

Introduction to Scientific Programming using GPGPU and CUDA



Day 2

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- Tools from CUDA-Toolkit

- Profiler
- CUDA-GDB
- CUDA-memcheck
- Parallel Nsight

- CUDA-Enabled Libraries

- CUBLAS
- CUFFT
- CUSPARSE
- CURAND
- MAGMA, THRUST, CUDDP, ...



Profiling tools: built-in

- CUDA toolkit provides useful tools for profiling your code

```
export CUDA_PROFILE=1  
export CUDA_PROFILE_CONFIG=$HOME/.config
```

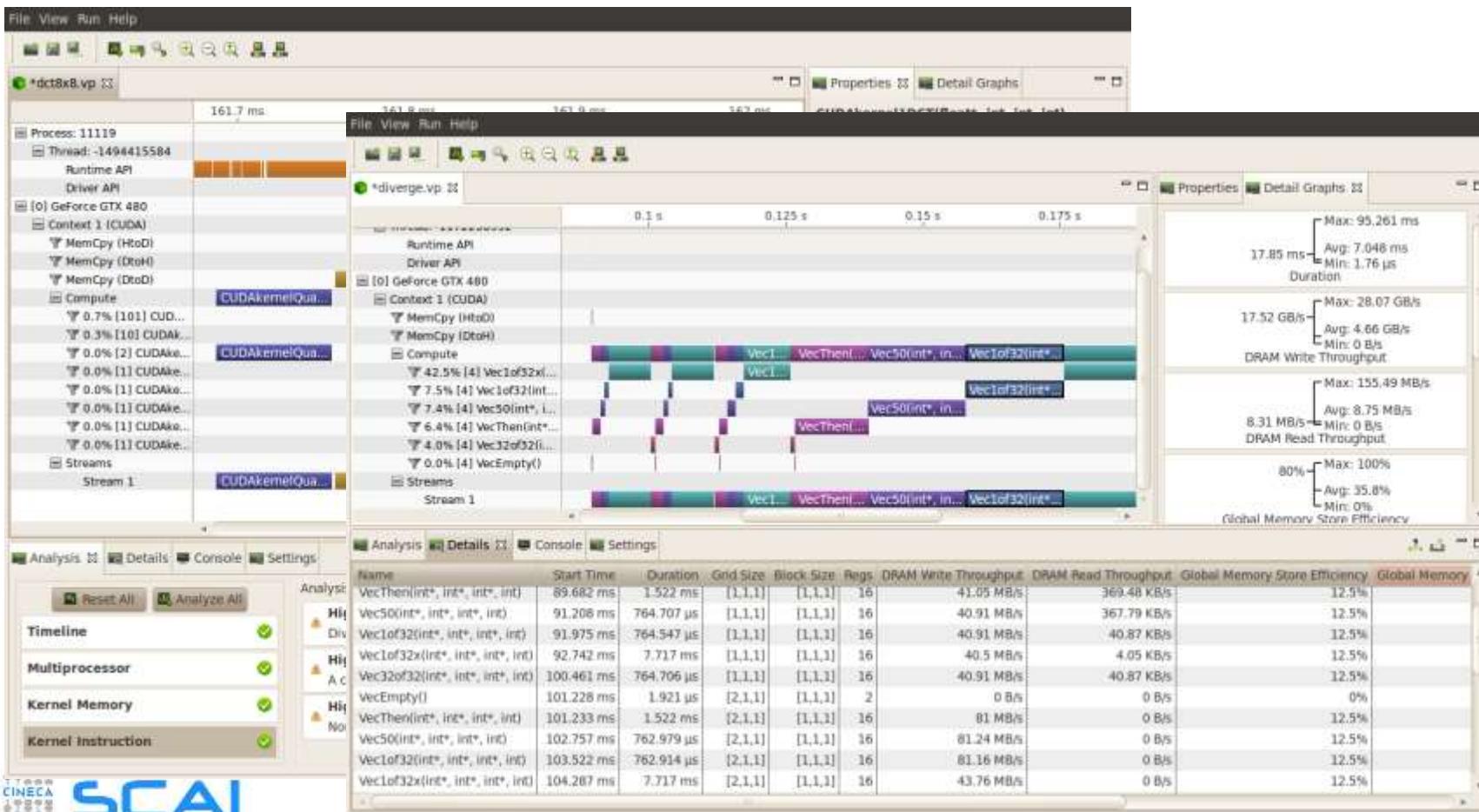
```
// Contents of config  
gld_coherent  
gld_incoherent  
gst_coherent  
gst_incoherent
```

```
gld_incoherent: Number of non-coalesced global memory loads  
gld_coherent: Number of coalesced global memory loads  
gst_incoherent: Number of non-coalesced global memory stores  
gst_coherent: Number of coalesced global memory stores  
local_load: Number of local memory loads  
local_store: Number of local memory stores  
branch: Number of branch events taken by threads  
divergent_branch: Number of divergent branches within a warp  
instructions: instruction count  
warp_serialize: Number of threads in a warp that serialize  
based on address conflicts to shared or constant memory  
cta_launched: executed thread blocks
```

```
method,gputime,cputime,occupancy,gld_incoherent,gld_coherent,gst_incoherent,gst_coherent  
method=[ memcpy ] gputime=[ 438.432 ]  
method=[ _Z17reverseArrayBlockPiS_ ] gputime=[ 267.520 ] cputime=[ 297.000 ] occupancy=[ 1.000 ]  
gld_incoherent=[ 0 ] gld_coherent=[ 1952 ] gst_incoherent=[ 62464 ] gst_coherent=[ 0 ]  
method=[ memcpy ] gputime=[ 349.344 ]  
...  
...  
...
```

Profiling: Visual Profiler

- Traces hosts and device calls, da transfers, kernel launches, shows overlapping streams and measure performances
- supports automated analysis (hardware counters)



Debugging: CUDA-GDB

- CUDA Toolkit also provides a cuda-gdb text debugger
 - the traditional gdb enhanced with CUDA extention

```
(cuda-gdb) info cuda threads
BlockIdx ThreadIdx To BlockIdx ThreadIdx Count Virtual PC Filename Line
Kernel 0* (0,0,0) (0,0,0) (0,0,0) (255,0,0) 256 0x0000000000866400
bitreverse.cu 9
(cuda-gdb) thread
[Current thread is 1 (process 16738) ]
(cuda-gdb) thread 1
[Switching to thread 1 (process 16738) ]
#0 0x000019d5 in main () at bitreverse.cu:34
34 bitreverse<<<1, N, N*sizeof(int)>>>(d);
(cuda-gdb) backtrace
#0 0x000019d5 in main () at bitreverse.cu:34
(cuda-gdb) info cuda kernels
Kernel Dev Grid SMS Mask GridDim BlockDim Name Args
0 0 1 0x00000001 (1,1,1) (256,1,1) bitreverse data=0x110000
```

Debugging: CUDA-MEMCHECK

- Very useful to detect buffer overflows, misaligned global memory accesses and memory leaks
- stand alone or fully integrated in CUDA-GDB

```
$ cuda-memcheck --continue ./memcheck_demo
=====
      CUDA-MEMCHECK
Allocating memory
Running unaligned_kernel
Ran unaligned_kernel: no error
Sync: no error
Running out_of_bounds_kernel
Ran out_of_bounds_kernel: no error
Sync: no error
=====
      Invalid __global__ write of size 4
      at 0x00000038 in memcheck_demo.cu:5:unaligned_kernel
      by thread (0,0,0) in block (0,0,0)
      Address 0x200200001 is misaligned
=====
      Invalid __global__ write of size 4
      at 0x00000030 in memcheck_demo.cu:10:out_of_bounds_kernel
      by thread (0,0,0) in block (0,0,0)
      Address 0x87654320 is out of bounds
=====
=====
=====
      ERROR SUMMARY: 2 errors
```

Parallel NSight

- Plug-in available for Eclipse and VisualStudio IDE
- Aggregates all external functionalities:
 - Debugger (fully integrated)
 - Visual Profiler
 - Memory correctness checker
- As a plug-in, it extends all the convenience of IDE development to CUDA
- Very fast growing community and feature rich:
 - supporto for multi-GPU
 - remote debug and profiling
 - PTX assembly view
 - warp inspector
 - expression lamination

Parallel NSight

Process: [1840] voxelpipe_demo.exe Thread: [2874912] <No Name> Stack Frame: CUmodule 05508fe0 - [2] trace - Line 148

Connections: localhost

CUDA Info 1

Warp Index PC Active Mask Status Exception File Name Source Lin Lanes

Current	blockIdx	Warp Index	PC	Active Mask	Status	Exception	File Name	Source Lin	Lanes
	(0, 0, 0)	0	0x003e1ad8	0xffffffff80	Breakpoint	None	rt_render.cu	163	
	(0, 0, 0)	1	0x003e1ad8	0xffffffff00	Breakpoint	None	rt_render.cu	163	
	(0, 0, 0)	2	0x003e1ad8	0xfffffc00	Breakpoint	None	rt_render.cu	163	
	(0, 0, 0)	3	0x003e1ad8	0xfffff800	None	None	rt_render.cu	163	
	(1, 0, 0)	0	0x003e1298	0x03e00000	Breakpoint	None	rt_render.cu	148	
	(1, 0, 0)	1	0x003e1298	0x07c00000	Breakpoint	None	rt_render.cu	148	
	(1, 0, 0)	2	0x003ede70	0xffffffff	None	None	ci_include.h	423	

CUDA WarpWatch 1

Name	Type	ray_inv.x	ray_inv.y	ray_inv.z
0	_local_float	-1.4444908	-1.7955524	-2.17
1	_local_float	-1.44425	-1.7967783	-2.17
2	_local_float	-1.4440092	-1.7980076	-2.17
3	_local_float	-1.4437686	-1.7992405	-2.17
4	_local_float	-1.4435281	-1.800477	-2.17
5	_local_float	-1.4432876	-1.8017174	-2.17
6	_local_float	-1.4430474	-1.8029615	-2.17
7	_local_float	-1.4428074	-1.8042094	-2.16
8	_local_float	-1.4425675	-1.8054608	-2.16
9	_local_float	-1.4423276	-1.8067161	-2.16
10	_local_float	-1.4420878	-1.8079749	-2.16
11	_local_float	-1.4418485	-1.8092378	-2.16
12	_local_float	-1.4416089	-1.8105046	-2.16
13	_local_float	-1.4413697	-1.8117749	-2.16
14	_local_float	-1.4411306	-1.8130492	-2.16
15	_local_float	-1.4408917	-1.8143274	-2.15
16	_local_float	-1.4406527	-1.8156093	-2.15
17	_local_float	-1.4404141	-1.8168953	-2.15
18	_local_float	-1.4401754	-1.818185	-2.15
19	_local_float	-1.439937	-1.8194786	-2.15
20	_local_float	-1.4396986	-1.820776	-2.15
21	_local_float	-1.4394605	-1.8220775	-2.15
22	_local_float	-1.4392225	-1.8233831	-2.14
23	_local_float	-1.4389844	-1.8246926	-2.14
24	_local_float	-1.4387469	-1.8260059	-2.14
25	_local_float	-1.4385092	-1.8273233	-2.14
26	_local_float	-1.4382718	-1.828645	-2.14
27	_local_float	-1.4380344	-1.8299706	-2.14
28	_local_float	-1.4377974	-1.8313001	-2.14
29	_local_float	-1.4375603	-1.832634	-2.13
30	_local_float	-1.4373236	-1.8339716	-2.13
31	_local_float	-1.4370868	-1.8353136	-2.13

CUDA Info 1

Modules

rt_render.cu

(Unknown Scope)

```

143     node_index = node.get_index(); // jump to child
144 }
145 else
146 {
147     // leaf intersection
148     const uint32 leaf_index = node.get_index();
149     const Bvh_leaf leaf    = geometry.m_bvh_leaves[ leaf_index ];
150     const uint32 leaf_end  = leaf.get_index() + leaf.get_size();
151     const uint32 leaf_begin = leaf.get_index();
152
153     for (uint32 tri_index = leaf_begin; tri_index < leaf_end; ++tri_index)
        !!!

```

Locals

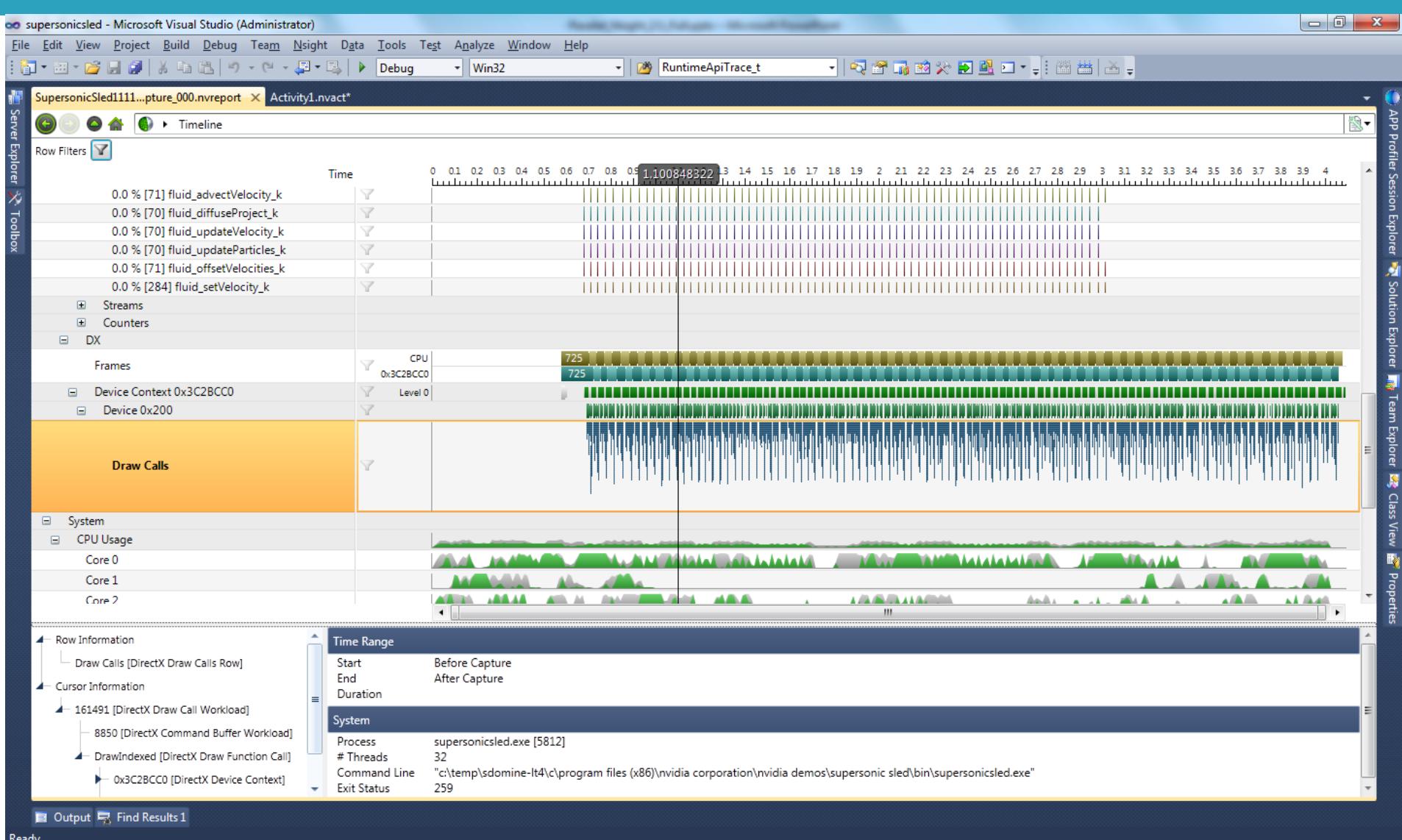
Name	Type	Value
leaf	_local_	{m_size = 67106176, m_index = 0}
leaf_index	_local_	'leaf_index' has no value at the target location.
leaf_end	_local_	'leaf_end' has no value at the target location.
leaf_begin	_local_	'leaf_begin' has no value at the target location.
node	_local_	{m_packed_data = 2147484877, m_skip_node = 248}
_T21669	_local_	{x = -1.4394605, y = -1.8220775, z = -2.150774}
ray_inv	_local_	{x = -1.4394605, y = -1.8220775, z = -2.150774}
node_index	_local_	'node_index' has no value at the target location.

Call Stack

Name	Language
CUmodule 05508fe0 - [2] trace - Line 148	CUDA
CUmodule 05508fe0 - [1] render_pixel - Line 409	CUDA
CUmodule 05508fe0 - [0] rt_trace_primary_kernel - Line 493	CUDA

Ready

Parallel NSight



CUDA Enabled Libraries

- CUDA Toolkit includes many usefull libraries:
 - **CUBLAS**: Basic Linear Algerba Subprograms
 - **NVBLAS**: multi-GPUs accelerated drop-in BLAS built on top of cuBLAS
 - **CUFFT** : Fast Fourier Transform
 - **CUSPARSE**: space matrix linear algebra
 - **CURAND**: pseudorandom and quasirandom number generator
 - **NPP**: image,signal processing, statistic (nVIDIA Performance Primitives)
 - **THRUST**: vector-based library for parallel algorythms in C++
 - other CUDA enabled libraries outside the CUDA Toolkit :
 - **MAGMA** (Matrix Algebra on GPU and Multicore Architectures)
<http://icl.cs.utk.edu/magma/>
 - **CUDPP** (Data Parallel Primitives): parallel prefix-sum, sort, reduction
<http://code.google.com/p/cudpp/>
 - **CULA**: CUDA LAPACK API (single precision version is freely available)
<http://www.culatools.com/contact/cuda-training/>
 - **CUSP**: CUDA Sparse Solver and graph:
http://developer.nvidia.com/object/npp_home.html
- for a complete list, visit CUDA-Zone and look for GPU-Accelerated Libraries*

CUDA Math Library

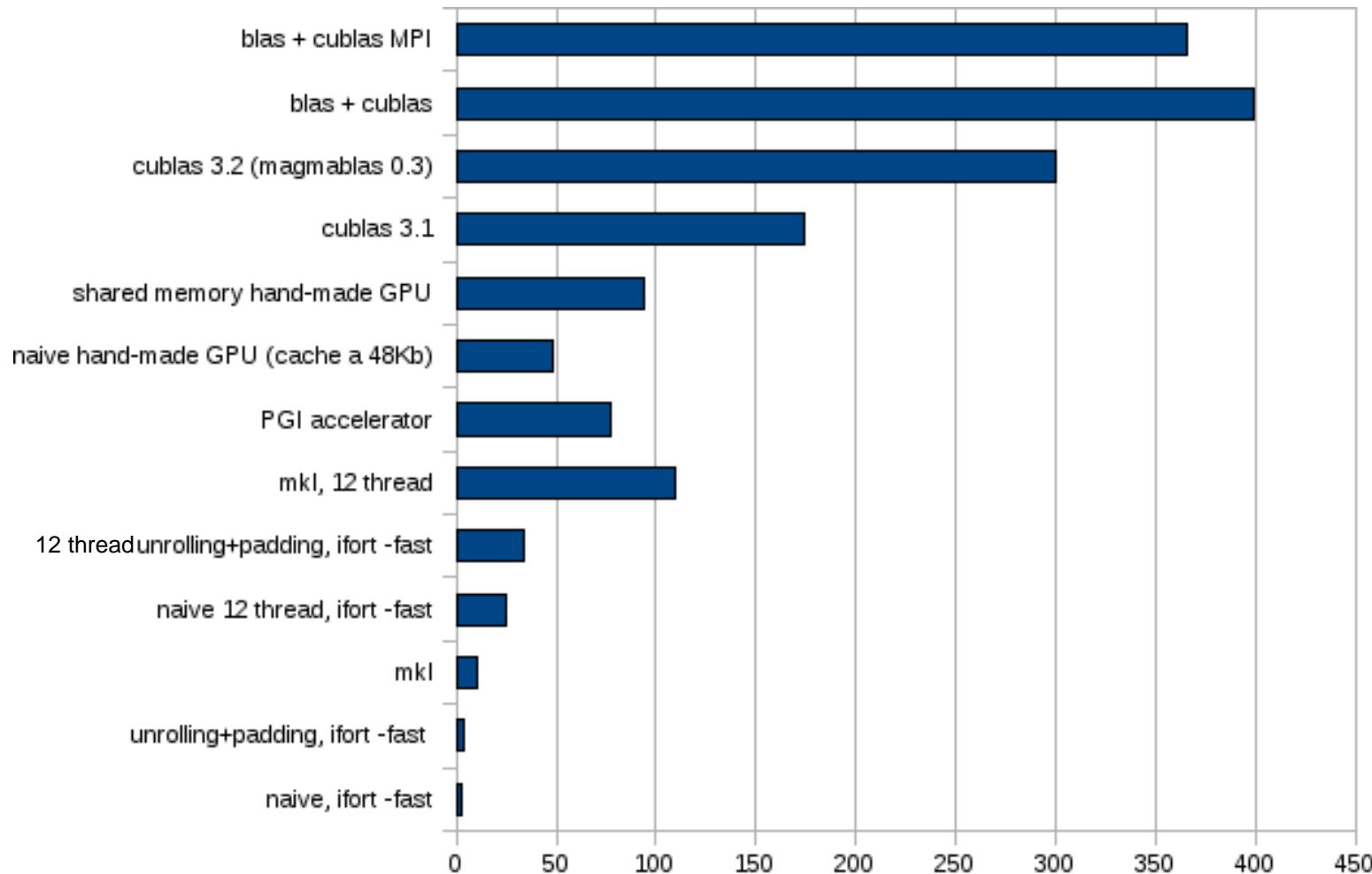
- Of course, CUDA Toolkit provides mathematical functions as other higher level language
 - by simply adding “#include `math.h`” in your source code
 - Complete support for all C99 standard float and double math functions
- IEEE-754 accurate for float, double, and all rounding modes
 - Extended Trigonometry and Exponential Functions
 - `cospi`, `sincos`, `sinpi`, `exp10`
 - Inverse Error Functions (`erfinv`, `erfcinv`)
 - Optimized Reciprocal Functions (`rsqrt`, `rcbrt`)
 - Bessel Functions (`j0,j1,jn,y0,y1,yn`)
- Floating Point Data Attributes (`signbit`, `isfinite`, `isinf`, `isnan`)
- intrinsic versions are also provided

cuBLAS Library

- BLAS is a standard in terms of interface and accuracy for most other libraries which implements linear algebra operations
 - BLAS Level 1:
 - BLAS Level 2:
 - BLAS Level 3:
- There are vendor optimized implementation for many hardware architecture (Intel, Power, ARM, etc)
- CUDA Toolkit provide a CUDA-enabled implementation of all BLAS
- WARNING: data layout in memory follows column-major ordering and 1-based indexing
- Function naming convention: cublas + BLAS name

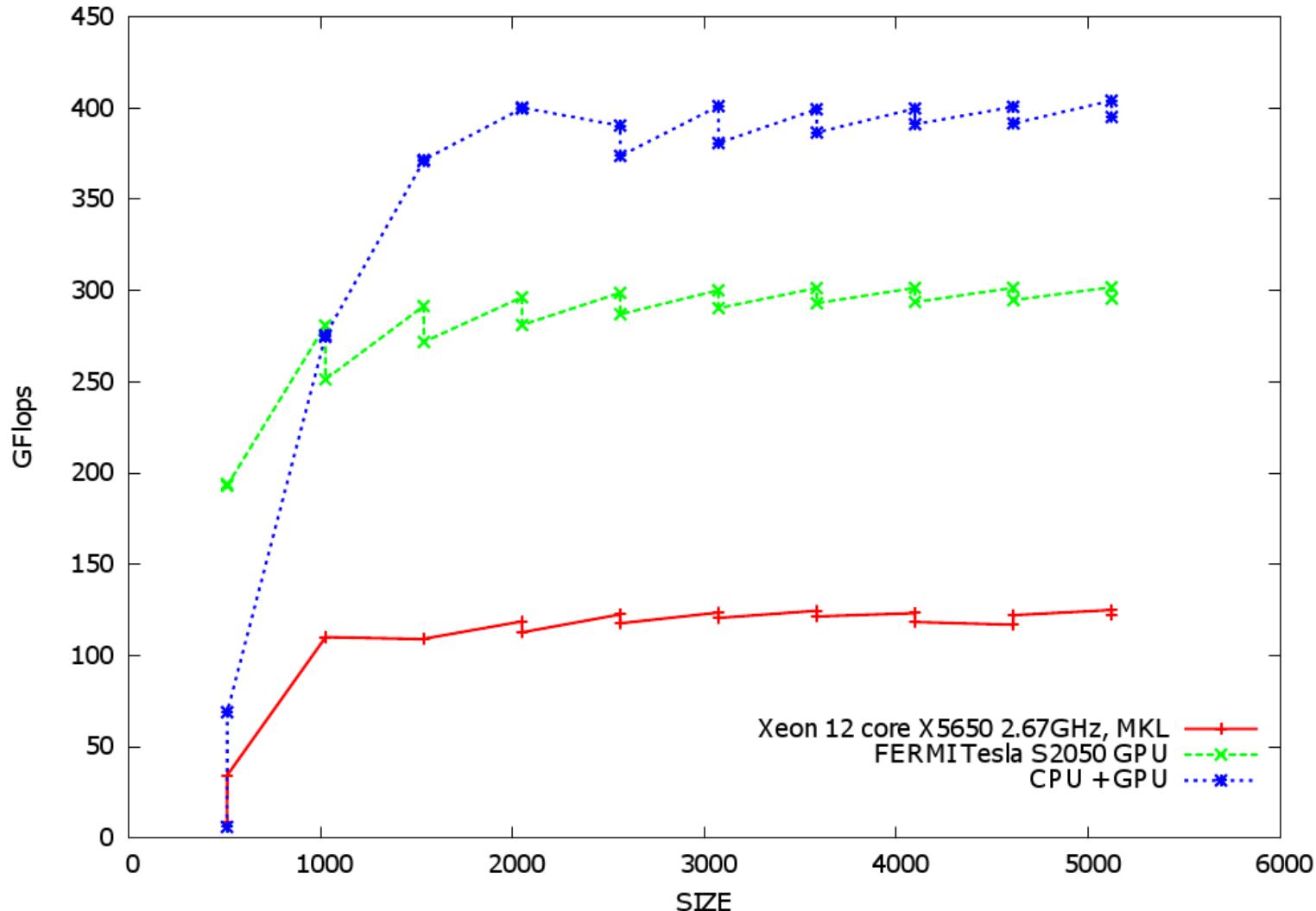
cuBLAS: DGEMM performance

- Here we show a simple performance study case for DGEMM (double-precision dense matrix matrix product - old system)



CUBLAS: DGEMM performance

- Performance versus matrix size dependency



cuBLASXT

- Starting with CUDA 6.0, the cuBLAS Library exposes two API
 - the regular cuBLAS API
 - the new cuBLASXT API
- With cuBLAS API
 - the application must allocate the required matrices and vectors in the GPU memory space
 - fill them with data, call the sequence of desired cuBLAS functions,
 - then upload the results from the GPU memory space back to the host
- With cuBLASXT API
 - the application must allocate data using managed memory
 - the library will take care of dispatching the operations to one or multiple GPUs present in the system

cuFFT

- cuFFT is the CUDA version of the Fast Fourier Transform
 - based on Cooley-Turkey and Bluestein algorithm
- cuFFT API is very similar to the FFTW one
 - as FFTW does, cuFFT use the *workplan* concept to optimize its work
 - once a *workplan* is computed, the library itself maintains necessary information to execute FFT operation on data many times efficiently
 - WARNING: cuFFT follow row-major convention for data in memory
- Other key features:
 - provides 1D, 2D, 3D transform
 - for many real and complex types (single, double, quad precision)
 - in-place and out-of-place transforms
 - non-normalized output:
 - $\text{IFFT}(\text{FFT}(A)) = \text{len}(A) * A$
 - support for asynchronous operation on CUDA streams
 - thread-safe (**CUDA 4.1**)

cuFFT sample: 2D complex-complex

```
#define NX 256
#define NY 128

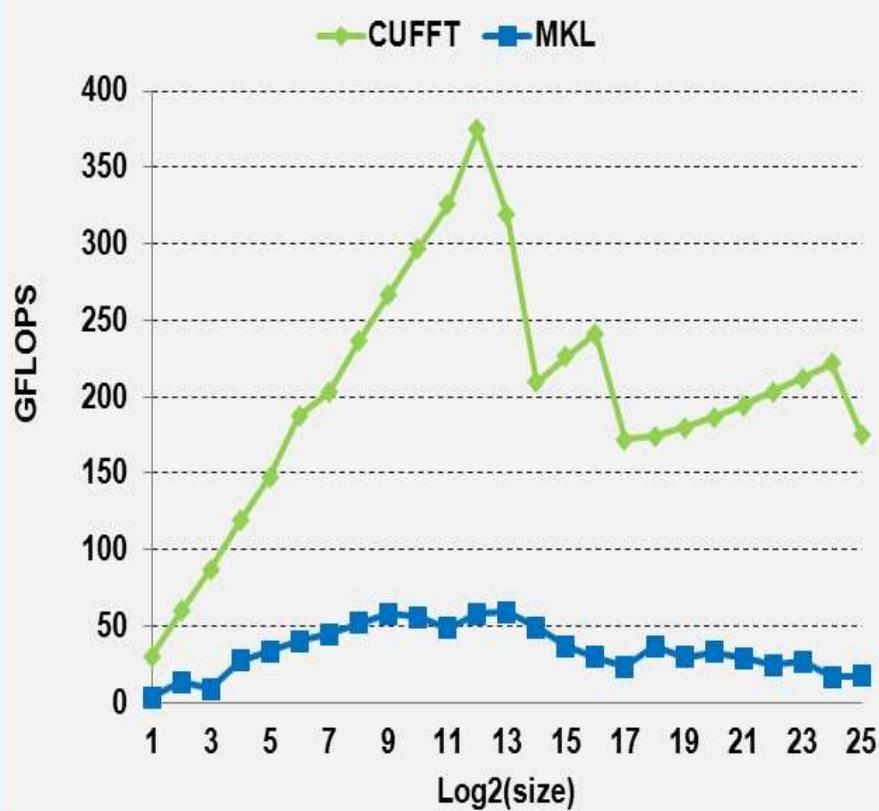
cufftHandle plan;
cufftComplex *idata, *odata;
cudaMalloc( (void**)&idata, sizeof(cufftComplex)*NX*NY);
cudaMalloc( (void**)&odata, sizeof(cufftComplex)*NX*NY);

...
/* create a plan for FFT 2D */
cufftPlan2d(&plan, NX,NY, CUFFT_C2C);
/* use plan for "out of place" transform */
cufftExecC2C(plan, idata, odata, CUFFT_FORWARD);
/* back transform "in place" */
cufftExecC2C(plan, odata, odata, CUFFT_INVERSE);
/* if input output pointers differ, "out of place" is implied */

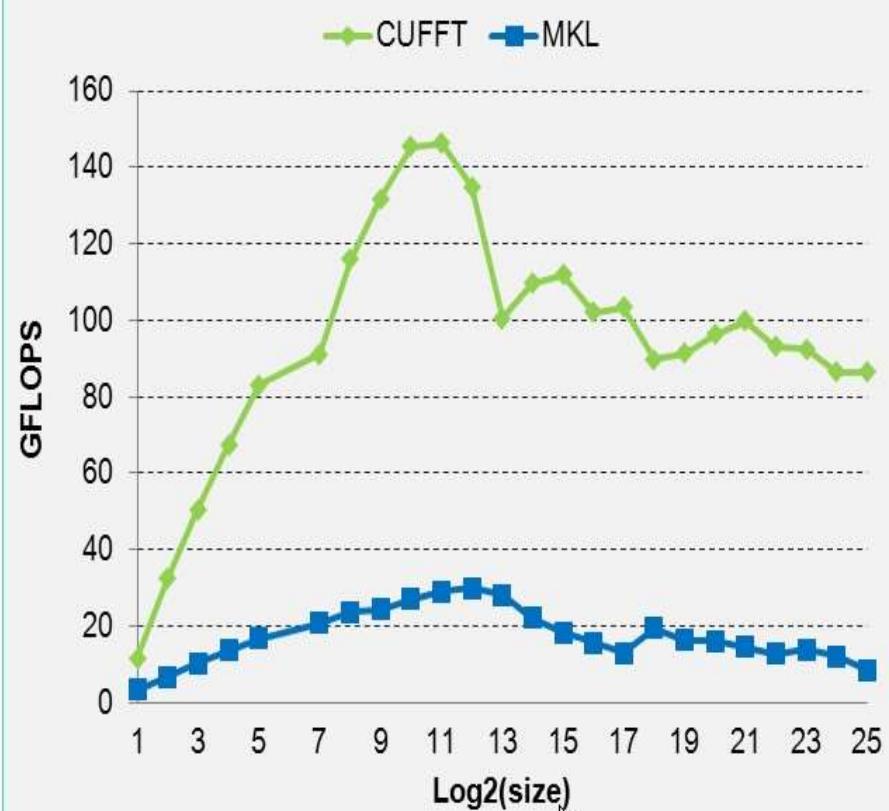
/* destroy plan and free resources */
cufftDestroy(plan);
cudaFree(idata), cudaFree(odata);
```

cuFFT: performances of FFT1D

cuFFT Single Precision



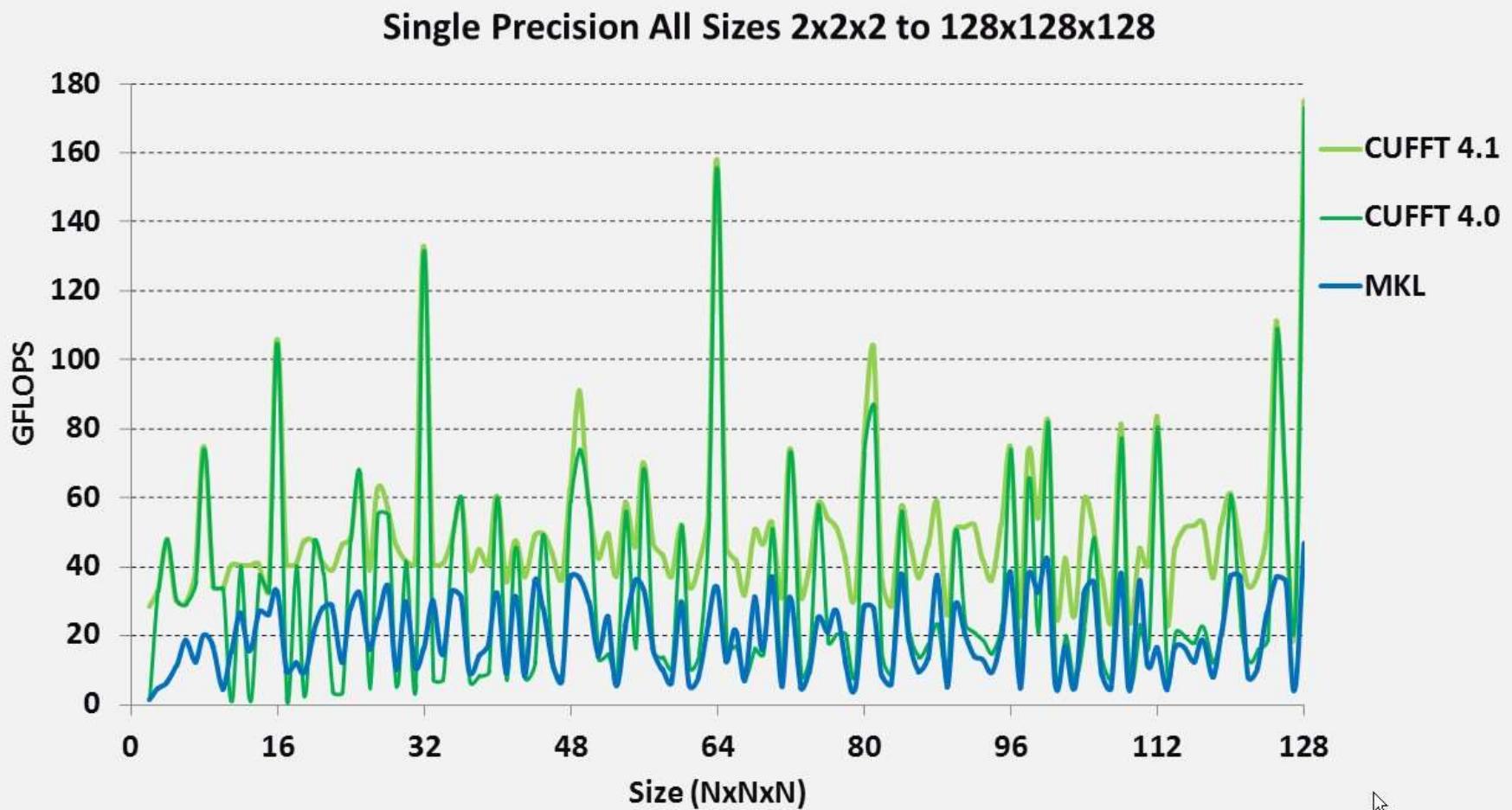
cuFFT Double Precision



- Measured on sizes that are exactly powers-of-2
- cuFFT 4.1 on Tesla M2090, ECC on
- MKL 10.2.3, TYAN FT72-B7015 Xeon x5680 Six-Core @ 3.33 GHz

- MKL 10.2.3, TYAN FT72-B7015 Xeon x5680 Six-Core @ 3.33 GHz
- Performance may vary based on OS version and motherboard configuration

cuFFT: performances of FFT3D



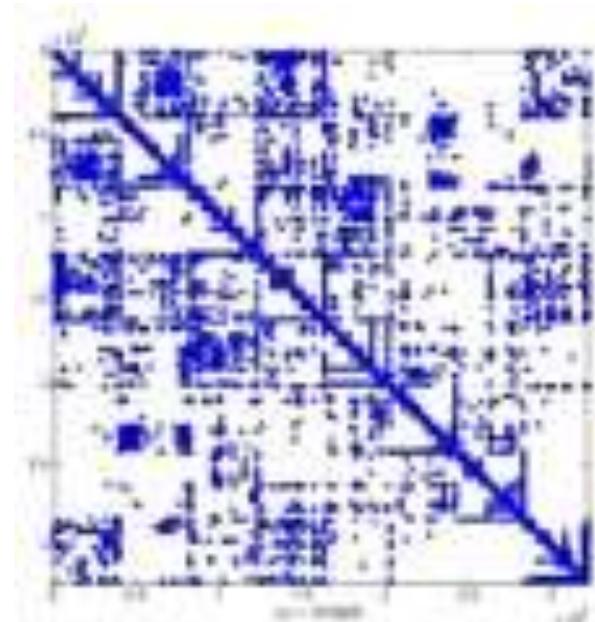
- cuFFT 4.1 on Tesla M2090, ECC on
- MKL 10.2.3, TYAN FT72-B7015 Xeon x5680 Six-Core @ 3.33 GHz

• Performance may vary based on OS ver. and motherboard config.



cuSPARSE

- support for dense, COO, CSR, CSC, ELL/HYB and Blocked CSR sparse matrix formats
- Level 1 routines for sparse vector x dense vector operations
- Level 2 routines for sparse matrix x dense vector operations
- Level 3 routines for sparse matrix x multiple dense vectors (tall matrix)
- Routines for sparse matrix by sparse matrix addition and multiplication
- Conversion routines that allow conversion between different matrix formats
- Sparse Triangular Solve
- Tri-diagonal solver
- Incomplete factorization preconditioners ilu0 and ic0



Flexible usage model

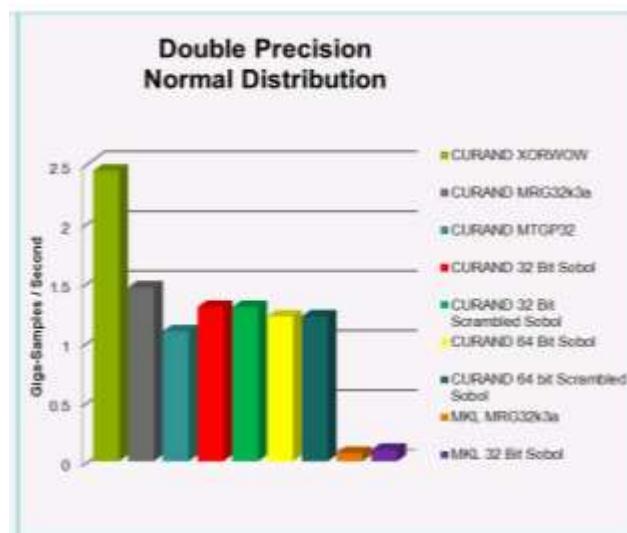
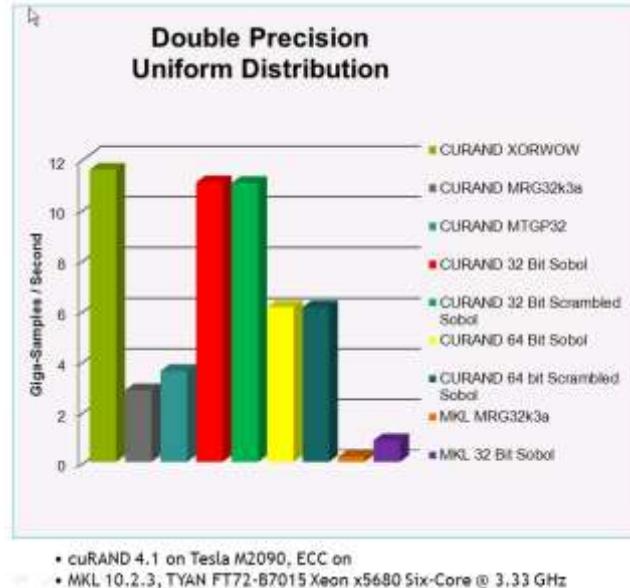
- Host API for generating random numbers in bulk on the GPU
- Inline implementation allows use inside GPU functions/kernels, or in your host code

Four high-quality RNG algorithms

- MRG32k3a
- MTGP Mersenne Twister
- XORWOW pseudo-random generation
- Sobol' quasi-random number generators, including support for scrambled and 64-bit RNG

Multiple RNG distribution options

- Uniform distribution
- Normal distribution
- Log-normal distribution
- Single-precision or double-precision



1. Create a generator:

curandCreateGenerator()

2. Set a seed:

curandSetPseudoRandomGeneratorSeed()

3. Generate the data from a distribution:

curandGenerateUniform()/curandGenerateUniformDouble() : Uniform

curandGenerateNormal()/cuRandGenerateNormalDouble() : Gaussian

curandGenerateLogNormal/curandGenerateLogNormalDouble() : Log-Normal

4. Destroy the generator:

curandDestroyGenerator()

cuRAND

```
#include <stdio.h>
#include <stdlib.h>
#include <cuda.h>
#include <curand.h>

int main() {

    int i, n = 100;
    curandGenerator_t gen;
    float *devData, *hostData;

    // Allocate n floats on host
    hostData = (float *) malloc (n,
        sizeof(float));

    // Allocate nfloats on device
    cudaMalloc((void **) &devData, n *
        sizeof(float));

    // Create psudo-random number generator
    curandCreateGenerator(&gen,
        CURAND_RNG_PSEUDO_DEFAULT);

    // set seed
    curandSetPseudoRandomGeneratorSeed(gen,
        1234ULL);
```

```
// generate n float on device
curandGenerateUniform(gen, devData, n);

// copy device memory to host
cudaMemcpy(hostData, devData, n *
    sizeof(float),
    cudaMemcpyDeviceToHost);

// show result
for (i = 0; i < n; i++) {
    printf("%1.4f ", hostData[i]);
}
printf("\n");

// Cleanup
curandDestroyGenerator(gen);

cudaFree(devData)
free(hostData)

return 0;
```

cuRAND

```
#include <stdio.h>
#include <stdlib.h>
#include <cuda.h>
#include <curand_kernel.h>

__global__ void
setup_kernel(curandState *state)
{
    int id = threadIdx.x - blockIdx.x *
64;
    // each thread gets same seed
    curand_init(1234, id, 0,
&state[id]);
}

__global__ void generate_kernel(
    curandState *state, int *result)
{
    int id = threadIdx.x + blockIdx.x *
64;
    int count = 0;
    unsigned int x;

    curandState localState = state[id];
```

```
// generate pseudo-random unsigned
for (int n = 0; n < 1000000; n++) {
    x = curand(&localState);
}

// copy state back to global memory
state[id] = localState;

// store results
result[id] += count;
}
```

CUDPP

- CUDPP: CUDA Data Parallel Primitives library
- collection of many *data-parallel* algorithms:
 - prefix-sum (“scan”)
 - parallel sort
 - reduction
- Important building blocks for a wide variety of data-parallel algorithms, including sorting, stream compaction, and building data structures such as trees and summed-area tables
- provides primitives and other complex operation functions such as:
 - hash table
 - array compaction
 - tridiagonal linear system solver
 - sparse matrix-vector product
- Specifications
 - open source project in C/C++
 - Support for Windows, Linux and OSX
 - open source: <http://cudpp.github.io/>

MAGMA: Matrix Algebra on GPU and Multicore Architectures

- LAPACK (Linear Algebra PACKage) is the de facto standard linear algebra operations
 - **built on BLAS**
- MAGMA is essentially a re-implementation of standard legacy) LAPACK on heterogeneous architectures such as GPU + CPU multicore
 - **built on top of cuBLAS**
- MAGMA 1.x support multi-GPU CUDA enabled environment, and its able to overlap computation on CPU cores (essentially through optimized multithreaded version of BLAS and LAPACK for the CPU side)
- Developed by the ICL group (Innovative Computing Laboratory) + many external collaborations + user community
- open source: <http://icl.cs.utk.edu/projectsfiles/magma/>
- WARNING: memory data layout follow the FORTRAN (column-major) convention

MAGMA: C/C++ usage

- MAGMA is entirely developed in C. So its usage is very easy in a C/c++ code
- The library interface is just in one file:
 - magma.h
- user must explicitly manage memory on host and device using traditional CUDA runtime APIs

```
// Reduction of a symmetric matrix into tridiagonal form
#include <cuda.h> //
#include <magma.h> //

// magma_int_t magma_dsytrd( char uplo, magma_int_t n, double *A,
//                           magma_int_t lda, double *d, double *e,
//                           double *tau, double *work, magma_int_t *lwork,
//                           double *da, double *dc, magma_int_t *info);

cudaError_t stat;
double *da, *dwork;
stat = cudaMalloc((void**)&da, n*n*sizeof(double));
stat = cudaMalloc((void**)&dwork, workSize* sizeof(double));
magma_dsytrd('U', n, A, lda, diagonal, offdiagonal, tau, work, lwork, da, dwork,
&info)
```

MAGMA: F90/2003 usage

- In order to use MAGMA with F90/2003 requires the programmer to provide interface and the ISO_C_BINDING module

```
!! Native C interface:  
!! magma_int_t magma_dsytrd( char uplo, magma_int_t n, double *A,  
!!                               magma_int_t lda, double *d, double *e,  
!!                               double *tau, double *work, magma_int_t *lwork,  
!!                               double *da, double *dc, magma_int_t *info);  
!!  
!! Interface for F90/2003:  
subroutine magma_dsytrd(uplo, n, a, lda, d, e, tau, work, lwork, da, dc, info)  
  
bind(C, name="magma_dsytrd")  
use iso_c_binding  
implicit none  
  
character, value:: uplo  
integer(C_INT), value :: n, lda  
integer(C_INT) :: info, lwork  
type(C_PTR), value :: a, d, e, tau, work, da, dc  
  
! NB: type(C_PTR), value == void*  
end subroutine magma_dsytrd
```

CUDA Thrust

A C++ template library for CUDA

- Mimics the C++ STL
- Two containers
 - Manage memory on host and device:

`thrust::host_vector<T>`

`thrust::device_vector<T>`

▪ Algorithms

- Sorting, reduction, scan, etc:

`thrust::sort()`

`thrust::reduce()`

`thrust::inclusive_scan()`

- act on ranges of the container data by pair of iterators (a sort of pointers)

The screenshot shows the GitHub project page for Thrust. At the top is the Thrust logo: a green and yellow swoosh with the word "Thrust" in white. Below the logo is a brief description: "Thrust is a CUDA library of parallel algorithms with an interface resembling the C++ Standard Template Library (STL). Thrust provides a modern, high-level interface for GPU programming that greatly enhances developer productivity. Develop high-performance applications rapidly with Thrust!" To the right of the logo is a "What is Thrust?" section. Below that is a "News" section listing several bullet points about recent releases. Under "Examples", there is a code snippet demonstrating how to generate random numbers on the host and transfer them to the device using Thrust's host and device vectors.

```
#include <thrust/host_vector.h>
#include <thrust/device_vector.h>
#include <thrust/generators.h>
#include <thrust/join.h>
#include <thrust/transform.h>
#include <thrust/reduce.h>

int main(int argc)
{
    // generate 2M random numbers on the host
    thrust::host_vector<float> h_vec(2 * 1024 * 1024);
    thrust::generate(h_vec.begin(), h_vec.end(), rand);

    // transfer data to the device
    thrust::device_vector<float> d_vec = h_vec;

    // sort data on the device (100M keys per second on a GTX 480)
    thrust::sort(d_vec.begin(), d_vec.end());

    // transfer data back to host
    thrust::copy(d_vec.begin(), d_vec.end(), h_vec.begin());
}
```

CUDA Thrust

```
#include <thrust/host_vector.h>
#include <thrust/device_vector.h>
#include <thrust/generate.h>
#include <thrust/sort.h>
#include <thrust/copy.h>
#include <cstdlib>

int main(void)
{
    // generate 32M random numbers on the host
    thrust::host_vector<int> h_vec(32 << 20);
    thrust::generate(h_vec.begin(), h_vec.end(), rand);

    // transfer data to the device
    thrust::device_vector<int> d_vec = h_vec;

    // sort data on the device (846M keys per second on GeForce GTX 480)
    thrust::sort(d_vec.begin(), d_vec.end());

    // transfer data back to host
    thrust::copy(d_vec.begin(), d_vec.end(), h_vec.begin());

    return 0;
}
```

CUDA Thrust

```
#include <thrust/host_vector.h>
#include <thrust/device_vector.h>
#include <thrust/generate.h>
#include <thrust/reduce.h>
#include <thrust/functional.h>
#include <cstdlib>

int main(void)
{
    // generate random data on the host
    thrust::host_vector<int> h_vec(100);
    thrust::generate(h_vec.begin(), h_vec.end(), rand);

    // transfer to device and compute sum
    thrust::device_vector<int> d_vec = h_vec;
    int x = thrust::reduce(d_vec.begin(), d_vec.end(), 0, thrust::plus<int>());
    return 0;
}
```

CUDA Thrust

Various Algorithms
(32M integer samples)



Sort
(32M integer samples)



- CUDA 4.1 on Tesla M2090, ECC on
- MKL 10.2.3, TYAN FT72-B7015 Xeon x5680 Six-Core @ 3.33 GHz

• Performance may vary based on OS ver. and motherboard config.

Lapack for CUDA: CULA Library

<http://www.culatools.com>

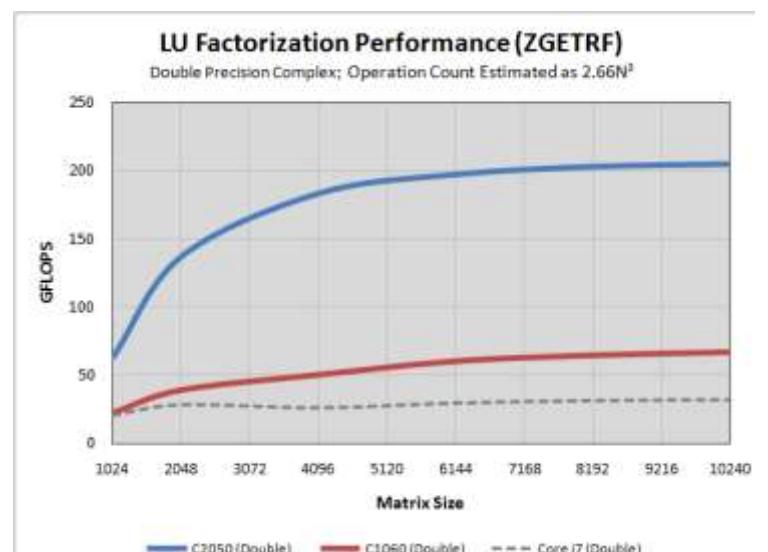
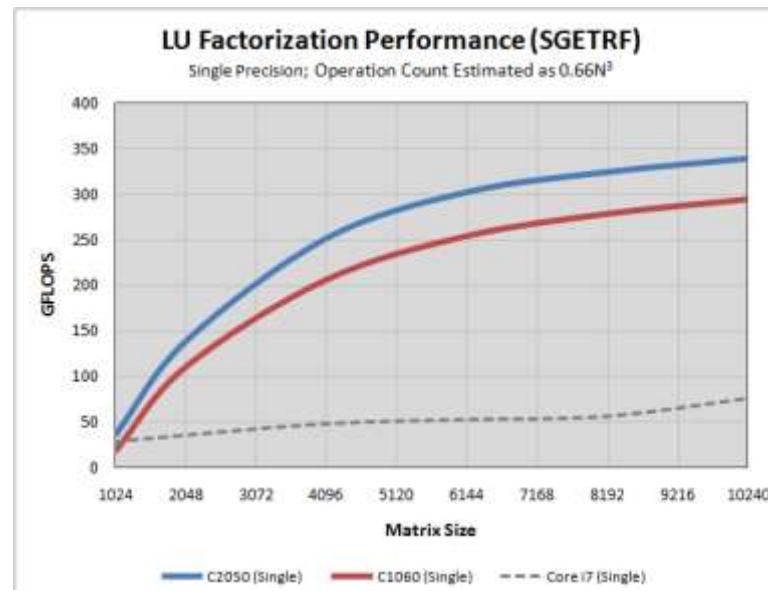
Proprietary library that implements the LAPACK in CUDA, which is available in several versions.

The speed-up of the picture on the right refers to:

CPU: Quad-core Intel Core i7 930 @ 2.8 GHZ CPU

GPU: NVIDIA Tesla C1060

GPU: NVIDIA Tesla C2050



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- ✓ P. Micikevicius, **Fundamental and Analysis-Driven Optimization**, GPU Technology Conference 2010 (GTC 2010)
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