0 6 7
      0 8 0 9

    * eigevales are { -0.5311, 7.5311, 9.0000, 4.0000 }
    * The largest eigenvalue is 9 and corresponding eigenvector is
    *  v = 0.3029
      0
      0.9350
      0.1844
*/
Appendix C: Examples of sparse matrix vector multiplication

```c
const int csrRowPtrA[n+1] = { 0, 3, 4, 7, 9 };
const int csrColIndA[nnzA] = {0, 2, 3, 1, 0, 2, 3, 1, 3 };
const double csrValA[nnzA] = {1.0, 2.0, 3.0, 4.0, 5.0, 6.0, 7.0, 8.0, 9.0 };
const double lambda_exact[n] = { 9.0000, 7.5311, 4.0000, -0.5311 };
const double x0[n] = {1.0, 2.0, 3.0, 4.0 }; /* initial guess */
double x[n]; /* numerical eigenvector */

int *d_csrRowPtrA = NULL;
int *d_csrColIndA = NULL;
double *d_csrValA = NULL;

double *d_x = NULL; /* eigenvector */
double *d_y = NULL; /* workspace */

const double tol = 1.e-6;
const int max_ites = 30;

const double h_one = 1.0;
const double h_zero = 0.0;

printf("example of csrmv_mp \n");
printf("tol = %E \n", tol);
printf("max. iterations = %d \n", max_ites);

printf("1st eigenvalue is %f\n", lambda_exact[0] );
printf("2nd eigenvalue is %f\n", lambda_exact[1] );

double alpha = lambda_exact[1]/lambda_exact[0] ;
printf("convergence rate is %f\n", alpha );

double est_iterations = log(tol)/log(alpha);
printf("# of iterations required is %d\n", (int)ceil(est_iterations) );

/* step 1: create cublas/cusparse handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

cublasStat = cublasCreate(&cublasH);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

cublasStat = cublasSetStream(cublasH, stream);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

cusparseStat = cusparseCreate(&cusparseH);
assert(CUSPARSE_STATUS_SUCCESS == cusparseStat);

cusparseStat = cusparseSetStream(cusparseH, stream);
assert(CUSPARSE_STATUS_SUCCESS == cusparseStat);

/* step 2: configuration of matrix A */
cusparseStat = cusparseCreateMatDescr(&descrA);
assert(CUSPARSE_STATUS_SUCCESS == cusparseStat);

cusparseSetMatIndexBase(descrA, CUSPARSE_INDEX_BASE_ZERO);
cusparseSetMatType(descrA, CUSPARSE_MATRIX_TYPE_GENERAL );
```
Appendix C: Examples of sparse matrix vector multiplication

power method

/* step 3: copy A and x0 to device */
cudaStat1 = cudaMalloc ((void**)&d_csrRowPtrA, sizeof(int) * (n+1) );
cudaStat2 = cudaMalloc ((void**)&d_csrColIndA, sizeof(int) * nnzA );
cudaStat3 = cudaMalloc ((void**)&d_csrValA   , sizeof(double) * nnzA );
cudaStat4 = cudaMalloc ((void**)&d_x         , sizeof(double) * n );
cudaStat5 = cudaMalloc ((void**)&d_y         , sizeof(double) * n );
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);
assert(cudaSuccess == cudaStat4);
assert(cudaSuccess == cudaStat5);

cudaStat1 = cudaMemcpy(d_csrRowPtrA, csrRowPtrA, sizeof(int) * (n+1)  ,
cudaMemcpyHostToDevice);
cudaStat2 = cudaMemcpy(d_csrColIndA, csrColIndA, sizeof(int) * nnzA  ,
cudaMemcpyHostToDevice);
cudaStat3 = cudaMemcpy(d_csrValA   , csrValA   , sizeof(double) * nnzA ,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);

/* step 4: power method */
double lambda = 0.0;
double lambda_next = 0.0;

/* 4.1: initial guess x0 */
cudaStat1 = cudaMemcpy(d_x, x0, sizeof(double) * n, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
for(int ite = 0 ; ite < max_ites ; ite++ ){
    /* 4.2: normalize vector x */
    double nrm2_x;
    cublasStat = cublasDnrm2_v2(cublasH,
                             n,
                             d_x, // incx,
                             &nrm2_x /* host pointer */
                          );
    assert(CUBLAS_STATUS_SUCCESS == cublasStat);
    double one_over_nrm2_x = 1.0 / nrm2_x;
    cublasStat = cublasDscal_v2( cublasH,
                                n,
                                &one_over_nrm2_x, /* host pointer */
                               d_x, // incx
                          );
    assert(CUBLAS_STATUS_SUCCESS == cublasStat);
}
Appendix C: Examples of sparse matrix vector multiplication

```c
/* 4.3: y = A*x */
cusparseStat = cusparseDcsrmv_mp(cusparseH,
    CUSPARSE_OPERATION_NON_TRANSPOSE,
    n, n, nnzA, &h_one,
    descrA, d_csrValA, d_csrRowPtrA, d_csrColIndA, d_x, &h_zero, d_y);
assert(CUSPARSE_STATUS_SUCCESS == cusparseStat);

/* 4.4: lambda = y**T*x */
cublasStat = cublasDdot_v2 ( cublasH,
    n, d_x, 1, // incx,
    d_y, 1, // incy,
    &lambda_next /* host pointer */
); 
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

double lambda_err = fabs( lambda_next - lambda_exact[0] );
printf("ite %d: lambda = %f, error = %E\n", ite, lambda_next, lambda_err);

/* 4.5: check if converges */
if ( (ite > 0) && fabs( lambda - lambda_next ) < tol
    }
    break; // converges
}

/* 4.6: x := y */
lambda = lambda_next

* so new approximation is (lambda, x), x is not normalized. */
cudaStat1 = cudaMemcpy(d_x, d_y, sizeof(double) * n , cudaMemcpyDeviceToDevice);
assert(cudaSuccess == cudaStat1);
lambda = lambda_next;
```

report largest eigenvalue and eigenvector.

```c
/* step 5: report eigen-pair */
cudaStat1 = cudaMemcpy(x, d_x, sizeof(double) * n, cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printf("largest eigenvalue is %E\n", lambda);
printf("eigenvector = (matlab base-1)\n");
printMatrix(n, 1, x, n, "V0");
printf("====\n");

/* free resources */
if (d_csrRowPtrA) cudaFree(d_csrRowPtrA);
if (d_csrColIndA) cudaFree(d_csrColIndA);
if (d_csrValA) cudaFree(d_csrValA);
if (d_x) cudaFree(d_x);
if (d_y) cudaFree(d_y);
if (cublasH) cublasDestroy(cublasH);
if (cusparseH) cusparseDestroy(cusparseH);
if (stream) cudaStreamDestroy(stream);
if (descrA) cusparseDestroyMatDescr(descrA);

cudaDeviceReset();
return 0;
```
Chapter 17.
APPENDIX D: EXAMPLES OF SORTING

17.1. COO sort

This chapter provides a simple example in the C programming language of sorting of COO format.

A is a 3x3 sparse matrix,

\[
A = \begin{pmatrix}
1.0 & 2.0 & 0.0 \\
0.0 & 5.0 & 0.0 \\
0.0 & 8.0 & 0.0
\end{pmatrix}
\]
int main(int argc, char*argv[]) {
  cusparseHandle_t handle = NULL;
  cudaStream_t stream = NULL;
  cusparseStatus_t status = CUSPARSE_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 = 3;
  const int nnz = 4;

  if (0) /* index starts at 0 */
  {
    int h_cooRows[nnz] = {2, 1, 0, 0};
    int h_cooCols[nnz] = {1, 1, 0, 1};
  } else /* index starts at -2 */
  {
    int h_cooRows[nnz] = {0, -1, -2, -2};
    int h_cooCols[nnz] = {-1, -1, -2, -1};
  }
  double h_cooVals[nnz] = {8.0, 5.0, 1.0, 2.0};
  int h_P[nnz];

  int *d_cooRows = NULL;
  int *d_cooCols = NULL;
  int *d_P = NULL;
  double *d_cooVals = NULL;
  double *d_cooVals_sorted = NULL;
  size_t pBufferSizeInBytes = 0;
  void *pBuffer = NULL;

  printf("m = %d, n = %d, nnz=%d \n", m, n, nnz);
Appendix D: Examples of sorting

...
Appendix D: Examples of sorting

/* step 4: sort COO format by Row */
status = cusparseXcoosortByRow(
    handle,
    m,
    n,
    nnz,
    d_cooRows,
    d_cooCols,
    d_P,
pBuffer
);  
assert( CUSPARSE_STATUS_SUCCESS == status);

/* step 5: gather sorted cooVals */
status = cusparseDgthr(
    handle,
    nnz,
    d_cooVals,
    d_cooVals_sorted,
    d_P,
    CUSPARSE_INDEX_BASE_ZERO
);  
assert( CUSPARSE_STATUS_SUCCESS == status);

cudaStat1 = cudaDeviceSynchronize(); /* wait until the computation is done */
cudaStat2 = cudaMemcpy(h_cooRows, d_cooRows, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat3 = cudaMemcpy(h_cooCols, d_cooCols, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat4 = cudaMemcpy(h_P,       d_P      , sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat5 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooRows, d_cooRows, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooCols, d_cooCols, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_P,       d_P      , sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooRows, d_cooRows, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooCols, d_cooCols, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_P,       d_P      , sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooRows, d_cooRows, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooCols, d_cooCols, sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_P,       d_P      , sizeof(int)*nnz   , cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);
cudaStat6 = cudaMemcpy(h_cooVals, d_cooVals_sorted, sizeof(double)*nnz, cudaMemcpyDeviceToHost);

printf("sorted coo: \n\n");
for(int j = 0; j < nnz; j++){
    printf("(%d, %d, %f) \n", h_cooRows[j], h_cooCols[j], h_cooVals[j] );
}

for(int j = 0; j < nnz; j++){
    printf("P[%d] = %d \n", j, h_P[j] );
}

/* free resources */
if (d_cooRows    ) cudaFree(d_cooRows);
if (d_cooCols    ) cudaFree(d_cooCols);
if (d_P          ) cudaFree(d_P);
if (d_cooVals    ) cudaFree(d_cooVals);
if (d_cooVals_sorted    ) cudaFree(d_cooVals_sorted);
if (pBuffer       ) cudaFree(pBuffer);
if (handle        ) cusparseDestroy(handle);
if (stream        ) cudaStreamDestroy(stream);
cudaDeviceReset();
return 0;
18.1. prune dense to sparse

This section provides a simple example in the C programming language of pruning a dense matrix to a sparse matrix of CSR format.

A is a 4x4 dense matrix,

\[
A = \begin{bmatrix}
1.0 & 0.0 & 2.0 & -3.0 \\
0.0 & 4.0 & 0.0 & 0.0 \\
5.0 & 0.0 & 6.0 & 7.0 \\
0.0 & 8.0 & 0.0 & 9.0
\end{bmatrix}
\]
Appendix E: Examples of prune

```c
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusparse.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) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

void printCsr(
    int m,
    int n,
    int nnz,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const char* name)
{
    const int base = (cusparseGetMatIndexBase(descrA) !=
                    CUSPARSE_INDEX_BASE_ONE)? 0:1 ;
    printf("matrix %s is %d-by-%d, nnz=%d, base=%d\n", name, m, n, nnz, base);
    for(int row = 0 ; row < m ; row++) {
        const int start = csrRowPtrA[row  ] - base;
        const int end   = csrRowPtrA[row+1] - base;
        for(int colidx = start ; colidx < end ; colidx++) {
            const int col = csrColIndA[colidx] - base;
            const float Areg = csrValA[colidx];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusparseHandle_t handle = NULL;
    cudaStream_t stream = NULL;
    cusparseMatDescr_t descrC = NULL;
    ...
Appendix E: Examples of prune

```c
 cusparseStatus_t status = CUSPARSE_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 = 4;
const int n = 4;
const int lda = m;
/*
 *      |    1     0     2     -3 |
 *      |    0     4     0     0   |
 *  A = |    5     0     6     7   |
 *      |    0     8     0     9   |
 */
const float A[lda*n] = {1, 0, 5, 0, 0, 4, 0, 8, 2, 0, 6, 0, -3, 0, 7, 9};
int* csrRowPtrC = NULL;
int* csrColIndC = NULL;
float* csrValC = NULL;
float *d_A = NULL;
int *d_csrRowPtrC = NULL;
int *d_csrColIndC = NULL;
float *d_csrValC = NULL;
size_t lworkInBytes = 0;
char *d_work = NULL;
int nnzC = 0;
float threshold = 4.1; /* remove Aij <= 4.1 */
// float threshold = 0; /* remove zeros */
printf("example of pruneDense2csr \n");
printf("prune |A(i,j)| <= threshold \n");
printf("threshold = %E \n", threshold);
printMatrix(m, n, A, lda, "A");
/* step 1: create cusparse handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);
status = cusparseCreate(&handle);
assert(CUSPARSE_STATUS_SUCCESS == status);
status = cusparseSetStream(handle, stream);
assert(CUSPARSE_STATUS_SUCCESS == status);
```

Appendix E: Examples of prune

/* step 2: configuration of matrix C */
status = cusparseCreateMatDescr(&descrC);
assert(CUSPARSE_STATUS_SUCCESS == status);

cusparseSetMatIndexBase(descrC, CUSPARSE_INDEX_BASE_ZERO);
cusparseSetMatType(descrC, CUSPARSE_MATRIX_TYPE_GENERAL);

cudaStat1 = cudaMalloc ((void**)&d_A         , sizeof(float)*lda*n );
cudaStat2 = cudaMalloc ((void**)&d_csrRowPtrC, sizeof(int)*(m+1) );
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

/* step 3: query workspace */
cudaStat1 = cudaMemcpy(d_A, A, sizeof(float)*lda*n, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

status = cusparseSpruneDense2csr_bufferSizeExt(
    handle,
    m, 
    n, 
    d_A, 
    lda, 
    &threshold, 
    descrC, 
    d_csrValC, 
    d_csrRowPtrC, 
    d_csrColIndC, 
    &lworkInBytes);
assert(CUSPARSE_STATUS_SUCCESS == status);
printf("lworkInBytes (prune) = %lld \n", (long long)lworkInBytes);

if (NULL != d_work) { cudaFree(d_work); }
cudaStat1 = cudaMalloc((void**)&d_work, lworkInBytes);
assert(cudaSuccess == cudaStat1);

/* step 4: compute csrRowPtrC and nnzC */
status = cusparseSpruneDense2csrNnz(
    handle,
    m, 
    n, 
    d_A, 
    lda, 
    &threshold, 
    descrC, 
    d_csrRowPtrC, 
    &nnzC, /* host */
    d_work);
assert(CUSPARSE_STATUS_SUCCESS == status);
cudaStat1 = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat1);

printf("nnzC = %d\n", nnzC);
if (0 == nnzC ){
    printf("C is empty \n");
    return 0;
}
/* step 5: compute csrColIndC and csrValC */
cudaStat1 = cudaMalloc ((void**)&d_csrColIndC, sizeof(int ) * nnzC );
cudaStat2 = cudaMalloc ((void**)&d_csrValC   , sizeof(float) * nnzC );
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);

status = cusparseSpruneDense2csr(
    handle,
    m,
    n,
    d_A,
    lda,
    &threshold,
    descrC,
    d_csrValC,
    d_csrRowPtrC,
    d_csrColIndC,
    d_work);
assert(CUSPARSE_STATUS_SUCCESS == status);
cudaStat1 = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat1);

/* step 6: output C */
csrRowPtrC = (int* )malloc(sizeof(int )*(m+1));
csrColIndC = (int* )malloc(sizeof(int )*nnzC);
csrValC    = (float*)malloc(sizeof(float)*nnzC);
assert( NULL != csrRowPtrC);
assert( NULL != csrColIndC);
assert( NULL != csrValC);

cudaStat1 = cudaMemcpy(csrRowPtrC, d_csrRowPtrC, sizeof(int )*(m+1),
cudaMemcpyDeviceToHost);
cudaStat2 = cudaMemcpy(csrColIndC, d_csrColIndC, sizeof(int )*nnzC ,
cudaMemcpyDeviceToHost);
cudaStat3 = cudaMemcpy(csrValC   , d_csrValC   , sizeof(float)*nnzC ,
cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
assert(cudaSuccess == cudaStat3);

printCsr(m, n, nnzC, descrC, csrValC, csrRowPtrC, csrColIndC, "C");

/* free resources */
if (d_A           ) cudaFree(d_A);
if (d_csrRowPtrC  ) cudaFree(d_csrRowPtrC);
if (d_csrColIndC  ) cudaFree(d_csrColIndC);
if (d_csrValC     ) cudaFree(d_csrValC);
if (csrRowPtrC  ) free(csrRowPtrC);
if (csrColIndC  ) free(csrColIndC);
if (csrValC     ) free(csrValC);
if (handle        ) cusparseDestroy(handle);
if (stream        ) cudaStreamDestroy(stream);
if (descrC        ) cusparseDestroyMatDescr(descrC);

cudaDeviceReset();
return 0;
18.2. prune sparse to sparse

This section provides a simple example in the C programming language of pruning a sparse matrix to a sparse matrix of CSR format. A is a $4 \times 4$ sparse matrix,

$$
A = \begin{bmatrix}
1.0 & 0.0 & 2.0 & -3.0 \\
0.0 & 4.0 & 0.0 & 0.0 \\
5.0 & 0.0 & 6.0 & 7.0 \\
0.0 & 8.0 & 0.0 & 9.0 \\
\end{bmatrix}
$$
Appendix E: Examples of prune

```c
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusparse.h>

void printCsr(
    int m,
    int n,
    int nnz,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const char* name)
{
    const int base = (cusparseGetMatIndexBase(descrA) !=
        CUSPARSE_INDEX_BASE_ONE)? 0:1 ;
    printf("matrix %s is %d-by-%d, nnz=%d, base=%d, output base-1\n", name, m,
        n, nnz, base);
    for(int row = 0 ; row < m ; row++){
        const int start = csrRowPtrA[row] - base;
        const int end   = csrRowPtrA[row+1] - base;
        for(int colidx = start ; colidx < end ; colidx++){
            const int col = csrColIndA[colidx] - base;
            const float Areg = csrValA[colidx];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusparseHandle_t handle = NULL;
    cudaStream_t stream = NULL;
    cusparseMatDescr_t descrA = NULL;
    cusparseMatDescr_t descrC = NULL;
    cusparseStatus_t status = CUSPARSE_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    const int m = 4;
    const int n = 4;
    const int nnzA = 9;
    /*
     | 1  0  2  -3 |
     | 0  4  0  0 |
     A = | 5  0  6  7 |
     | 0  8  0  9 |
     */
```
const int csrRowPtrA[m+1] = {1, 4, 5, 8, 10};
const int csrColIndA[nnzA] = {1, 3, 4, 2, 1, 3, 4, 2, 4};
const float csrValA[nnzA] = {1, 3, 4, 2, 1, 3, 4, 2, 4};

int* csrRowPtrC = NULL;
int* csrColIndC = NULL;
float* csrValC = NULL;

int *d_csrRowPtrA = NULL;
int *d_csrColIndA = NULL;
float *d_csrValA = NULL;

int *d_csrRowPtrC = NULL;
int *d_csrColIndC = NULL;
float *d_csrValC = NULL;

size_t lworkInBytes = 0;
char *d_work = NULL;

int nnzC = 0;

float threshold = 4.1; /* remove Aij <= 4.1 */
// float threshold = 0; /* remove zeros */
printf("example of pruneCsr2csr \n");
printf("prune |A(i,j)| <= threshold \n");
printf("threshold = %E \n", threshold);

/* step 1: create cusparse handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);
status = cusparseCreate(&handle);
assert(CUSPARSE_STATUS_SUCCESS == status);
status = cusparseSetStream(handle, stream);
assert(CUSPARSE_STATUS_SUCCESS == status);

/* step 2: configuration of matrix A and C */
status = cusparseCreateMatDescr(&descrA);
assert(CUSPARSE_STATUS_SUCCESS == status);
/* A is base=1 */
cusparseSetMatIndexBase(descrA, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descrA, CUSPARSE_MATRIX_TYPE_GENERAL);

status = cusparseCreateMatDescr(&descrC);
assert(CUSPARSE_STATUS_SUCCESS == status);
/* C is base=0 */
cusparseSetMatIndexBase(descrC, CUSPARSE_INDEX_BASE_ZERO);
cusparseSetMatType(descrC, CUSPARSE_MATRIX_TYPE_GENERAL);

printCsr(m, n, nnzA, descrA, csrValA, csrRowPtrA, csrColIndA, "A");
Appendix E: Examples of prune

```c
... cudaStat1 = cudaMalloc ((void**)&d_csrRowPtrA, sizeof(int)*(m+1) );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_csrColIndA, sizeof(int)*nnzA );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_csrValA   , sizeof(float)*nnzA );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_csrRowPtrC, sizeof(int)*(m+1) );
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(d_csrRowPtrA, csrRowPtrA, sizeof(int)*(m+1), cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_csrColIndA, csrColIndA, sizeof(int)*nnzA, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_csrValA   , csrValA   , sizeof(float)*nnzA, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

/* step 3: query workspace */
status = cusparseSpruneCsr2csr_bufferSizeExt(
    handle,
    m, n, nnzA, descrA,
    d_csrValA, d_csrRowPtrA, d_csrColIndA, &threshold,
    descrC, d_csrValC, d_csrRowPtrC, d_csrColIndC, &lworkInBytes);
assert(CUSPARSE_STATUS_SUCCESS == status);

printf("lworkInBytes (prune) = %lld \n", (long long)lworkInBytes);

if (NULL != d_work) { cudaFree(d_work); }
cudaStat1 = cudaMalloc((void**)&d_work, lworkInBytes);
assert(cudaSuccess == cudaStat1);

/* step 4: compute csrRowPtrC and nnzC */
status = cusparseSpruneCsr2csrNnz(
    handle,
    m, n, nnzA, descrA,
    d_csrValA, d_csrRowPtrA, d_csrColIndA, &threshold,
    descrC, d_csrValC, d_csrRowPtrC, &nnzC, /* host */
    d_work);
assert(CUSPARSE_STATUS_SUCCESS == status);
cudaStat1 = cudaMemcpy(d_work, lworkInBytes);
assert(cudaSuccess == cudaStat1);
```
...
18.3. prune dense to sparse by percentage

This section provides a simple example in the C programming language of pruning a dense matrix to a sparse matrix by percentage.

A is a 4x4 dense matrix,

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

The percentage is 50, which means to prune 50 percent of the dense matrix. The matrix has 16 elements, so 8 out of 16 must be pruned out. Therefore 7 zeros are pruned out, and value 1.0 is also out because it is the smallest among 9 nonzero elements.
Appendix E: Examples of prune

...  

#include <stdio.h>    
#include <stdlib.h>    
#include <assert.h>    
#include <cuda_runtime.h>    
#include <cusparse.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) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

void printCsr(  
    int m,  
    int n,  
    int nnz,  
    const cusparseMatDescr_t descrA,  
    const float *csrValA,  
    const int *csrRowPtrA,  
    const int *csrColIndA,  
    const char* name)
{
    const int base = (cusparseGetMatIndexBase(descrA) !=  
                     CUSPARSE_INDEX_BASE_ONE)? 0:1 ;

    printf("matrix %s is %d-by-%d, nnz=%d, base=%d, output base-1\n", name, m, n, nnz, base);
    for(int row = 0 ; row < m ; row++){
        int start = csrRowPtrA[row] - base;
        int end   = csrRowPtrA[row+1] - base;
        for(int colidx = start ; colidx < end ; colidx++){
            int col = csrColIndA[colidx] - base;
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, csrValA[colidx]);
        }
    }
}

int main(int argc, char*argv[])
{
    cusparseHandle_t handle = NULL;
    cudaStream_t stream = NULL;
    cusparseMatDescr_t descrC = NULL;
    pruneInfo_t  info = NULL;
    cusparseStatus_t status = CUSPARSE_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 = 4;
    const int n = 4;
    const int lda = m;  
    ...
Appendix E: Examples of prune

```c
/*
  |    1     0     2     -3  |
  |    0     4     0     0   |
  A = |    5     0     6     7   |
  |    0     8     0     9   |
*/
const float A[lda*n] = {1, 0, 5, 0, 4, 0, 8, 2, 0, 6, 0, -3, 0, 7, 9};
int* csrRowPtrC = NULL;
int* csrColIndC = NULL;
float* csrValC = NULL;
float *d_A = NULL;
int *d_csrRowPtrC = NULL;
int *d_csrColIndC = NULL;
float *d_csrValC = NULL;
size_t lworkInBytes = 0;
char *d_work = NULL;
int nnzC = 0;
float percentage = 50; /* 50% of nnz */
printf("example of pruneDense2csrByPercentage \n");
printf("prune out %.1f percentage of A \n", percentage);
printMatrix(m, n, A, lda, "A");
/* step 1: create cusparse handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);
status = cusparseCreate(&handle);
assert(CUSPARSE_STATUS_SUCCESS == status);
status = cusparseSetStream(handle, stream);
assert(CUSPARSE_STATUS_SUCCESS == status);
status = cusparseCreatePruneInfo(&info);
assert(CUSPARSE_STATUS_SUCCESS == status);
/* step 2: configuration of matrix C */
status = cusparseCreateMatDescr(&descrC);
assert(CUSPARSE_STATUS_SUCCESS == status);
cusparseSetMatIndexBase(descrC, CUSPARSE_INDEX_BASE_ZERO);
cusparseSetMatType(descrC, CUSPARSE_MATRIX_TYPE_GENERAL);
cudaStat1 = cudaMalloc ((void**)&d_A, sizeof(float)*lda*n);
cudaStat2 = cudaMalloc ((void**)&d_csrRowPtrC, sizeof(int)*(m+1));
assert(cudaSuccess == cudaStat1);
assert(cudaSuccess == cudaStat2);
cudaStat1 = cudaMemcpy(d_A, A, sizeof(float)*lda*n, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
```
/* step 3: query workspace */
    status = cusparseSpruneDense2csrByPercentage_bufferSizeExt(
        handle,
        m,
        n,
        d_A,
        lda,
        percentage,
        descrC,
        d csrValC,
        d csrRowPtrC,
        d csrColIndC,
        info,
        lworkInBytes);
    assert(CUSPARSE_STATUS_SUCCESS == status);

    printf("lworkInBytes = %lld \n", (long long) lworkInBytes);
    if (NULL != d_work) { cudaFree(d_work); }
    cudaStat1 = cudaMalloc((void**)& d_work, lworkInBytes);
    assert(cudaSuccess == cudaStat1);

/* step 4: compute csrRowPtrC and nnzC */
    status = cusparseSpruneDense2csrNnzByPercentage(
        handle,
        m,
        n,
        d_A,
        lda,
        percentage,
        descrC,
        d csrRowPtrC,
        &nnzC, /* host */
        info,
        d_work);
    assert(CUSPARSE_STATUS_SUCCESS == status);
    cudaStat1 = cudaDeviceSynchronize();
    assert(cudaSuccess == cudaStat1);

    printf("nnzC = %d\n", nnzC);
    if (0 == nnzC ){
        printf("C is empty \n");
        return 0;
    }

/* step 5: compute csrColIndC and csrValC */
    cudaStat1 = cudaMalloc ((void**)& d csrColIndC, sizeof(int ) * nnzC );
    cudaStat2 = cudaMalloc ((void**)& d csrValC , sizeof(float ) * nnzC );
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
Appendix E: Examples of prune

```c
... status = cusparseSpruneDense2csrByPercentage(
    handle,
    m,
    n,
    d_A,
    lda,
    percentage,
    descrC,
    d_csrValC,
    d_csrRowPtrC,
    d_csrColIndC,
    info,
    d_work);
    assert(CUSPARSE_STATUS_SUCCESS == status);
    cudaStat1 = cudaDeviceSynchronize();
    assert(cudaSuccess == cudaStat1);

    /* step 7: output C */
    csrRowPtrC = (int*) malloc(sizeof(int)*(m+1));
    csrColIndC = (int*) malloc(sizeof(int)*nnzC);
    csrValC = (float*) malloc(sizeof(float)*nnzC);
    assert( NULL != csrRowPtrC);
    assert( NULL != csrColIndC);
    assert( NULL != csrValC);
    cudaStat1 = cudaMemcpy(csrRowPtrC, d_csrRowPtrC, sizeof(int)*(m+1),
                        cudaMemcpyDeviceToHost);
    cudaStat2 = cudaMemcpy(csrColIndC, d_csrColIndC, sizeof(int)*nnzC ,
                        cudaMemcpyDeviceToHost);
    cudaStat3 = cudaMemcpy(csrValC   , d_csrValC   , sizeof(float)*nnzC ,
                        cudaMemcpyDeviceToHost);
    assert(cudaSuccess == cudaStat1);
    assert(cudaSuccess == cudaStat2);
    assert(cudaSuccess == cudaStat3);
    printCsr(m, n, nnzC, descrC, csrValC, csrRowPtrC, csrColIndC, "C");

    /* free resources */
    if (d_A) cudaFree(d_A);
    if (d_csrRowPtrC) cudaFree(d_csrRowPtrC);
    if (d_csrColIndC) cudaFree(d_csrColIndC);
    if (d_csrValC) cudaFree(d_csrValC);
    if (csrRowPtrC) free(csrRowPtrC);
    if (csrColIndC) free(csrColIndC);
    if (csrValC) free(csrValC);
    if (handle) cusparseDestroy(handle);
    if (stream) cudaStreamDestroy(stream);
    if (descrC) cusparseDestroyMatDescr(descrC);
    if (info) cusparseDestroyPruneInfo(info);
    cudaDeviceReset();
    return 0;
}
```

www.nvidia.com

cuSPARSE Library
18.4. prune sparse to sparse by percentage

This section provides a simple example in the C programming language of pruning a sparse matrix to a sparse matrix by percentage.

A is a 4x4 sparse matrix,

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

The percentage is 20, which means to prune 20 percent of the nonzeros. The sparse matrix has 9 nonzero elements, so 1.4 elements must be pruned out. The function removes 1.0 and 2.0 which are first two smallest numbers of nonzeros.
Appendix E: Examples of prune

```c
/*
 * How to compile (assume cuda is installed at /usr/local/cuda/)
 *   nvcc -c -I/usr/local/cuda/include prunecsr2csrByP.cpp
 *   g++ -o prunecsr2csrByP prunecsr2csrByP.o -L/usr/local/cuda/lib64 -lcusparse -lcudart
 */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusparse.h>

void printCsr(
    int m,
    int n,
    int nnz,
    const cusparseMatDescr_t descrA,
    const float *csrValA,
    const int *csrRowPtrA,
    const int *csrColIndA,
    const char* name)
{
    const int base = (cusparseGetMatIndexBase(descrA) !=
    CUSPARSE_INDEX_BASE_ONE)? 0:1 ;

    printf("matrix %s is %d-by-%d, nnz=%d, base=%d, output base-1\n", name, m, n, nnz, base);
    for(int row = 0 ; row < m ; row++){
        const int start = csrRowPtrA[row] - base;
        const int end   = csrRowPtrA[row+1] - base;
        for(int colidx = start ; colidx < end ; colidx++){
            const int col = csrColIndA[colidx] - base;
            const float Areg = csrValA[colidx];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
}

int main(int argc, char*argv[])
{
    cusparseHandle_t handle = NULL;
    cudaStream_t stream = NULL;
    cusparseMatDescr_t descrA = NULL;
    cusparseMatDescr_t descrC = NULL;
    pruneInfo_t info = NULL;
    cusparseStatus_t status = CUSPARSE_STATUS_SUCCESS;
    cudaError_t cudaStat1 = cudaSuccess;
    const int m = 4;
    const int n = 4;
    const int nnzA = 9;

    // A matrix
    float csrValA[9] = {1, 0, 2, -3, 0, 4, 0, 0, 0};
    int csrRowPtrA[5] = {1, 3, 4, 5, 6};
    int csrColIndA[9] = {1, 2, 2, 3, 4, 5, 5, 6, 6};

    printf("matrix A is 4-by-4, nnz=9, base=0, output base-1\n", m, n, nnzA);
    for(int row = 0 ; row < m ; row++){
        const int start = csrRowPtrA[row] - base;
        const int end   = csrRowPtrA[row+1] - base;
        for(int colidx = start ; colidx < end ; colidx++){
            const int col = csrColIndA[colidx] - base;
            const float Areg = csrValA[colidx];
            printf("%s(%d,%d) = %f\n", name, row+1, col+1, Areg);
        }
    }
    printf("\n");
```
Appendix E: Examples of prune

```c
const int csrRowPtrA[m+1] = { 1, 4, 5, 8, 10};
const int csrColIndA[nnzA] = { 1, 3, 4, 2, 1, 3, 4, 2, 4};
const float csrValA[nnzA] = {1, 2, -3, 4, 5, 6, 7, 8, 9};

int* csrRowPtrC = NULL;
int* csrColIndC = NULL;
float* csrValC = NULL;

int *d_csrRowPtrA = NULL;
int *d_csrColIndA = NULL;
float *d_csrValA = NULL;

int *d_csrRowPtrC = NULL;
int *d_csrColIndC = NULL;
float *d_csrValC = NULL;

size_t lworkInBytes = 0;
char *d_work = NULL;

int nnzC = 0;

float percentage = 20; /* remove 20% of nonzeros */

printf("example of pruneCsr2csrByPercentage \n");
printf("prune %.1f percent of nonzeros \n", percentage);

/* step 1: create cusparse handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusparseCreate(&handle);
assert(CUSPARSE_STATUS_SUCCESS == status);

status = cusparseSetStream(handle, stream);
assert(CUSPARSE_STATUS_SUCCESS == status);

status = cusparseCreatePruneInfo(&info);
assert(CUSPARSE_STATUS_SUCCESS == status);

/* step 2: configuration of matrix C */
status = cusparseCreateMatDescr(&descrA);
assert(CUSPARSE_STATUS_SUCCESS == status);
/* A is base-1*/
cusparseSetMatIndexBase(descrA, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descrA, CUSPARSE_MATRIX_TYPE_GENERAL );

status = cusparseCreateMatDescr(&descrC);
assert(CUSPARSE_STATUS_SUCCESS == status);
/* C is base-0 */
cusparseSetMatIndexBase(descrC, CUSPARSE_INDEX_BASE_ZERO);
cusparseSetMatType(descrC, CUSPARSE_MATRIX_TYPE_GENERAL );

printCsr(m, n, nnzA, descrA, csrValA, csrRowPtrA, csrColIndA, "A");
```

www.nvidia.com
cuSPARSE Library

DU-06709-001_v10.1 | 356
Appendix E: Examples of prune

```c
... 

cudaStat1 = cudaMalloc ((void**)&d_csrRowPtrA, sizeof(int)*(m+1) );
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMalloc ((void**)&d_csrColIndA, sizeof(int)*nnzA );
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMalloc ((void**)&d_csrValA , sizeof(float)*nnzA );
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMalloc ((void**)&d_csrRowPtrC, sizeof(int)*(m+1) );
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(d_csrRowPtrA, csrRowPtrA, sizeof(int)*(m+1),
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(d_csrColIndA, csrColIndA, sizeof(int)*nnzA,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(d_csrValA , csrValA , sizeof(float)*nnzA,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

/* step 3: query workspace */
status = cusparseSpruneCsr2csrByPercentage_bufferSizeExt(
handle,
  m, 
nnzA, 
descrA,
  d_csrValA,
  d_csrRowPtrA,
  d_csrColIndA,
  percentage,
  descrC,
  d_csrValC,
  d_csrRowPtrC,
  d_csrColIndC,
  info,
  &lworkInBytes);
assert(CUSPARSE_STATUS_SUCCESS == status);

printf("lworkInBytes = %lld \n", (long long)lworkInBytes);

if (NULL != d_work) { cudaFree(d_work); }
cudaStat1 = cudaMalloc((void**)&d_work, lworkInBytes);
assert(cudaSuccess == cudaStat1);

/* step 4: compute csrRowPtrC and nnzC */
status = cusparseSpruneCsr2csrNnzByPercentage(
handle,
  m, 
nnzA, 
descrA,
  d_csrValA,
  d_csrRowPtrA,
  d_csrColIndA,
  percentage,
  descrC,
  d_csrRowPtrC,
  &nnzC, /* host */
  info,
  d_work);
```

www.nvidia.com

cuSPARSE Library
assert(CUSPARSE_STATUS_SUCCESS == status);
cudaStat1 = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat1);

printf("nnzC = %d\n", nnzC);
if (0 == nnzC ){
    printf("C is empty \n");
    return 0;
}

/* step 5: compute csrColIndC and csrValC */
cudaStat1 = cudaMalloc ((void**)&d_csrColIndC, sizeof(int  ) * nnzC );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_csrValC   , sizeof(float) * nnzC );
assert(cudaSuccess == cudaStat1);

status = cusparseSpruneCsr2csrByPercentage(  
    handle,  
    m,  
    n,  
    nnzA,  
    descrA,  
    d_csrValA,  
    d_csrRowPtrA,  
    d_csrColIndA,  
    percentage,  
    descrC,  
    d_csrValC,  
    d_csrRowPtrC,  
    d_csrColIndC,  
    info,  
    d_work);
assert(CUSPARSE_STATUS_SUCCESS == status);
cudaStat1 = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat1);

/* step 6: output C */

csrRowPtrC = (int  *)(void*)malloc(sizeof(int  )*(m+1));
csrColIndC = (int  *)(void*)malloc(sizeof(int  )*nnzC);
csrValC   = (float  *)(void*)malloc(sizeof(float)*nnzC);
assert( NULL != csrRowPtrC);
assert( NULL != csrColIndC);
assert( NULL != csrValC);

cudaStat1 = cudaMemcpy(csrRowPtrC, d_csrRowPtrC, sizeof(int  )*(m+1),  
cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(csrColIndC, d_csrColIndC, sizeof(int  )*nnzC ,  
cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(csrValC   , d_csrValC   , sizeof(float)*nnzC ,  
cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);

printCsr(m, n, nnzC, descrC, csrValC, csrRowPtrC, csrColIndC, "C");
Appendix E: Examples of prune

```c
/* free resources */
if (d_csrRowPtrA) cudaFree(d_csrRowPtrA);
if (d_csrColIndA) cudaFree(d_csrColIndA);
if (d_csrValA) cudaFree(d_csrValA);
if (d_csrRowPtrC) cudaFree(d_csrRowPtrC);
if (d_csrColIndC) cudaFree(d_csrColIndC);
if (d_csrValC) cudaFree(d_csrValC);
if (csrRowPtrC) free(csrRowPtrC);
if (csrColIndC) free(csrColIndC);
if (csrValC) free(csrValC);
if (handle) cusparseDestroy(handle);
if (stream) cudaStreamDestroy(stream);
if (descrA) cusparseDestroyMatDescr(descrA);
if (descrC) cusparseDestroyMatDescr(descrC);
if (info) cusparseDestroyPruneInfo(info);
cudaDeviceReset();
return 0;
```
19.1. batched tridiagonal solver

This section provides a simple example in the C programming language of `gtsvInterleavedBatch`.

The example solves two linear systems and assumes data layout is NOT interleaved format. Before calling `gtsvInterleavedBatch`, `cublasXgeam` is used to transform the data layout, from aggregate format to interleaved format. If the user can prepare interleaved format, no need to transpose the data.
Appendix F: Examples of gtsv

/* How to compile (assume cuda is installed at /usr/local/cuda/)
   nvcc -c -I/usr/local/cuda/include gtsv.cpp
   g++ -o gtsv gtsv.o -L/usr/local/cuda/lib64 -lcusparse -lcublas -lcudart
*/
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusparse.h>
#include <cublas_v2.h>

/*
 * compute \| b - A*x \|_\infty
 */
void residual_eval(
   int n,
   const float *dl,
   const float *d,
   const float *du,
   const float *b,
   const float *x,
   float *r_nrminf_ptr)
{
   float r_nrminf = 0;
   for(int i = 0 ; i < n ; i++){
      float dot = 0;
      if (i > 0 ){
         dot += dl[i]*x[i-1];
      }
      dot += d[i]*x[i];
      if (i < (n-1) ){
         dot += du[i]*x[i+1];
      }
      float ri = b[i] - dot;
      r_nrminf = (r_nrminf > fabs(ri))? r_nrminf : fabs(ri);
   }
   *r_nrminf_ptr = r_nrminf;
}

int main(int argc, char*argv[])
{
   cusparseHandle_t cusparseH = NULL;
   cublasHandle_t cublasH = NULL;
   cudaStream_t stream = NULL;
}
cusparseStatus_t status = CUSPARSE_STATUS_SUCCESS;
cublasStatus_t cublasStat = CUBLAS_STATUS_SUCCESS;
cudaError_t cudaStat1 = cudaSuccess;
const int n = 3;
const int batchSize = 2;
/*
  | 1  6  0 | 1 | -0.603960 |
  A1 =| 4  2  7 |, b1 =| 2 |, x1 =| 0.267327 |
  | 0  5  3 | 3 | 0.554455 |
  *
  | 8 13  0 | 4 | -0.063291 |
  A2 =| 11  9 14 |, b2 =| 5 |, x2 =| 0.346641 |
  | 0 12 10 | 6 | 0.184031 |
*/

/* A = (d1, d, du), B and X are in aggregate format */
const float d1[n * batchSize] = {0, 4, 5, 0, 11, 12};
const float d[n * batchSize] = {1, 2, 3, 8, 9, 10};
const float du[n * batchSize] = {6, 7, 0, 13, 14, 0};
const float B[n * batchSize] = {1, 2, 3, 4, 5, 6};
float X[n * batchSize]; /* Xj = Aj \ Bj */

/* device memory */
(d_d10, d_d0, d_du0) is aggregate format
(d_d1, d_d, d_du) is interleaved format

float *d_d10 = NULL;
float *d_d0 = NULL;
float *d_du0 = NULL;
float *d_d1 = NULL;
float *d_d = NULL;
float *d_du = NULL;
float *d_B = NULL;
float *d_X = NULL;

size_t lworkInBytes = 0;
char *d_work = NULL;

/* algo = 0: cuThomas (unstable)
algo = 1: LU with pivoting (stable)
algo = 2: QR (stable)
*/
const int algo = 2;

const float h_one = 1;
const float h_zero = 0;

printf("example of gtsv (interleaved format) \n");
printf("choose algo = 0,1,2 to select different algorithms \n");
printf("n = %d, batchSize = %d, algo = %d \n", n, batchSize, algo);
/* step 1: create cusparse/cublas handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);
status = cusparseCreate(&cusparseH);
assert(CUSPARSE_STATUS_SUCCESS == status);
status = cusparseSetStream(cusparseH, stream);
assert(CUSPARSE_STATUS_SUCCESS == status);
cublasStat = cublasCreate(&cublasH);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
cublasStat = cublasStream(cublasH, stream);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

/* step 2: allocate device memory */
cudaStat1 = cudaMalloc ((void**)&d_dl0 , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_d0  , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_du0 , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_dl  , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_d   , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_du  , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_B   , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_X   , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);

cudaDeviceSynchronize();

/* step 3: prepare data in device, interleaved format */
cudaStat1 = cudaMemcpy(d_d10, dl, sizeof(float)*n*batchSize, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_d0 , d , sizeof(float)*n*batchSize, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_du0, du, sizeof(float)*n*batchSize, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_B  ,  B, sizeof(float)*n*batchSize, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_X  ,  X, sizeof(float)*n*batchSize, cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

cudaDeviceSynchronize();

/* convert dl to interleaved format */

/* convert dl to interleaved format */

cublasH,
CUBLAS_OP_T, /* transa */
CUBLAS_OP_T, /* transb, don't care */
batchSize, /* number of rows of dl */
n, /* number of columns of dl */
&h_one,
d_d10, /* dl is n-by-batchSize */
n, /* leading dimension of dl */
&h_zero,
NULL,
n, /* don't care */
d_dl, /* dl is batchSize-by-n */
batchSize /* leading dimension of dl */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
/* convert d to interleaved format
 *  d = transpose(d0)
 */
cublasStat = cublasSgeam{
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of d */
    n, /* number of columns of d */
    &h_one,
    d_d0, /* d0 is n-by-batchSize */
    n, /* leading dimension of d0 */
    &h_zero,
    NULL,
    n, /* don't care */
    d_d, /* d is batchSize-by-n */
    batchSize /* leading dimension of d */
};
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

/* convert du to interleaved format
 *  du = transpose(du0)
 */
cublasStat = cublasSgeam{
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of du */
    n, /* number of columns of du */
    &h_one,
    d_du0, /* du0 is n-by-batchSize */
    n, /* leading dimension of du0 */
    &h_zero,
    NULL,
    n, /* don't care */
    d_du, /* du is batchSize-by-n */
    batchSize /* leading dimension of du */
};
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

/* convert B to interleaved format
 *  X = transpose(B)
 */
cublasStat = cublasSgeam{
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of X */
    n, /* number of columns of X */
    &h_one,
    d_B, /* B is n-by-batchSize */
    n, /* leading dimension of B */
    &h_zero,
    NULL,
    n, /* don't care */
    d_X, /* X is batchSize-by-n */
    batchSize /* leading dimension of X */
};
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
/* step 4: prepare workspace */
status = cusparseSgtsvInterleavedBatch_bufferSizeExt(
cusparseH,
algo,
n,
d_dl,
d_d,
d_du,
d_X,
batchSize,
lworkInBytes);
assert(CUSPARSE_STATUS_SUCCESS == status);
printf("lworkInBytes = %lld \n", (long long)lworkInBytes);
cudaStat1 = cudaMalloc((void**)&d_work, lworkInBytes);
assert(cudaSuccess == cudaStat1);
/* step 5: solve A_j*x_j = b_j */
status = cusparseSgtsvInterleavedBatch(
cusparseH,
algo,
n,
d_dl,
d_d,
d_du,
d_X,
batchSize,
d_work);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSPARSE_STATUS_SUCCESS == status);
assert(cudaSuccess == cudaStat1);
/* step 6: convert X back to aggregate format */
/* B = transpose(X) */
cublasStat = cublasSgeam(
cublasH,
CUBLAS_OP_T, /* transa */
CUBLAS_OP_T, /* transb, don't care */
n, /* number of rows of B */
batchSize, /* number of columns of B */
&h_one,
d_X, /* X is batchSize-by-n */
batchSize, /* leading dimension of X */
&h_zero,
NULL,
n, /* don't care */
d_B, /* B is n-by-batchSize */
n /* leading dimension of B */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
cudaDeviceSynchronize();
/* step 7: residual evaluation */
cudaStat1 = cudaMemcpy(X, d_B, sizeof(float)*n*batchSize,
cudamemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);
cudaDeviceSynchronize();
```c
... printf("==== x1 = inv(A1)*b1 \n");
for(int j = 0 ; j < n; j++){
    printf("x1[%d] = %f\n", j, X[j]);
}
float r1_nrminf;
residaul_eval(
    n,
    d1,
    d,
    du,
    B,
    X,
    &r1_nrminf
); printf("|b1 - A1*x1| = %E\n", r1_nrminf);

printf("\n==== x2 = inv(A2)*b2 \n");
for(int j = 0 ; j < n; j++){
    printf("x2[%d] = %f\n", j, X[n+j]);
}
float r2_nrminf;
residaul_eval(
    n,
    d1 + n,
    d + n,
    du + n,
    B + n,
    X + n,
    &r2_nrminf
); printf("|b2 - A2*x2| = %E\n", r2_nrminf);
/* free resources */
if (d_d0       ) cudaFree(d_d0);
if (d_d        ) cudaFree(d_d);
if (d_du0      ) cudaFree(d_du0);
if (d_d1       ) cudaFree(d_d1);
if (d_du       ) cudaFree(d_du);
if (d_B        ) cudaFree(d_B);
if (d_X        ) cudaFree(d_X);
if (cusparseH   ) cusparseDestroy(cusparseH);
if (cublasH     ) cublasDestroy(cublasH);
if (stream      ) cudaStreamDestroy(stream);
cudaDeviceReset();
    return 0;
}
```
20.1. batched penta-diagonal solver

This section provides a simple example in the C programming language of gpsvInterleavedBatch.

The example solves two penta-diagonal systems and assumes data layout is NOT interleaved format. Before calling gpsvInterleavedBatch, cublasXgeam is used to transform the data layout, from aggregate format to interleaved format. If the user can prepare interleaved format, no need to transpose the data.
Appendix G: Examples of gsv

* How to compile (assume cuda is installed at /usr/local/cuda/)
  * nvcc -c -I/usr/local/cuda/include gsv.cpp
  * g++ -o gsv gsv.o -L/usr/local/cuda/lib64 -lcusparse -lcublas -lcudart
  */
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusparse.h>
#include <cublas_v2.h>

/*
 * compute | b - A*x|_inf
 */
void residual_eval(
  int n,
  const float *ds,
  const float *dl,
  const float *d,
  const float *du,
  const float *dw,
  const float *b,
  const float *x,
  float *r_nrminf_ptr)
{
  float r_nrminf = 0;
  for(int i = 0 ; i < n ; i++){
    float dot = 0;
    if (i > 1 ){
      dot += ds[i]*x[i-2];
    }
    if (i > 0 ){
      dot += dl[i]*x[i-1];
    }
    dot += d[i]*x[i];
    if (i < (n-1) ){
      dot += du[i]*x[i+1];
    }
    if (i < (n-2) ){
      dot += dw[i]*x[i+2];
    }
    float ri = b[i] - dot;
    r_nrminf = (r_nrminf > fabs(ri))? r_nrminf : fabs(ri);
  }
  *r_nrminf_ptr = r_nrminf;
}

int main(int argc, char*argv[])
{
  cusparseHandle_t cusparseH = NULL;
  cublasHandle_t cublasH = NULL;
  cudaStream_t stream = NULL;
...
Appendix G: Examples of gspv

```c

cusparseStatus_t status = CUSPARSE_STATUS_SUCCESS;
cublasStatus_t cublasStat = CUBLAS_STATUS_SUCCESS;
cudaError_t cudaStat1 = cudaSuccess;

const int n = 4;
const int batchSize = 2;

/*
 *      |  1    8   13   0  |       | 1 |       | -0.0592 |
 * A1 =|  5    2    9   14 |, b1 = | 2 | , x1 = | 0.3428 |
 *      | 11   6    3   10 |       | 3 |       | -1.1295 |
 *      |  0   12   7    4 |       | 4 |       | 0.1982 |
 *
 *      | 15   22  27   0  |       | 5 |       | -0.0012 |
 * A2 =| 19   16   23   28 |, b2 = | 6 | , x2 = | 0.2792 |
 *      | 25   20  17   24 |       | 7 |       | 0.0416 |
 *      |  0  26  21   18 |       | 8 |       | 0.0898 |
 */

/*
 * A = (ds, dl, d, du, dw), B and X are in aggregate format
 */
const float ds[n * batchSize] = {0, 0, 11, 12, 0, 0, 25, 26};
const float dl[n * batchSize] = {0, 5, 6, 7, 0, 19, 20, 21};
const float du[n * batchSize] = {1, 2, 3, 4, 15, 16, 17, 18};
const float dw[n * batchSize] = {9, 10, 0, 22, 23, 24, 0, 0};
const float B[n * batchSize] = {1, 2, 3, 4, 5, 6, 7, 8};
float X[n * batchSize]; /* Xj = Aj Bj */

/* device memory
 * (d_ds0, d_dl0, d_d0, d_du0, d_dw0) is aggregate format
 * (d_ds, d_dl, d_d, d_du, d_dw) is interleaved format
 */
float *d_ds0 = NULL;
float *d_dl0 = NULL;
float *d_d0 = NULL;
float *d_du0 = NULL;
float *d_dw0 = NULL;
float *d_ds = NULL;
float *d_dl = NULL;
float *d_d = NULL;
float *d_du = NULL;
float *d_dw = NULL;
float *d_B = NULL;
float *d_X = NULL;

size_t lworkInBytes = 0;
char *d_work = NULL;

const float h_one = 1;
const float h_zero = 0;

int algo = 0 ; /* QR factorization */
printf("example of gspv (interleaved format) \n");
printf("n = %d, batchSize = %d\n", n, batchSize);

/* step 1: create cusparse/cublas handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);
status = cusparseCreate(&cusparseH);
assert(CUSPARSE_STATUS_SUCCESS == status);
```
status = cusparseSetStream(cusparseH, stream);
assert(CUSPARSE_STATUS_SUCCESS == status);
cublasStat = cublasCreate(&cublasH);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
cublasStat = cublasSetStream(cublasH, stream);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
/* step 2: allocate device memory */
cudaStat1 = cudaMalloc (&d_d0 , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc (&d_d1 , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc (&d_d  , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc (&d_du , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc (&d_dw , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc (&d_B  , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc (&d_X  , sizeof(float)*n*batchSize );
assert(cudaSuccess == cudaStat1);
/* step 3: prepare data in device, interleaved format */
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaMemcpyHostToDevice();
assert(cudaSuccess == cudaStat1);
cudaDeviceSynchronize();
/* convert ds to interleaved format */
cublasStat = cublasSgeam(
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of ds */
    n, /* number of columns of ds */
    &h_one,
    d_d0, /* ds is n-by-batchSize */
    n, /* leading dimension of ds0 */
    &h_zero,
    NULL,
    n, /* don't care */
    d_d , /* ds is batchSize-by-n */
    batchSize); /* leading dimension of ds */
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
/* convert dl to interleaved format
* dl = transpose(dl0)
*/
cublasStat = cublasSgeam(
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of dl */
    n, /* number of columns of dl */
    &h_one,
    d_dl0, /* dl0 is n-by-batchSize */
    n, /* leading dimension of dl0 */
    &h_zero,
    NULL,
    n, /* don't care */
    d_dl, /* dl is batchSize-by-n */
    batchSize /* leading dimension of dl */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

/* convert d to interleaved format
* d = transpose(d0)
*/
cublasStat = cublasSgeam(
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of d */
    n, /* number of columns of d */
    &h_one,
    d_d0, /* d0 is n-by-batchSize */
    n, /* leading dimension of d0 */
    &h_zero,
    NULL,
    n, /* don't care */
    d_d, /* d is batchSize-by-n */
    batchSize /* leading dimension of d */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

/* convert du to interleaved format
* du = transpose(du0)
*/
cublasStat = cublasSgeam(
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of du */
    n, /* number of columns of du */
    &h_one,
    d_du0, /* du0 is n-by-batchSize */
    n, /* leading dimension of du0 */
    &h_zero,
    NULL,
    n, /* don't care */
    d_du, /* du is batchSize-by-n */
    batchSize /* leading dimension of du */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
Appendix G: Examples of gpsi

/* convert dw to interleaved format 
 * dw = transpose(dw0) */
cublasStat = cublasSgeam(
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of dw */
    n, /* number of columns of dw */
    &h_one,
    d_dw0, /* dw0 is n-by-batchSize */
    n, /* leading dimension of dw0 */
    &h_zero,
    NULL,
    n, /* don't care */
    d_dw, /* dw is batchSize-by-n */
    batchSize /* leading dimension of dw */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

/* convert B to interleaved format 
 * X = transpose(B) */
cublasStat = cublasSgeam(
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    batchSize, /* number of rows of X */
    n, /* number of columns of X */
    &h_one,
    d_B, /* B is n-by-batchSize */
    n, /* leading dimension of B */
    &h_zero,
    NULL,
    n, /* don't care */
    d_X, /* X is batchSize-by-n */
    batchSize /* leading dimension of X */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);

/* step 4: prepare workspace */
status = cusparseSgpsvInterleavedBatch_bufferSizeExt(
    cusparseH,
    algo,
    n,
    d_ds,
    d_dl,
    d_d,
    d_du,
    d_dw,
    d_X,
    batchSize,
    &lworkInBytes);
assert(CUSPARSE_STATUS_SUCCESS == status);

printf("lworkInBytes = %lld \n", (long long)lworkInBytes);
cudaStat1 = cudaMalloc((void**)&d_work, lworkInBytes);
assert(cudaSuccess == cudaStat1);
Appendix G: Examples of gpsv

.../* step 5: solve A\(j\)x\(j\) = b\(j\) */
status = cusparseSgpsvInterleavedBatch(
    cusparseH,
    algo,
    n,
    d_ds,
    d_dl,
    d_d,
    d_du,
    d_dw,
    d_X,
    batchSize,
    d_work);
cudaStat1 = cudaDeviceSynchronize();
assert(CUSPARSE_STATUS_SUCCESS == status);
assert(cudaSuccess == cudaStat1);

/* step 6: convert X back to aggregate format */
/* B = transpose(X) */
cublasStat = cublasSgeam(
    cublasH,
    CUBLAS_OP_T, /* transa */
    CUBLAS_OP_T, /* transb, don't care */
    n, /* number of rows of B */
    batchSize, /* number of columns of B */
    &h_one,
    d_X, /* X is batchSize-by-n */
    batchSize, /* leading dimension of X */
    &h_zero,
    NULL,
    n, /* don't care */
    d_B, /* B is n-by-batchSize */
    n /* leading dimension of B */
);
assert(CUBLAS_STATUS_SUCCESS == cublasStat);
cudaDeviceSynchronize();

/* step 7: residual evaluation */
cudaStat1 = cudaMemcpy(X, d_B, sizeof(float)*n*batchSize,
    cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);
cudaDeviceSynchronize();

printf("==== x1 = inv(A1)*b1 \n");
for(int j = 0 ; j < n; j++){
    printf("x1[%d] = %f\n", j, X[j]);
}

float r1_nrminf;
residual_eval(
    n,
    ds,
    dl,
    d,
    du,
    dw,
    B,
    X,
    &r1_nrminf
);
printf("|b1 - A1*x1| = %E\n", r1_nrminf);
Appendix G: Examples of gpusv

```c
...;

  float r2_nrminf;
  residual_eval(
  n,
  ds + n,
  dl + n,
  d + n,
  du + n,
  dw + n,
  B + n,
  X + n,
  &r2_nrminf
  );

  printf("\n|b2 - A2*x2| = %E\n", r2_nrminf);

  free resources */
  if (d_ds0 ) cudaFree(d_ds0);
  if (d_d10 ) cudaFree(d_d10);
  if (d_d0  ) cudaFree(d_d0);
  if (d_du0  ) cudaFree(d_du0);
  if (d_dw0  ) cudaFree(d_dw0);
  if (d_ds  ) cudaFree(d_ds);
  if (d_d1  ) cudaFree(d_d1);
  if (d_d   ) cudaFree(d_d);
  if (d_du  ) cudaFree(d_du);
  if (d_dw  ) cudaFree(d_dw);
  if (d_B   ) cudaFree(d_B);
  if (d_X   ) cudaFree(d_X);

  if (cusparseH ) cusparseDestroy(cusparseH);
  if (cublasH    ) cublasDestroy(cublasH);
  if (stream     ) cudaStreamDestroy(stream);

  cudaDeviceReset();

  return 0;
}
```
21.1. forward triangular solver

This section provides a simple example in the C programming language of csrsm2. The example solves a lower triangular system with 2 right hand side vectors.
Appendix H: Examples of csrsm2

/* How to compile (assume cuda is installed at /usr/local/cuda/)
   nvcc -c -I/usr/local/cuda/include csrms2.cpp
   g++ -o csrm2 csrsm2.o -L/usr/local/cuda/lib64 -lcusparse -lcudart
*/

#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include <cuda_runtime.h>
#include <cusparse.h>

/* compute | b - A*x|_inf */
void residual_eval(
    int n,
    const cusparseMatDescr_t descrA,
    const float *csrVal,
    const int *csrRowPtr,
    const int *csrColInd,
    const float *b,
    const float *x,
    float *r_nrminf_ptr)
{
    const int base = (cusparseGetMatIndexBase(descrA) !=
        CUSPARSE_INDEX_BASE_ONE)? 0:1 ;
    const int lower = (CUSPARSE_FILL_MODE_LOWER ==
        cusparseGetMatFillMode(descrA))? 1:0;
    const int unit  = (CUSPARSE_DIAG_TYPE_UNIT ==
        cusparseGetMatDiagType(descrA))? 1:0;

    float r_nrminf = 0;
    for(int row = 0 ; row < n ; row++){
        const int start = csrRowPtr[row] - base;
        const int end   = csrRowPtr[row+1] - base;
        float dot = 0;
        for(int colidx = start ; colidx < end; colidx++){
            const int col = csrColInd[colidx] - base;
            float Aij = csrVal[colidx];
            float xj  = x[col];
            if ( (row == col) && unit ){
                Aij = 1.0;
            }
            if ( valid ){
                dot += Aij*xj;
            }
        }
        float ri = b[row] - dot;
        r_nrminf = (r_nrminf > fabs(ri))? r_nrminf : fabs(ri);
    }
    *r_nrminf_ptr = r_nrminf;
}

int main(int argc, char*argv[]) {
    cusparseHandle_t handle = NULL;
    cudaStream_t stream = NULL;
    cusparseMatDescr_t descrA = NULL;
    csrsm2Info_t info = NULL;

    ...
Appendix H: Examples of csrsm2

cusparseStatus_t status = CUSPARSE_STATUS_SUCCESS;
cudaError_t cudaStat1 = cudaSuccess;
const int nrhs = 2;
const int n = 4;
const int nnzA = 9;
const cusparseSolvePolicy_t policy = CUSPARSE_SOLVE_POLICY_NO_LEVEL;
const float h_one = 1.0;

\[
\begin{array}{cccc}
1 & 0 & 2 & -3 \\
0 & 4 & 0 & 0 \\
5 & 0 & 6 & 7 \\
0 & 8 & 0 & 9 \\
\end{array}
\]

* Regard A as a lower triangle matrix L with non-unit diagonal.
* Given \( B = \begin{bmatrix} 2 & 6 \\ 3 & 7 \\ 4 & 8 \end{bmatrix} \), \( X = L \backslash B = \begin{bmatrix} 0.5 & 1.5 \\ -0.3333 & -3 \\ 0 & -0.4444 \end{bmatrix} \)

\[
\begin{array}{cccc}
1 & 5 & 6 & 7 \\
2 & 4 & 8 & 9 \\
3 & 7 & 8 & 9 \\
4 & 8 & 9 & 0 \\
\end{array}
\]

const int csrRowPtrA[n+1] = { 1, 4, 5, 8, 10};
const int csrColIndA[nnzA] = { 1, 3, 4, 2, 1, 3, 4, 2, 4};
const float csrValA[nnzA] = {1, 2, -3, 4, 5, 6, 7, 8, 9};
const float B[n*nrhs] = {1,2,3,4,5,6,7,8};
float X[n*nrhs];

int *d_csrRowPtrA = NULL;
int *d_csrColIndA = NULL;
float *d_csrValA = NULL;
float *d_B = NULL;

size_t lworkInBytes = 0;
char *d_work = NULL;

const int algo = 0; /* non-block version */

printf("example of csrsm2 \n");

/* step 1: create cusparse handle, bind a stream */
cudaStat1 = cudaStreamCreateWithFlags(&stream, cudaStreamNonBlocking);
assert(cudaSuccess == cudaStat1);

status = cusparseCreate(&handle);
assert(CUSPARSE_STATUS_SUCCESS == status);

status = cusparseSetStream(handle, stream);
assert(CUSPARSE_STATUS_SUCCESS == status);

status = cusparseCreateCsrsm2Info(&info);
assert(CUSPARSE_STATUS_SUCCESS == status);

/* step 2: configuration of matrix A */

status = cusparseCreateMatDescr(&descrA);
assert(CUSPARSE_STATUS_SUCCESS == status);

/* A is base-1*/
cusparseSetMatIndexBase(descrA, CUSPARSE_INDEX_BASE_ONE);
cusparseSetMatType(descrA, CUSPARSE_MATRIX_TYPE_GENERAL);
/* A is lower triangle */
cusparseSetMatFillMode(descrA, CUSPARSE_FILL_MODE_LOWER);
/* A has non unit diagonal */
cusparseSetMatDiagType(descrA, CUSPARSE_DIAG_TYPE_NON_UNIT);
Appendix H: Examples of csrsm2

... cudaStat1 = cudaMalloc ((void**)&d_csrRowPtrA, sizeof(int)*(n+1) );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_csrColIndA, sizeof(int)*nnzA );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_csrValA , sizeof(float)*nnzA );
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMalloc ((void**)&d_B , sizeof(float)*n*nrhs );
assert(cudaSuccess == cudaStat1);

cudaStat1 = cudaMemcpy(d_csrRowPtrA, csrRowPtrA, sizeof(int)*(n+1),
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_csrColIndA, csrColIndA, sizeof(int)*nnzA,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_csrValA , csrValA , sizeof(float)*nnzA,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);
cudaStat1 = cudaMemcpy(d_B , B , sizeof(float)*n*nrhs,
cudaMemcpyHostToDevice);
assert(cudaSuccess == cudaStat1);

/* step 3: query workspace */
status = cusparseScsrsm2_bufferSizeExt(
    handle,
    algo,
    CUSPARSE_OPERATION_NON_TRANSPOSE, /* transA */
    CUSPARSE_OPERATION_NON_TRANSPOSE, /* transB */
    n,
    nrhs,
    nnzA,
    &h_one,
    descrA,
    d_csrValA,
    d_csrRowPtrA,
    d_csrColIndA,
    d_B,
    n, /* ldb */
    info,
    policy,
    &lworkInBytes);
assert(CUSPARSE_STATUS_SUCCESS == status);

printf("lworkInBytes = %lld \n", (long long)lworkInBytes);
if (NULL != d_work) { cudaFree(d_work); }
cudaStat1 = cudaMalloc((void**)&d_work, lworkInBytes);
assert(cudaSuccess == cudaStat1);

/* step 4: analysis */
status = cusparseScsrsm2_analysis(
    handle,
    algo,
    CUSPARSE_OPERATION_NON_TRANSPOSE, /* transA */
    CUSPARSE_OPERATION_NON_TRANSPOSE, /* transB */
    n,
    nrhs,
    nnzA,
    &h_one,
    descrA,
    d_csrValA,
    d_csrRowPtrA,
    d_csrColIndA,
    d_B,
    n, /* ldb */
    info,
    policy,
    d_work);
assert(CUSPARSE_STATUS_SUCCESS == status);
/* step 5: solve L * X = B */
status = cusparseScsrm2_solve(
    handle,
    algo,
    CUSPARSE_OPERATION_NON_TRANSPOSE, /* transA */
    CUSPARSE_OPERATION_NON_TRANSPOSE, /* transB */
    n,
    nrhs,
    nnzA,
    &h_one,
    descrA,
    d_csrValA,
    d_csrRowPtrA,
    d_csrColIndA,
    d_B,
    n, /* ldb */
    info,
    policy,
    d_work);
assert(CUSPARSE_STATUS_SUCCESS == status);
cudaStat1 = cudaDeviceSynchronize();
assert(cudaSuccess == cudaStat1);
/* step 6: measure residual B - A*X */
cudaStat1 = cudaMemcpy(X, d_B, sizeof(float)*n*nrhs,
cudaMemcpyDeviceToHost);
assert(cudaSuccess == cudaStat1);
cudaDeviceSynchronize();

printf("==== x1 = inv(A)*b1 \n");
for(int j = 0 ; j < n; j++){
    printf("x1[%d] = %f\n", j, X[j]);
}
float r1_nrminf;
residaul_eval(
    n,
    descrA,
    csrValA,
    csrRowPtrA,
    csrColIndA,
    B,
    X,
    &r1_nrminf
);
printf("|b1 - A*x1| = %E\n", r1_nrminf);

printf("==== x2 = inv(A)*b2 \n");
for(int j = 0 ; j < n; j++){
    printf("x2[%d] = %f\n", j, X[n+j]);
}
float r2_nrminf;
residaul_eval(
    n,
    descrA,
    csrValA,
    csrRowPtrA,
    csrColIndA,
    B+n,
    X+n,
    &r2_nrminf
);
printf("|b2 - A*x2| = %E\n", r2_nrminf);
Appendix H: Examples of csrsm2

/* free resources */
    if (d_csrRowPtrA) cudaFree(d_csrRowPtrA);
    if (d_csrColIndA) cudaFree(d_csrColIndA);
    if (d_csrValA) cudaFree(d_csrValA);
    if (d_B) cudaFree(d_B);

    if (handle) cusparseDestroy(handle);
    if (stream) cudaStreamDestroy(stream);
    if (descrA) cusparseDestroyMatDescr(descrA);
    if (info) cusparseDestroyCsrsm2Info(info);

    cudaDeviceReset();

    return 0;
}
NVIDIA would like to thank the following individuals and institutions for their contributions:

- The cusparse<t>gtsv implementation is derived from a version developed by Li-Wen Chang from the University of Illinois.
- the cusparse<t>gtsvInterleavedBatch adopts cuThomasBatch developed by Pedro Valero-Lara and Ivan Martínez-Pérez from Barcelona Supercomputing Center and BSC/UPC NVIDIA GPU Center of Excellence.
Chapter 23.
BIBLIOGRAPHY


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

© 2007-2019 NVIDIA Corporation. All rights reserved.