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Chapter 1.  
CUDNN OVERVIEW

NVIDIA cuDNN is a GPU-accelerated library of primitives for deep neural networks. It provides highly tuned implementations of routines arising frequently in DNN applications:

- Convolution forward and backward, including cross-correlation
- Pooling forward and backward
- Softmax forward and backward
- Neuron activations forward and backward:
  - Rectified linear (ReLU)
  - Sigmoid
  - Hyperbolic tangent (TANH)
- Tensor transformation functions
- LRN, LCN and batch normalization forward and backward

cuDNN’s convolution routines aim for performance competitive with the fastest GEMM (matrix multiply) based implementations of such routines while using significantly less memory.

cuDNN features customizable data layouts, supporting flexible dimension ordering, striding, and subregions for the 4D tensors used as inputs and outputs to all of its routines. This flexibility allows easy integration into any neural network implementation and avoids the input/output transposition steps sometimes necessary with GEMM-based convolutions.

cuDNN offers a context-based API that allows for easy multi-threading and (optional) interoperability with CUDA streams.
Chapter 2.
CUDNN RELEASE NOTES V7.1.4

Key Features and Enhancements
The following enhancements have been added to this release:

‣ Improved performance for some cases of data-gradient convolutions and maxpooling. This is expected to improve performance of ResNet-50 like networks.
‣ The runtime of the RNN Find algorithm suite is improved in v7.1.4 resulting in slightly improved runtime of cudnnFindRNN***AlgorithmEx.

Known Issues
Following are known issues in this release:

‣ cudnnGet picks a slow algorithm that does not use Tensor Cores on Volta when inputs are FP16 and it is possible to do so.
‣ The cudnnConvolutionBackwardFilter() function may output incorrect results for CUDNN_CONVOLUTION_BWD_FILTER_ALGO_FFT_TILING when the convolution mode is CUDNN_CONVOLUTION. This function should not be used in this mode.

Fixed Issues
The following issues have been fixed in this release:

‣ cudnnAddTensorNd might cause a segmentation fault if called with bad arguments (e.g. null pointer), this issue is in 7.1.3 only and fixed in 7.1.4.
‣ cudnnRNNBackwardData LSTM cell with fp16 (half) inputs might generate wrong values (silently), this issue exists in cudnn 7.1.3 binaries compiled with cuda toolkit 9.0 and toolkit cuda 9.2, and does not exist in cudnn 7.1.3 binaries compiled with toolkit 9.1.
‣ cudnnGetRNNLinLayerMatrixParams wrongly returns CUDNN_STATUS_BAD_PARAM when cudnnSetRNNDescriptor is called with dataType == CUDNN_DATA_FLOAT. This is an issue in 7.1.3 only and will be fixed
in 7.1.4. The dataType argument as of today supports only CUDNN_DATA_FLOAT and we plan to support additional compute types in the future.

- There is a small memory leak issue when calling cudnnRNNBackwardData with CUDNN_RNN_ALGO_STANDARD. This issue also affects previous cuDNN v7 releases. This is fixed in 7.1.4.

- RNN with half precision returns CUDNN_EXECUTION_FAILED on Kepler gpu in 7.1.3. This is fixed in 7.1.4 to use pseudo-fp16 computation.

- The RNN Find algorithm suite mistakenly did not test CUDNN_RNN_ALGO_PERSIST_STATIC and CUDNN_RNN_ALGO_PERSIST_DYNAMIC kernels with tensor operations enabled when it was possible to do so. This is fixed in v7.1.4.
Key Features and Enhancements

The following enhancements have been added to this release:

- None.

Known Issues

Following are known issues in this release:

- `cudnnGet` picks a slow algorithm that does not use Tensor Cores on Volta when inputs are FP16 and it is possible to do so.
- The `cudnnConvolutionBackwardFilter()` function may output incorrect results for `CUDNN_CONVOLUTION_BWD_FILTER_ALGO_FFT_TILING` when the convolution mode is `CUDNN_CONVOLUTION` and the product "n*k" (n - batch size, k - number of output feature maps) is large, i.e., several thousand or more. It appears that the `CUDNN_CROSS_CORRELATION` mode is not affected by this bug.
- There is a small memory leak issue when calling `cudnnRNNBackwardData` with `CUDNN_RNN_ALGO_STANDARD`. This issue also affects previous cuDNN v7 releases.
- RNN with half precision will not work on Kepler GPUs and will return `CUDNN_EXECUTION_FAILED`. This will be fixed in future releases to return `CUDNN_STATUS_UNSUPPORTED`.

Fixed Issues

The following issues have been fixed in this release:

- `cudnnRNNBackwardData` for LSTM with recurrent projection in half precision may fail in rare cases with misaligned memory access on Pascal and Maxwell.
- `cudnnRNNBackwardData` for bidirectional LSTM with recurrent projection may produce inaccurate results, or `CUDNN_STATUS_UNSUPPORTED`. 
‣ Algo 1 for forward convolution and dgrad may produce erroneous results when the filter size is greater than the input size. This issue is fixed in 7.1.3.

‣ For very large RNN networks, the function `cudnnGetRNNWorkspaceSize` and `cudnnGetRNNTrainingReserveSize` may internally overflow and give incorrect results.

‣ The small performance regression on multi-layer RNNs using the STANDARD algorithm and Tensor Core math in 7.1.2, as compared to 7.0.5, is fixed in this release.

‣ Fixed an issue with Persistent LSTM backward pass with a hidden state size in the range 257 to 512 on GPUs with number of SMs between 22 and 31 might hang. This issue also exists in 7.1.1. This is fixed in 7.1.3.

‣ Fixed an issue Persistent GRU backward pass with a hidden state size in the range 513->720 on GPUs with exactly 30 SMs would hang. This issue also exists in 7.1.1. This is fixed in 7.1.3.
Chapter 4.
CUDNN RELEASE NOTES V7.1.2

Key Features and Enhancements

The following enhancements have been added to this release:

‣ RNN search API extended to support all RNN algorithms.
‣ Newly added projection Layer supported for inference bidirectional RNN cells and for backward data and gradient.
‣ Support IDENTITY Activation for all cudnnConvolutionBiasActivationForward data types for CUDNN_CONVOLUTION_FWD_ALGO_IMPLICIT_GEMM.
‣ Added documentation to clarify RNN/LSTM weight formats.

Known Issues

Following are known issues in this release:

‣ cudnnGet picks a slow algorithm that does not use Tensor Cores on Volta when inputs are FP16 and it is possible to do so.
‣ There may be a small performance regression on multi-layer RNNs using the STANDARD algorithm with Tensor Core math in this release compared to v7.0.5.
‣ LSTM projection dgrad half precision may fail in rare cases with misaligned memory access on Pascal and Maxwell.
‣ Dgrad for bidirectional LSTM with projection should not be used, may produce inaccurate results, or CUDNN_STATUS_UNSUPPORTED.
‣ The cudnnConvolutionBackwardFilter() function may output incorrect results for CUDNN_CONVOLUTION_BWD_FILTER_ALGO_FFT_TILING when the convolution mode is CUDNN_CONVOLUTION and the product “n*k” (n - batch size, k - number of output feature maps) is large, i.e., several thousand or more. It appears that the CUDNN_CROSS_CORRELATION mode is not affected by this.
Persistent LSTM backward pass with a hidden state size in the range 257 to 512 on GPUs with number of SMs between 22 and 31 might hang. This issue also exists in 7.1.1 and will be fixed in 7.1.3.

Persistent GRU backward pass with a hidden state size in the range 513 to 720 on GPUs with exactly 30 SMs would hang. This issue also exists in 7.1.1 and will be fixed in 7.1.3.

Algo 1 for forward convolution and dgrad may produce erroneous results when the filter size is greater than the input size.

Fixed Issues

The following issues have been fixed in this release:

The uint8 input for convolution is restricted to Volta and later. We added support for older architectures, for algo: CUDNN_CONVOLUTION_FWD_ALGO_IMPLICIT_GEMM.

In some cases when algorithm CUDNN_CONVOLUTION_BWD_FILTER_ALGO1 was selected, the routine cudnnConvolutionBackwardFilter could fail at runtime and return CUDNN_STATUS_EXECUTION_FAILED. It now returns CUDNN_STATUS_NOT_SUPPORTED.

cudnnSetRNNDescriptor no longer needs valid Dropout Descriptor in inference mode, user can pass NULL for Dropout Descriptor in inference mode.
Chapter 5.
CUDNN RELEASE NOTES V7.1.1

Key Features and Enhancements

The following enhancements have been added to this release:

- Added new API `cudnnSetRNNProjectionLayers` and `cudnnGetRNNProjectionLayers` to support Projection Layer for the RNN LSTM cell. In this release only the inference use case will be supported. The bi-directional and the training forward and backward for training is not supported in 7.1.1 but will be supported in the upcoming 7.1.2 release without API changes. For all the unsupported cases in this release, `CUDNN_NOT_SUPPORTED` is returned when projection layer is set and the RNN is called.

- The `cudnnGetRNNLinLayerMatrixParams()` function was enhanced and a bug was fixed without modifying its prototype. Specifically:
  - The `cudnnGetRNNLinLayerMatrixParams()` function was updated to support the RNN projection feature. An extra linLayerID value of 8 can be used to retrieve the address and the size of the “recurrent” projection weight matrix when "mode" in `cudnnSetRNNDescriptor()` is configured to `CUDNN_LSTM` and the recurrent projection is enabled via `cudnnSetRNNProjectionLayers()`.
  - Instead of reporting the total number of elements in each weight matrix in the “linLayerMatDesc” filter descriptor, the `cudnnGetRNNLinLayerMatrixParams()` function returns the matrix size as two dimensions: rows and columns. This allows the user to easily print and initialize RNN weight matrices. Elements in each weight matrix are arranged in the row-major order. Due to historical reasons, the minimum number of dimensions in the filter descriptor is three. In previous versions of the cuDNN library, `cudnnGetRNNLinLayerMatrixParams()` returned the total number of weights as follows: `filterDimA[0]=total_size, filterDimA[1]=1, filterDimA[2]=1`. In v7.1.1, the format was changed to: `filterDimA[0]=1, filterDimA[1]=rows, filterDimA[2]=columns`. In both cases, the
"format" field of the filter descriptor should be ignored when retrieved by `cudnnGetFilterNdDescriptor()`.

- A bug in `cudnnGetRNNLinLayerMatrixParams()` was fixed to return a zeroed filter descriptor when the corresponding weight matrix does not exist. This occurs, for example, for `linLayerID` values of 0-3 when the first RNN layer is configured to exclude matrix multiplications applied to RNN input data (`inputMode=CUDNN_SKIP_INPUT` in `cudnnSetRNNDescriptor()` specifies implicit, fixed identity weight matrices for RNN input). Such cases in previous versions of the cuDNN library caused `cudnnGetRNNLinLayerMatrixParams()` to return corrupted filter descriptors with some entries from the previous call. A workaround was to create a new filter descriptor for every invocation of `cudnnGetRNNLinLayerMatrixParams()`.

- The `cudnnGetRNNLinLayerBiasParams()` function was updated to report the bias column vectors in "linLayerBiasDesc" in the same format as `cudnnGetRNNLinLayerMatrixParams()`. In previous versions of the cuDNN library, `cudnnGetRNNLinLayerBiasParams()` returned the total number of adjustable bias parameters as follows: `filterDimA[0]=total_size, filterDimA[1]=1, filterDimA[2]=1`. In v7.1.1, the format was changed to: `filterDimA[0]=1, filterDimA[1]=rows, filterDimA[2]=1` (number of columns). In both cases, the "format" field of the filter descriptor should be ignored when retrieved by `cudnnGetFilterNdDescriptor()` . The recurrent projection GEMM does not have a bias so the range of valid inputs for the "linLayerID" argument remains the same.

- Added support for use of Tensor Core for the `CUDNN_RNN_ALGO_PERSIST_STATIC`. This required cuda cuDNN v7.1 build with CUDA 9.1 and 387 or higher driver. It will not work with CUDA 9.0 and 384 driver.

- Added RNN search API that allows the application to provide an RNN descriptor and get a list of possible algorithm choices with performance and memory usage, to allow applications to choose between different implementations. For more information, refer to the documentation of: `cudnnFindRNNForwardInferenceAlgorithmEx`, `cudnnFindRNNForwardTrainingAlgorithmEx`, `cudnnFindRNNBackwardDataAlgorithmEx`, and `cudnnFindRNNBackwardWeightsAlgorithmEx`. In this release, the search will operate on STANDARD algorithm and will not support PERSISTENT algorithms of RNN.

- Added `uint8` for support for the input data for `cudnnConvolutionBiasActivationForward` and `cudnnConvolutionForward`. Currently the support is on Volta (sm 70) and later architectures. Support for older architectures will be gradually added in the upcoming releases.
Suport for CUDNN_ACTIVATION_IDENTITY is added to cudnnConvolutionBiasActivationForward. This allows users to perform Convolution and Bias without Activation.

All API functions now support logging. User can trigger logging by setting environment variable “CUDNN_LOGINFO_DBG=1” and “CUDNN_LOGDEST_DBG= <option>” where <option> (i.e., the output destination of the log) can be chosen from “stdout”, “stderr”, or a file path. User may also use the new Set/GetCallBack functions to install their customized callback function. Log files can be added to the reported bugs or shared with us for analysis and future optimizations through partners.nvidia.com.

Improved performance of 3D convolution on Volta architecture.

The following algo-related functions have been added for this release: cudnnGetAlgorithmSpaceSize, cudnnSaveAlgorithm, cudnnRestoreAlgorithm, cudnnCreateAlgorithmDescriptor, cudnnSetAlgorithmDescriptor, cudnnGetAlgorithmDescriptor, cudnnDestroyAlgorithmDescriptor, cudnnCreateAlgorithmPerformance, cudnnSetAlgorithmPerformance, cudnnGetAlgorithmPerformance, cudnnDestroyAlgorithmPerformance.

All algorithms for convolutions now support groupCount > 1. This includes cudnnConvolutionForward(), cudnnConvolutionBackwardData(), and cudnnConvolutionBackwardFilter().

Known Issues

Following are known issues in this release:

- RNN search Algorithm is restricted to STANDARD algorithm.
- Newly added projection Layer supported for inference and one directional RNN cells.
- uint8 input for convolution is restricted to Volta and later.
- cudnnGet picks a slow algorithm that doesn’t use Tensor Cores on Volta when inputs are FP16 and it is possible to do so.
- There may be a small performance regression on multi-layer RNNs using the STANDARD algorithm with Tensor Core math in this release compared to 7.0.5.

Fixed Issues

The following issues have been fixed in this release:

- 3D convolution performance improvements for Volta.
- Added support for Algorithm 0 data gradients to cover cases previously not supported.
- Removed the requirement for dropout Descriptor in RNN inference. Before application had to set a non point for the dropout Descriptor which was not used.
Use of CUDNN_TENSOR_NCHW_VECT_C with non-zero padding resulted in a
return status of CUDNN_STATUS_INTERNAL_ERROR. This issue is now fixed.
**Key Features and Enhancements**

The following enhancements have been added to this release:

- None.

**Known Issues**

Following are known issues in this release:

- cuDNN library may trigger a CPU floating point exception when FP exceptions are enabled by user. This issue exists for all 7.0.x releases.

- There are heavy use cases of RNN layers that might hit a memory allocation issue in the CUDA driver when using cuDNN v7 with CUDA 8.0 and R375 driver on pre-Pascal architectures (Kepler and Maxwell). In these cases, subsequent CUDA kernels may fail to launch with an Error Code 30. To resolve the issue, it is recommended to use the latest R384 driver (from NVIDIA driver downloads) or to ensure that the persistence daemon is started. This behavior is observed on all 7.0.x releases.

- When using TENSOR_OP_MATH mode with `cudnnConvolutionBiasActivationForward`, the pointer to the bias must be aligned to 16 bytes and the size of allocated memory must be multiples of 256 elements. This behavior exists for all 7.0.x releases.

**Fixed Issues**

The following issues have been fixed in this release:

- Corrected the algorithm fallback behavior in RNN when user set to use `CUDNN_TENSOR_OP_MATH` when using compute card without HMMA. Instead of returning `CUDNN_STATUS_NOT_SUPPORTED`, the RNN algorithm will now continue to run using `CUDNN_DEFAULT_MATH`. The correct behavior is to fall back to using default math when Tensor Core is not supported. Fixed to the expected behavior.
On Volta hardware, `BWD_FILTER_ALGO_1` and `BWD_DATA_ALGO_1` convolutions using a number of filter elements greater than 512 were causing `CUDA_ERROR_ILLEGAL_ADDRESS` and `CUDNN_STATUS_INTERNAL_ERROR` errors. Logic was added to fall back to a generic kernel for these filter sizes.

cuDNN v7 with CUDA 8.0 produced erroneous results on Volta for some common cases of Algo 1. Logic was added to fall back to a generic kernel when cudnn v7 with CUDA 8.0 is used on Volta.
Chapter 7.
CUDNN RELEASE NOTES V7.0.4

Key Features and Enhancements

Performance improvements for grouped convolutions when input channels and output channels per group are 1, 2, or 4 for the following algorithms:

- CUDNN_CONVOLUTION_FWD_ALGO_IMPLICIT_GEMM
- CUDNN_CONVOLUTION_BWD_DATA_ALGO0
- CUDNN_CONVOLUTION_BWD_DATA_ALGO_1
- CUDNN_CONVOLUTION_BWD_FILTER_ALGO_0
- CUDNN_CONVOLUTION_BWD_FILTER_ALGO_1

Known Issues

Following are known issues in this release:

- The CUDA 8.0 build of cuDNN may produce incorrect computations when run on Volta.
- cuDNN library triggers CPU floating point exception when FP exceptions are enabled by user. This issue exists for all 7.0.x releases.
- There are heavy use cases of RNN layers that might hit a memory allocation issue in the CUDA driver when using cuDNN v7 with CUDA 8.0 and R375 driver on pre-Pascal architectures (Kepler and Maxwell). In these cases, subsequent CUDA kernels may fail to launch with an Error Code 30. To resolve the issue, it is recommended to use the latest R384 driver (from NVIDIA driver downloads) or to ensure that the persistence daemon is started. This behavior is observed on all 7.0.x releases.
- When using TENSOR_OP_MATH mode with cudnnConvolutionBiasActivationForward, the pointer to the bias must be aligned to 16 bytes and the size of allocated memory must be multiples of 256 elements. This behavior exists for all 7.0.x releases.
Fixed Issues

The following issues have been fixed in this release:

‣ Fixed out-of-band global memory accesses in the 256-point 1D FFT kernel. The problem affected convolutions with 1x1 filters and tall but narrow images, e.g., 1x500 (WxH). In those cases, the workspace size for the FFT_TILING algo was computed incorrectly. There was no error in the FFT kernel.

‣ Eliminated a source of floating point exceptions in the CUDNN_CONVOLUTION_FWD_ALGO_WINOGRAD_NONFUSED algorithm. The host code to generate a negative infinity floating point value was substituted with a different logic. By default, FP exceptions are disabled. However, a user program enabled them by invoking feenableexcept(). There are at least two other sources of FP exceptions in the cuDNN library, affecting for example BATCHNORM_SPATIAL_PERSISTENT. Those sources of FP exceptions will be eliminated in future releases of the cuDNN library.
Chapter 8.
CUDNN RELEASE NOTES V7.0.3

Key Features and Enhancements
Performance improvements for various cases:

- Forward Grouped Convolutions where input channel per groups is 1, 2 or 4 and hardware is Volta or Pascal.
- `cudnnTransformTensor()` where input and output tensor is packed.

This is an improved fallback, improvements will not be seen in all cases.

Known Issues
The following are known issues in this release:

- `CUDNN_CONVOLUTION_FWD_ALGO_FFT_TILING` may cause `CUDA_ERROR_ILLEGAL_ADDRESS`. This issue affects input images of just one 1 pixel in width and certain \( n, c, k, h \) combinations.

Fixed Issues
The following issues have been fixed in this release:

- `AddTensor` and `TensorOp` produce incorrect results for half and INT8 inputs for various use cases.
- `cudnnPoolingBackward()` can produce incorrect values for rare cases of non-deterministic MAX pooling with `window_width > 256`. These rare cases are when the maximum element in a window is duplicated horizontally (along width) by a stride of \( 256 \times k \) for some \( k \). The behavior is now fixed to accumulate derivatives for the duplicate that is left-most.
- `cudnnGetConvolutionForwardWorkspaceSize()` produces incorrect workspace size for algorithm `FFT_TILING` for 1d convolutions. This only occurs for large sized...
convolutions where intermediate calculations produce values greater than $2^{31}$ (2 to the power of 31).

- **CUDNN_STATUS_NOT_SUPPORTED** returned by `cudnnPooling*()` functions for small image (channels * height * width < 4).
Chapter 9.
CUDNN RELEASE NOTES V7.0.2

Key Features and Enhancements

This is a patch release of cuDNN 7.0 and includes bug fixes and performance improvements mainly on Volta.

Algo 1 Convolutions Performance Improvements

Performance improvements were made to
- `CUDNN_CONVOLUTION_FWD_ALGO_IMPLICIT_PRECOMP_GEMM`,
- `CUDNN_CONVOLUTION_BWD_FILTER_ALGO_1`, and
- `CUDNN_CONVOLUTION_BWD_DATA_ALGO_1`. These improvements consist of new SASS kernels and improved heuristics. The new kernels implement convolutions over various data sizes and tile sizes. The improved heuristics take advantage of these new kernels.

Known Issues

The following are known issues in this release:

- `cudnnGetConvolutionForwardWorkspaceSize()` returns overflowed size_t value for certain input shape for `CUDNN_CONVOLUTION_*_ALGO_FFT_TILING`.
- `cudnnPoolingBackward()` fails for pooling window size > 256.

Fixed Issues

The following issues have been fixed in this release:

- Batch Norm `CUDNN_BATCHNORM_SPATIAL_PERSISTENT` might get into race conditions in certain scenarios.
- cuDNN convolution layers using `TENSOR_OP_MATH` with fp16 inputs and outputs and fp32 compute will use “round to nearest” mode instead of “round to zero” mode as in 7.0.1. This rounding mode has proven to achieve better results in training.
• Fixed synchronization logic in the **CUDNN_CTC_LOSS_ALGO_DETERMINISTIC** algo for CTC. The original code would hang in rare cases.

• Convolution algorithms using **TENSOR_OP_MATH** returned a workspace size from *GetWorkspaceSize()* smaller than actually necessary.

• The results of int8 are inaccurate in certain cases when calling **cudnnConvolutionForward()** in convolution layer.

• **cudnnConvolutionForward()** called with **xDesc’s channel = yDesc’s channel = groupCount** could compute incorrect values when vertical padding > 0.
cuDNN v7.0.1 is the first release to support the Volta GPU architecture. In addition, cuDNN v7.0.1 brings new layers, grouped convolutions, and improved convolution find as error query mechanism.

**Key Features and Enhancements**

This cuDNN release includes the following key features and enhancements.

**Tensor Cores**

Version 7.0.1 of cuDNN is the first to support the Tensor Core operations in its implementation. Tensor Cores provide highly optimized matrix multiplication building blocks that do not have an equivalent numerical behavior in the traditional instructions, therefore, its numerical behavior is slightly different.

The `cudnnSetConvolutionMathType`, `cudnnSetRNNMatrixMathType`, and `cudnnMathType_t` functions enable you to choose whether or not to use Tensor Core operations in the convolution and RNN layers respectively by setting the math mode to either `CUDNN_TENSOR_OP_MATH` or `CUDNN_DEFAULT_MATH`.

Tensor Core operations perform parallel floating point accumulation of multiple floating point products.

Setting the math mode to `CUDNN_TENSOR_OP_MATH` indicates that the library will use Tensor Core operations.

The default is `CUDNN_DEFAULT_MATH`. This default indicates that the Tensor Core operations will be avoided by the library. The default mode is a serialized operation.
whereas, the Tensor Core is a parallelized operation, therefore, the two might result in slightly different numerical results due to the different sequencing of operations.

The library falls back to the default math mode when Tensor Core operations are not supported or not permitted.

cudnnSetConvolutionGroupCount
A new interface that allows applications to perform convolution groups in the convolution layers in a single API call.

cudnnCTCLoss
The cudnnCTCLoss provides a GPU implementation of the Connectionist Temporal Classification (CTC) loss function for RNNs. The CTC loss function is used for phoneme recognition in speech and handwriting recognition.

CUDNN_BATCHNORM_SPATIAL_PERSISTENT
The CUDNN BATCHNORM SPATIAL.Persistent function is a new batch normalization mode for cudnnBatchNormalizationForwardTraining and cudnnBatchNormalizationBackward. This mode is similar to CUDNN BATCHNORM SPATIAL, however, it can be faster for some tasks.

cudnnQueryRuntimeError
The cudnnQueryRuntimeError function reports error codes written by GPU kernels when executing cudnnBatchNormalizationForwardTraining and cudnnBatchNormalizationBackward with the CUDNN BATCHNORM SPATIAL.Persistent mode.

cudnnGetConvolutionForwardAlgorithm_v7
This new API returns all algorithms sorted by expected performance (using internal heuristics). These algorithms are output similarly to cudnnFindConvolutionForwardAlgorithm.

cudnnGetConvolutionBackwardDataAlgorithm_v7
This new API returns all algorithms sorted by expected performance (using internal heuristics). These algorithms are output similarly to cudnnFindConvolutionBackwardDataAlgorithm.

cudnnGetConvolutionBackwardFilterAlgorithm_v7
This new API returns all algorithms sorted by expected performance (using internal heuristics). These algorithms are output similarly to cudnnFindConvolutionBackwardFilterAlgorithm.

CUDNN_REDUCE_TENSOR_MUL_NO_ZEROS
The MUL NO ZEROS function is a multiplication reduction that ignores zeros in the data.
CUDNN_OP_TENSOR_NOT
The OP_TENSOR_NOT function is a unary operation that takes the negative of (alpha*A).

cudnnGetDropoutDescriptor
The cudnnGetDropoutDescriptor function allows applications to get dropout values.

Using cuDNN v7.0.1
Ensure you are familiar with the following notes when using this release.

‣ Multi-threading behavior has been modified. Multi-threading is allowed only when using different cuDNN handles in different threads.
‣ In cudnnConvolutionBackwardFilter, dilated convolution did not support cases where the product of all filter dimensions was odd for half precision floating point. These are now supported by CUDNN_CONVOLUTION_BWD_FILTER_ALGO1.
‣ Fixed bug that produced a silent computation error for when a batch size was larger than 65536 for CUDNN_CONVOLUTION_FWD_ALGO_IMPLICIT_PRECOMP_GEMM.
‣ In getConvolutionForwardAlgorithm, an error was not correctly reported in v5 when the output size was larger than expected. In v6 the CUDNN_STATUS_NOT_SUPPORTED error message displayed. In v7, this error is modified to CUDNN_STATUS_BAD_PARAM.
‣ In cudnnConvolutionBackwardFilter, cuDNN now runs some exceptional cases correctly where it previously erroneously returned CUDNN_STATUS_NOT_SUPPORTED. This impacted the algorithms CUDNN_CONVOLUTION_BWD_FILTER_ALGO0 and CUDNN_CONVOLUTION_BWD_FILTER_ALGO3.

Deprecated Features
The following routines have been removed:

‣ cudnnSetConvolution2dDescriptor_v4
‣ cudnnSetConvolution2dDescriptor_v5
‣ cudnnGetConvolution2dDescriptor_v4
‣ cudnnGetConvolution2dDescriptor_v5

Only the non-suffixed versions of these routines remain.

The following routines have been created and have the same API prototype as their non-suffixed equivalent from cuDNN v6:
- **cudnnSetRNNDescriptor_v5** - The non-suffixed version of the routines in cuDNN v7.0.1 are now mapped to their _v6 equivalent.


| Attention | It is strongly advised to use the non-suffixed version as the _v5 and _v6 routines will be removed in the next cuDNN release. |

- **cudnnGetConvolutionForwardAlgorithm**, **cudnnGetConvolutionBackwardDataAlgorithm**, and **cudnnGetConvolutionBackwardFilterAlgorithm** - A _v7 version of this routine has been created. For more information, see the Backward compatibility and deprecation policy chapter of the cuDNN documentation for details.

**Known Issues**

- cuDNN pooling backwards fails for pooling window size > 256.
Notice

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