



NVIDIA PROFILING TOOLS

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PROGRAMMING THE NVIDIA PLATFORM

CPU, GPU, and Network

ACCELERATED STANDARD LANGUAGES

ISO C++, ISO Fortran

```
std::transform(par, x, x+n, y, y,
    [=](float x, float y){ return y +
a*x; })
;
```

```
do concurrent (i = 1:n)
    y(i) = y(i) + a*x(i)
enddo
```

```
import cunumeric as np
...
def saxpy(a, x, y):
    y[:] += a*x
```

INCREMENTAL PORTABLE OPTIMIZATION

OpenACC, OpenMP

```
#pragma acc data copy(x,y) {
...
std::transform(par, x, x+n, y, y,
    [=](float x, float y){
        return y + a*x;
});
...
}

#pragma omp target data map(x,y) {
...
std::transform(par, x, x+n, y, y,
    [=](float x, float y){
        return y + a*x;
});
...
}
```

PLATFORM SPECIALIZATION

CUDA

```
__global__
void saxpy(int n, float a,
            float *x, float *y) {
    int i = blockIdx.x*blockDim.x +
            threadIdx.x;
    if (i < n) y[i] += a*x[i];
}

int main(void) {
    ...
    cudaMemcpy(d_x, x, ...);
    cudaMemcpy(d_y, y, ...);

    saxpy<<<(N+255)/256,256>>>(...);

    cudaMemcpy(y, d_y, ...);
```

ACCELERATION LIBRARIES

Core

Math

Communication

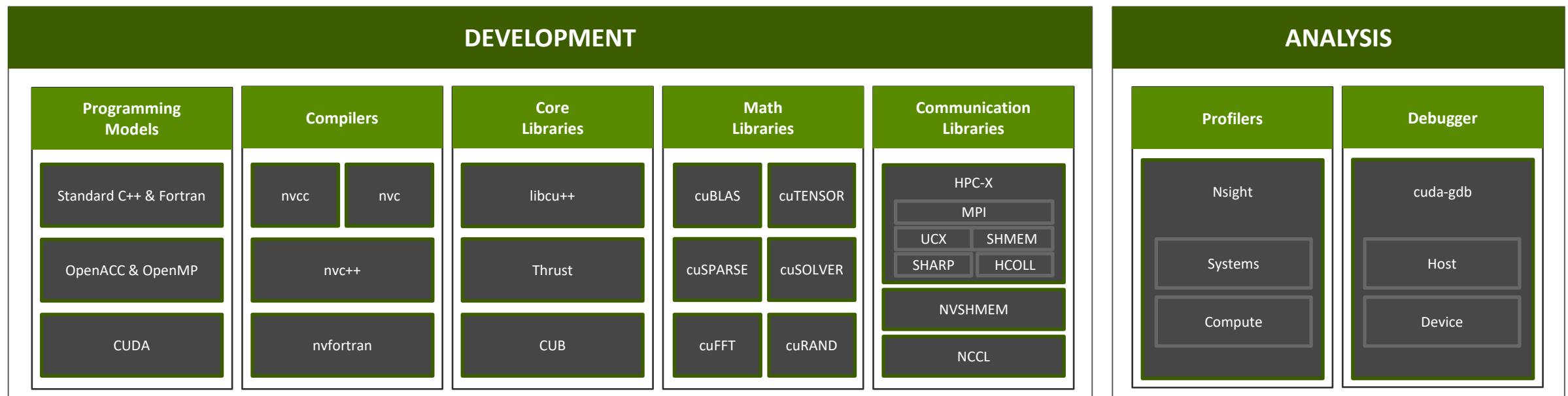
Data Analytics

AI

Quantum

NVIDIA HPC SDK

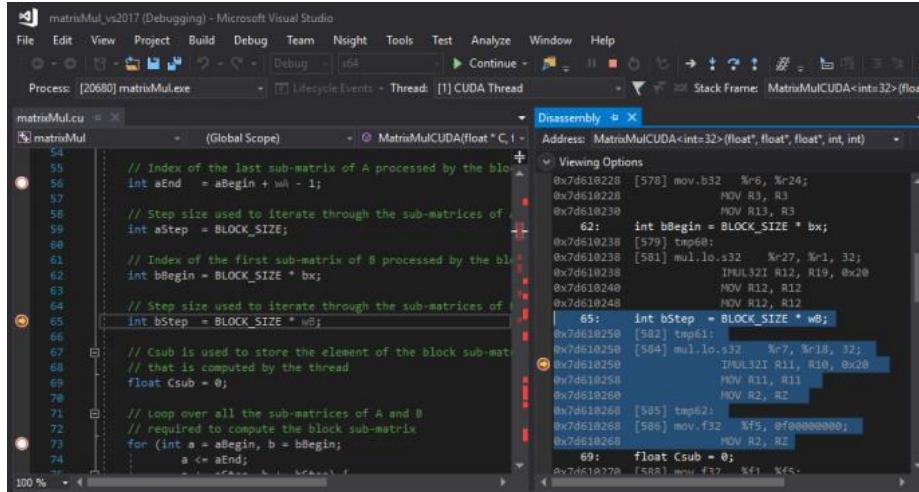
Available at developer.nvidia.com/hpc-sdk, on NGC, via Spack, and in the Cloud



Develop for the NVIDIA Platform: GPU, CPU and Interconnect
Libraries | Accelerated C++ and Fortran | Directives | CUDA
7-8 Releases Per Year | Freely Available

DEVELOPER TOOLS

Debuggers: cuda-gdb, Nsight Visual Studio Edition



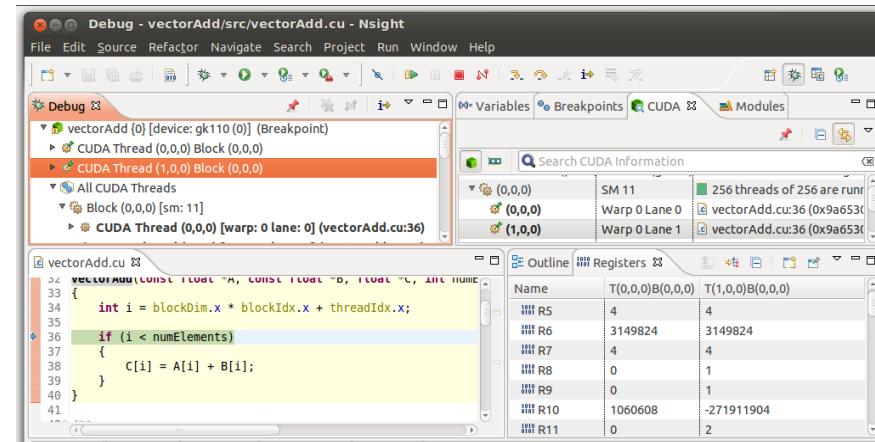
Profilers: Nsight Systems, Nsight Compute, CUPTI, [NVIDIA Tools eXtension \(NVTX\)](#)



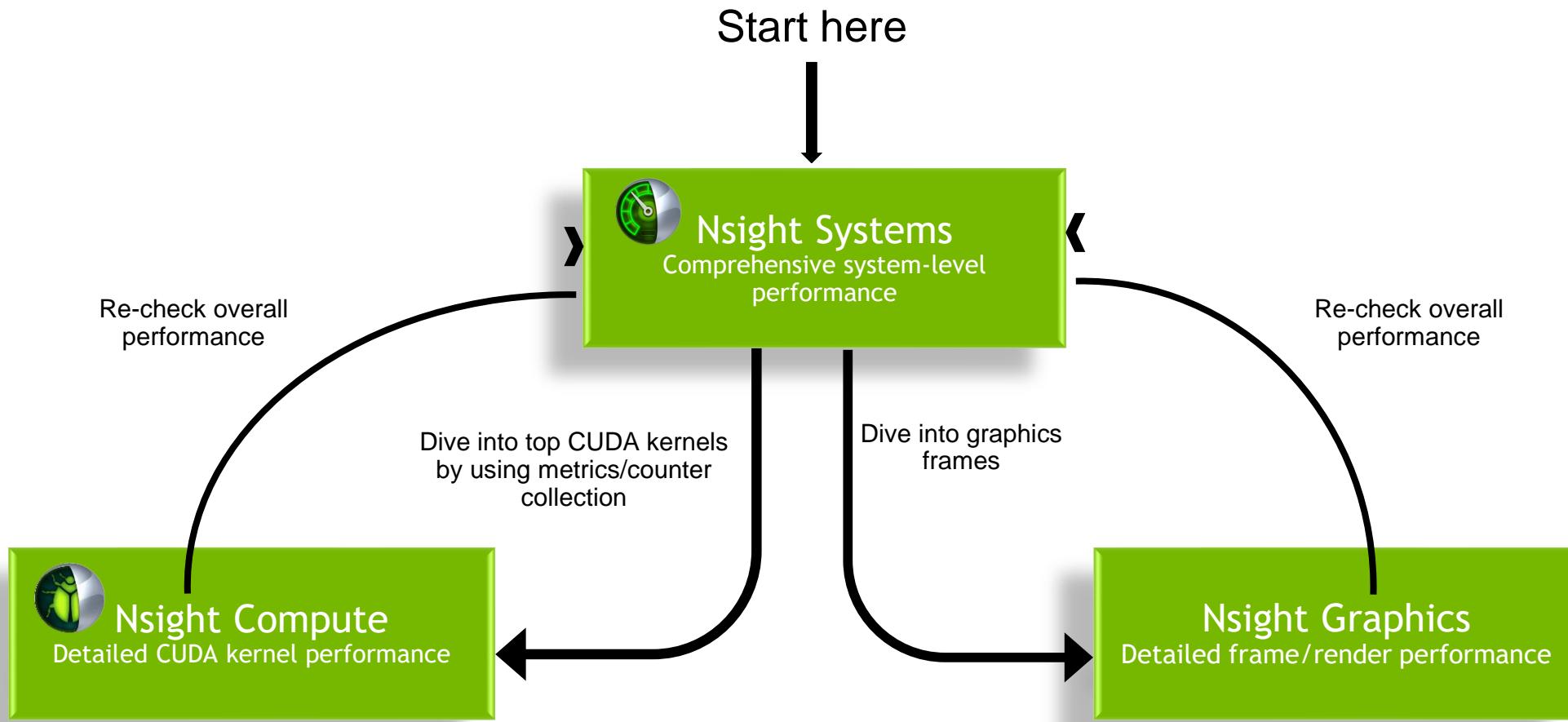
Correctness Checker: Compute Sanitizer

```
$ compute-sanitizer --leak-check full memcheck_demo
===== COMPUTE-SANITIZER
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 bytes
=====      at 0x60 in memcheck_demo.cu:6:unaligned_kernel(void)
=====      by thread (0,0,0) in block (0,0,0)
=====      Address 0x400100001 is misaligned
```

IDE integrations: Nsight Eclipse Edition
Nsight Visual Studio Edition
Nsight Visual Studio Code Edition



NSIGHT TOOLS WORKFLOW





NSIGHT SYSTEMS

System Profiler

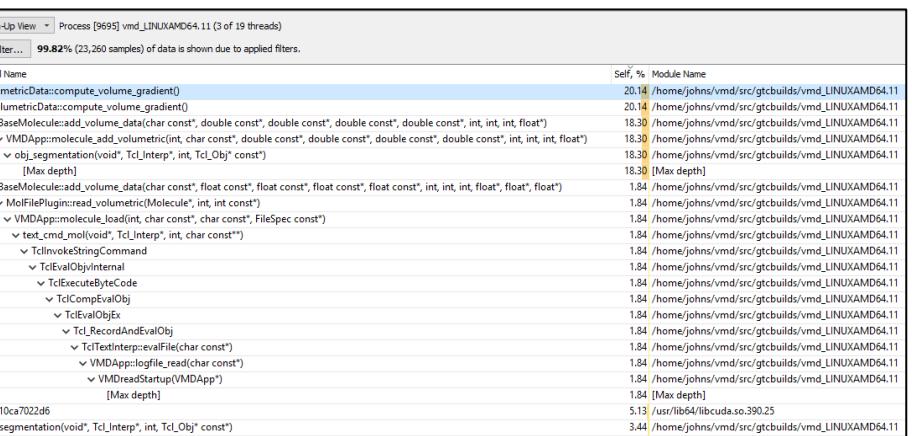
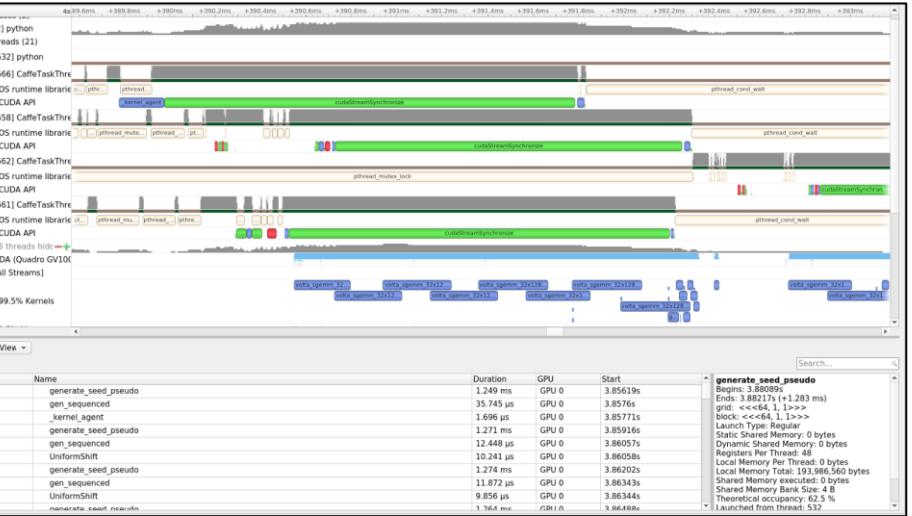
Key Features:

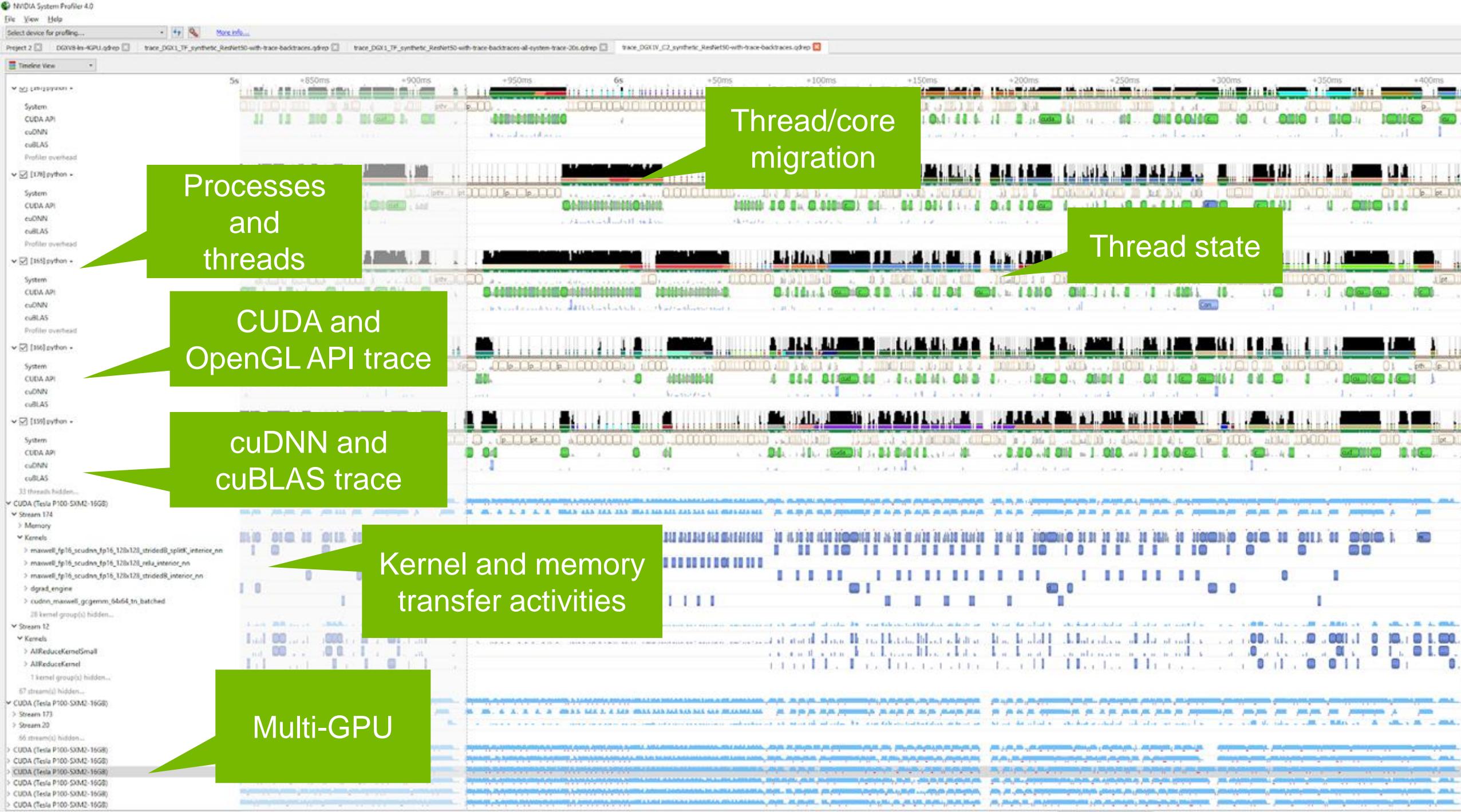
- System-wide application algorithm tuning
 - Multi-process tree support
- Locate optimization opportunities
 - Visualize millions of events on a very fast GUI timeline
 - Or gaps of unused CPU and GPU time
- Balance your workload across multiple CPUs and GPUs
 - CPU algorithms, utilization and thread state
 - GPU streams, kernels, memory transfers, etc
- Command Line, Standalone, IDE Integration

OS: Linux (x86, Power, Arm SBSA, Tegra), Windows, MacOSX (host)

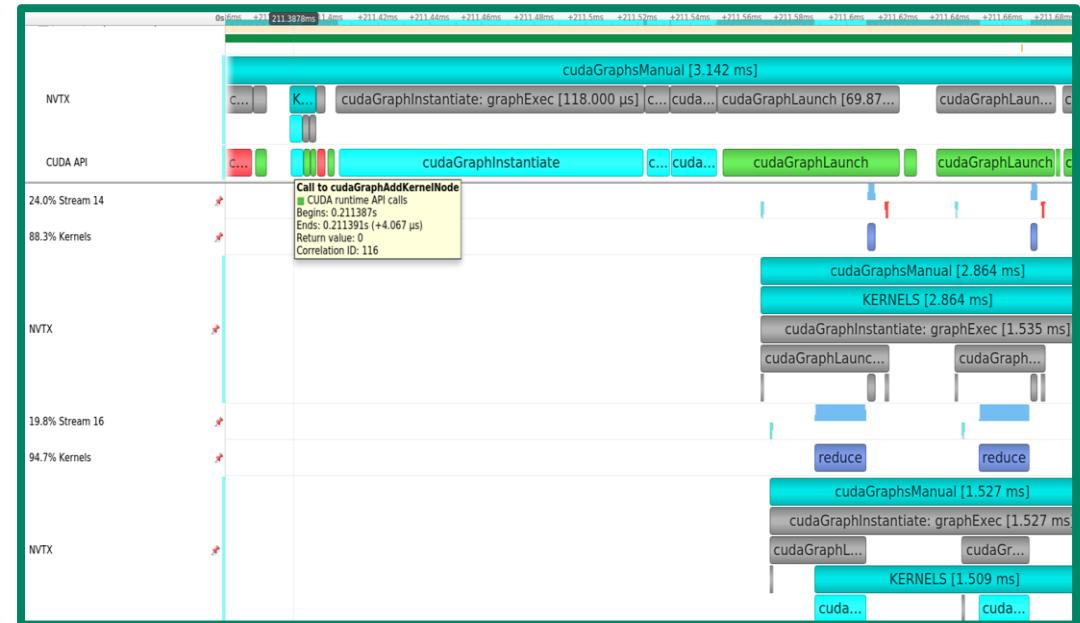
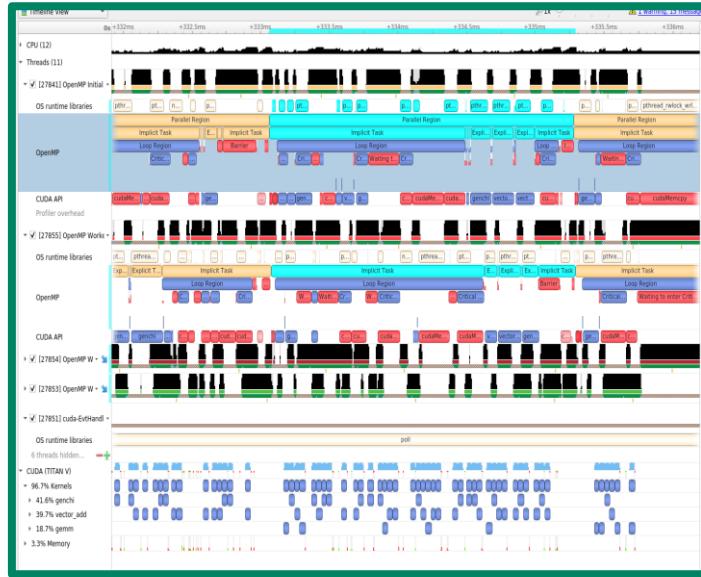
GPUs: Pascal+

Docs/product: <https://developer.nvidia.com/nsight-systems>



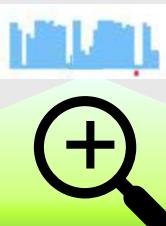


ZOOM/FILTER TO EXACT AREAS OF INTEREST



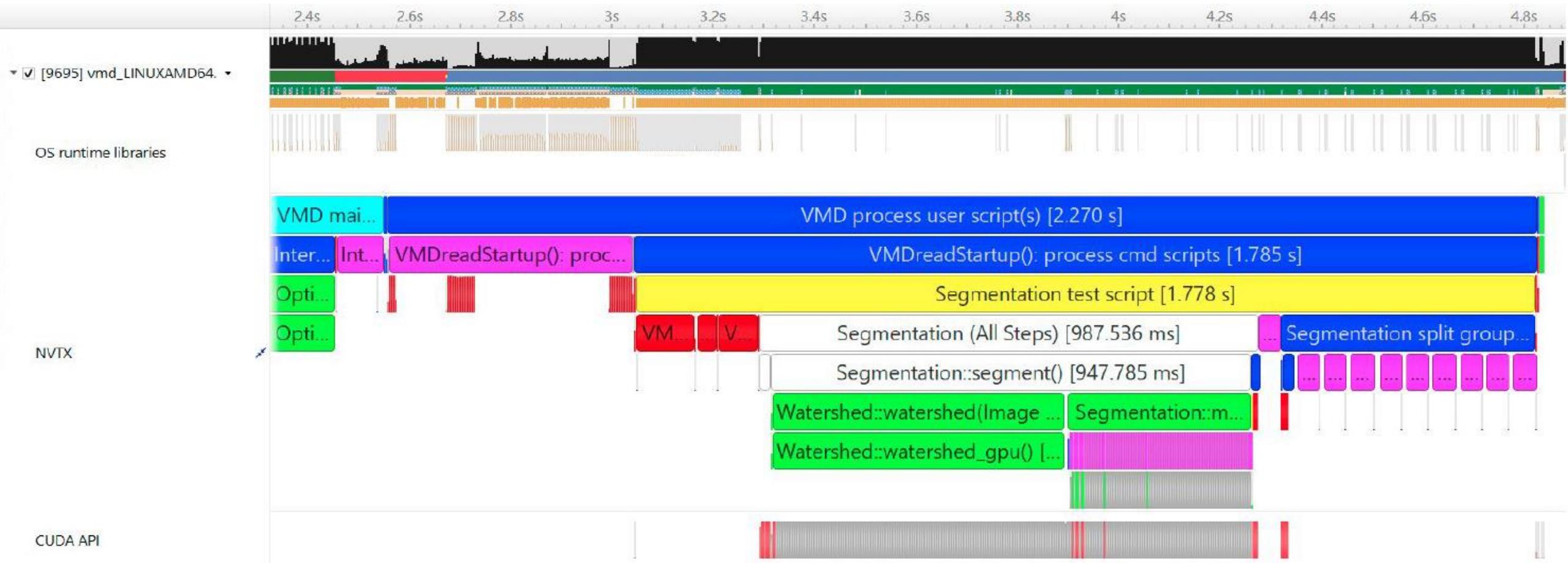
- Zoom in valleys to find gaps!

CUDA (Graphics Device, 0001:01:00.0)



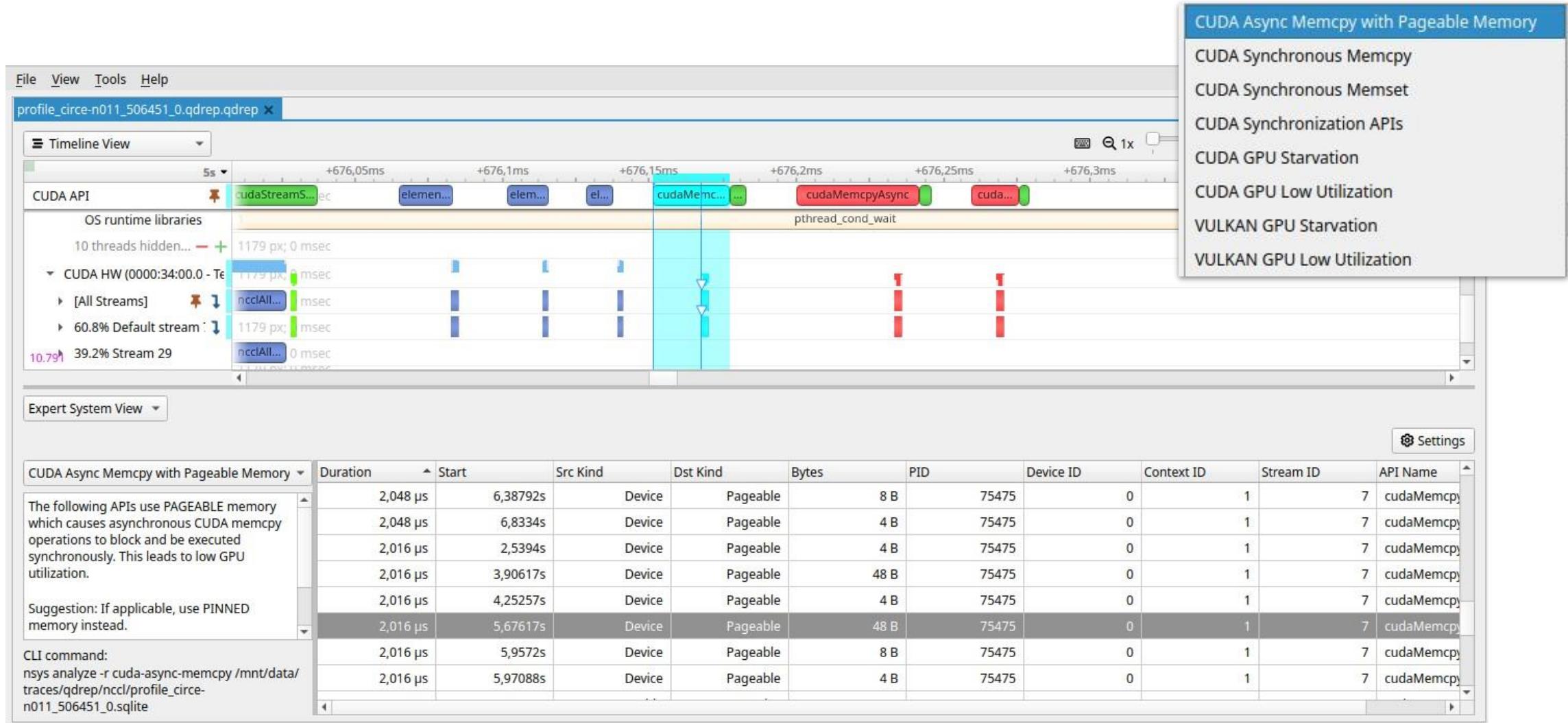
NVTX: NVIDIA TOOLS EXTENSIONS

Code Annotation API



EXPERT SYSTEMS & STATISTICS

Built-in Data Analytics with Advice



MULTI-REPORT TILING

Visualize More Parallel Activity

Open multiple reports

Loaded on same timeline based on wall-clock



APPLICATION PROFILES WITH NSIGHT SYSTEMS

```
$ nsys profile -o report --stats=true ./myapp.exe
```

- Generated file: report.qdrep (or report.nsys-rep)
Open for viewing in the Nsight Systems UI
- When using MPI, recommended to use *nsys* after mpirun/srun:

```
$ mpirun -n 4 nsys profile ./myapp.exe
```

PROFILING DL MODELS

- Pytorch
 - DNN Layer annotations are disabled by default
 - `++ "with torch.autograd.profiler.emit_nvtx():"`
 - Manually with `torch.cuda.nvtx.range_(push/pop)`
 - TensorRT backend is already annotated
- Tensorflow
 - Annotated by default with NVTX in NVIDIA TF containers
 - `TF_DISABLE_NVTX_RANGES=1` to disable for production



NSIGHT COMPUTE

Kernel Profiling Tool

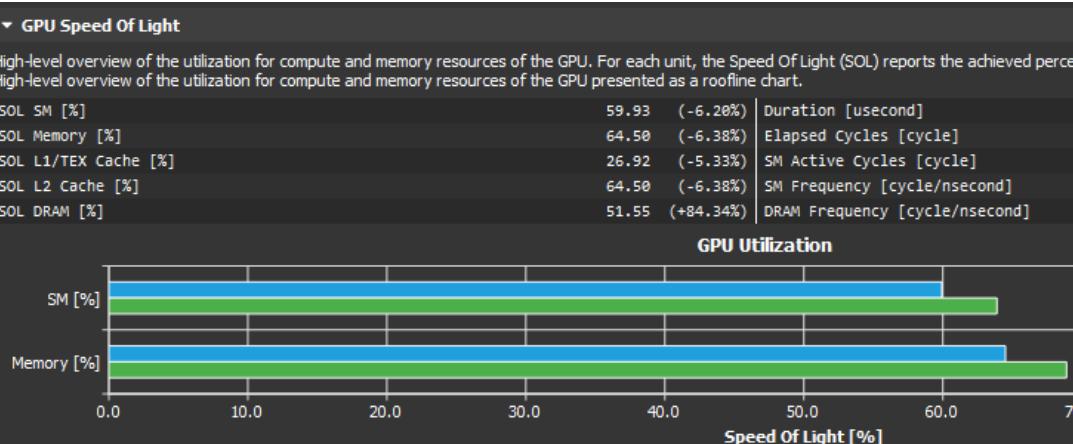
Key Features:

- Interactive CUDA API debugging and kernel profiling
- Built-in rules expertise
- Fully customizable data collection and display
- Command Line, Standalone, IDE Integration, Remote Targets

OS: Linux (x86, Power, Tegra, Arm SBSA), Windows, MacOSX
(host only)

GPUs: Volta+

Docs/product: <https://developer.nvidia.com/nsight-compute>



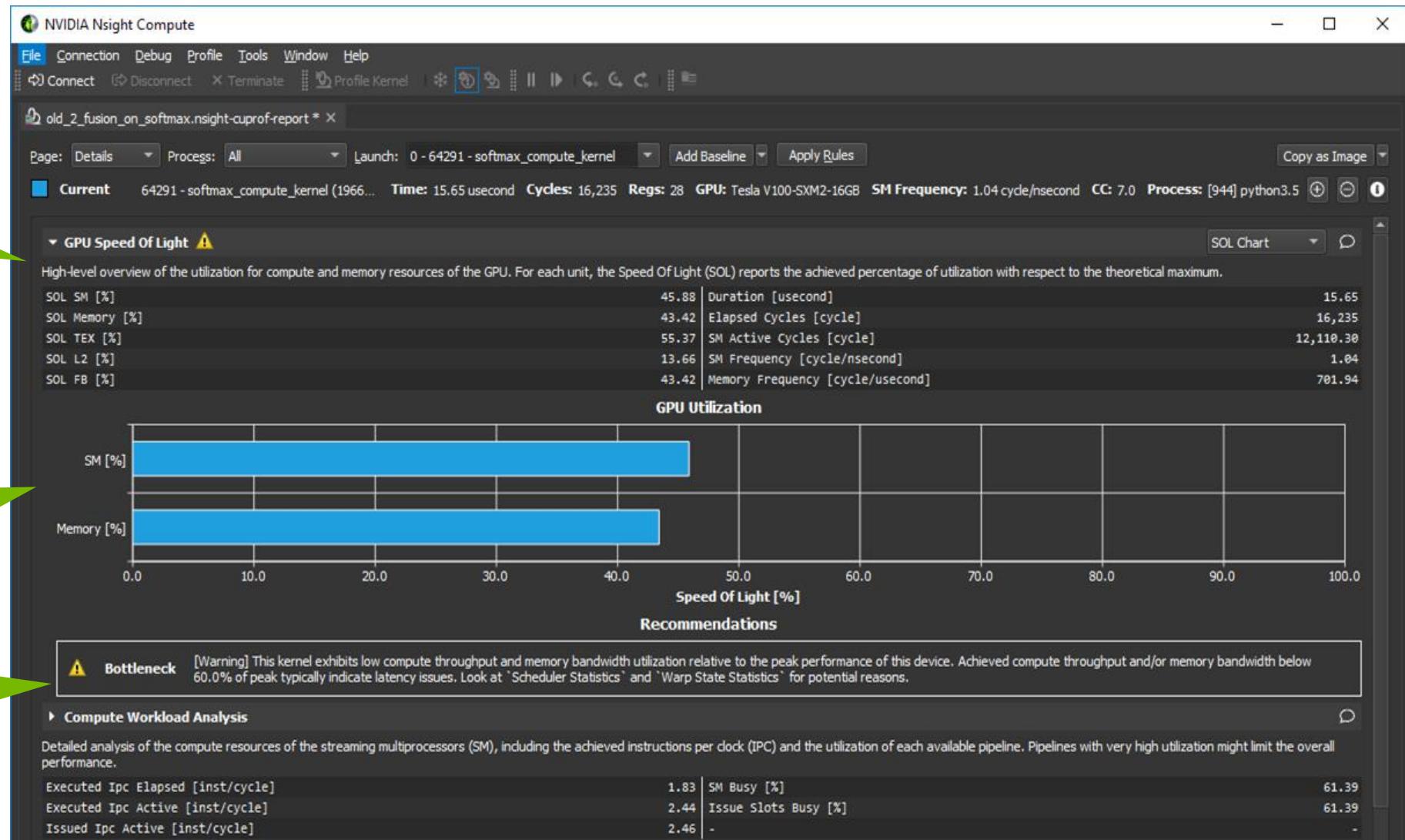
inst_executed [inst]	63,021,056 (284 instances)
l1tex_data_bank_conflicts_pipe_lsu_mem_shared_op_Id.sum	0
l1tex_data_bank_conflicts_pipe_lsu_mem_shared_op_st.sum	0
l1tex_data_bank_reads.avg.pct_of_peak_sustained_elapsed [%]	9.66
l1tex_data_bank_writes.avg.pct_of_peak_sustained_elapsed [%]	3.23
l1tex_data_pipe_lsu_wavefronts.avg.pct_of_peak_sustained_elapsed [%]	46.16
l1tex_data_pipe_lsu_wavefronts_mem_shared_cmd_read.sum	25,165,824
l1tex_data_pipe_lsu_wavefronts_mem_shared_cmd_read.sum.pct_of_peak_sustained_active [%]	40.75
l1tex_data_pipe_lsu_wavefronts_mem_shared_cmd_write.sum	2,097,152
l1tex_data_pipe_lsu_wavefronts_mem_shared_cmd_write.sum.pct_of_peak_sustained_active [%]	3.40
l1tex_data_pipe_tex_wavefronts.avg.pct_of_peak_sustained_elapsed [%]	0
l1tex_f_wavefronts.avg.pct_of_peak_sustained_elapsed [%]	0.00
l1tex_lsu_writeback_active.avg.pct_of_peak_sustained_elapsed [%]	42.59
l1tex_lsu_writeback_active.sum [cycle]	27,803,648
l1tex_lsu_writeback_active.sum.pct_of_peak_sustained_active [%]	45.03
l1tex_lsu_requests.avg.pct_of_peak_sustained_elapsed [%]	66.00
l1tex_m_l1tex2xbar_req_cycles_active.avg.pct_of_peak_sustained_elapsed [%]	3.40
l1tex_m_l1tex2xbar_write_bytes.sum [Mbyte]	4.19
l1tex_m_l1tex2xbar_write_bytes_mem_global_op_red.sum [byte]	0



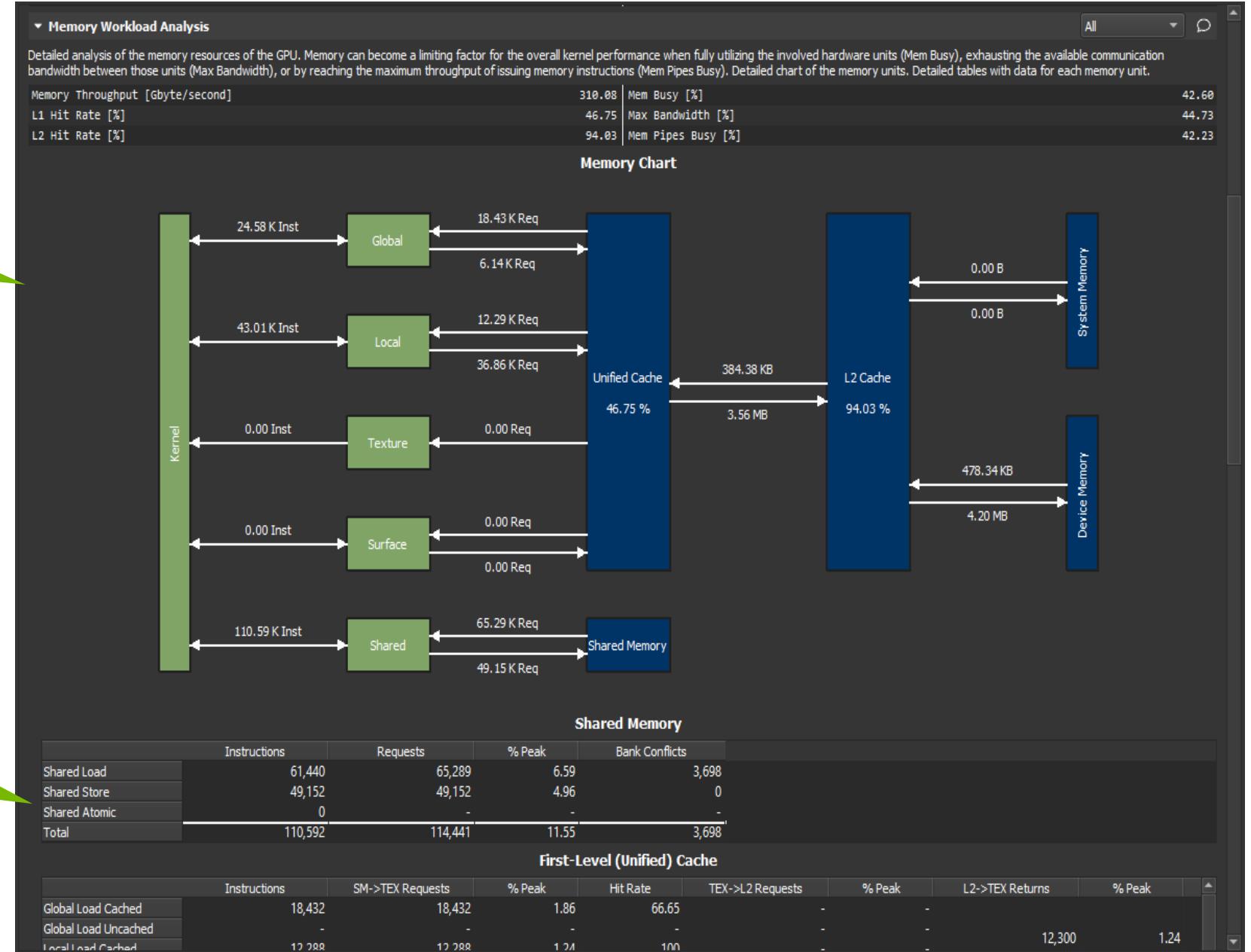
Targeted metric sections

Customizable data collection and presentation

Built-in expertise for Guided Analysis and optimization



Visual memory analysis chart



Metrics for peak performance ratios

The screenshot displays the Nsight Compute interface with two main panes: "Source" and "Sampling Data (All)".

Source Metrics: The left pane shows the C++ source code for a softmax computation kernel. A green callout box highlights the instruction `cuda::Reduce1D<red::maximum, x_bits>(smem);` at line 249, which is highlighted with a blue selection bar. Below this instruction, a green box contains the text "Source metrics per instruction".

Sampling Data Analysis: The right pane shows the "Sampling Data (All)" for the kernel. A green callout box highlights the instruction `@!P0 LDS.U R4, [R14+0x100]` at line 141, which is highlighted with a blue selection bar. A tooltip for this instruction provides performance metrics: Total Sample Count: 111, Barrier: 43 (38.7%), Mio Throttle: 21 (18.9%), Not Selected: 8 (7.2%), Selected: 7 (6.3%), Short Scoreboard: 16 (14.4%), and Wait: 16 (14.4%). Another green callout box highlights the heatmap at the bottom of the pane, with the text "Metric heatmap to quickly identify hotspots".

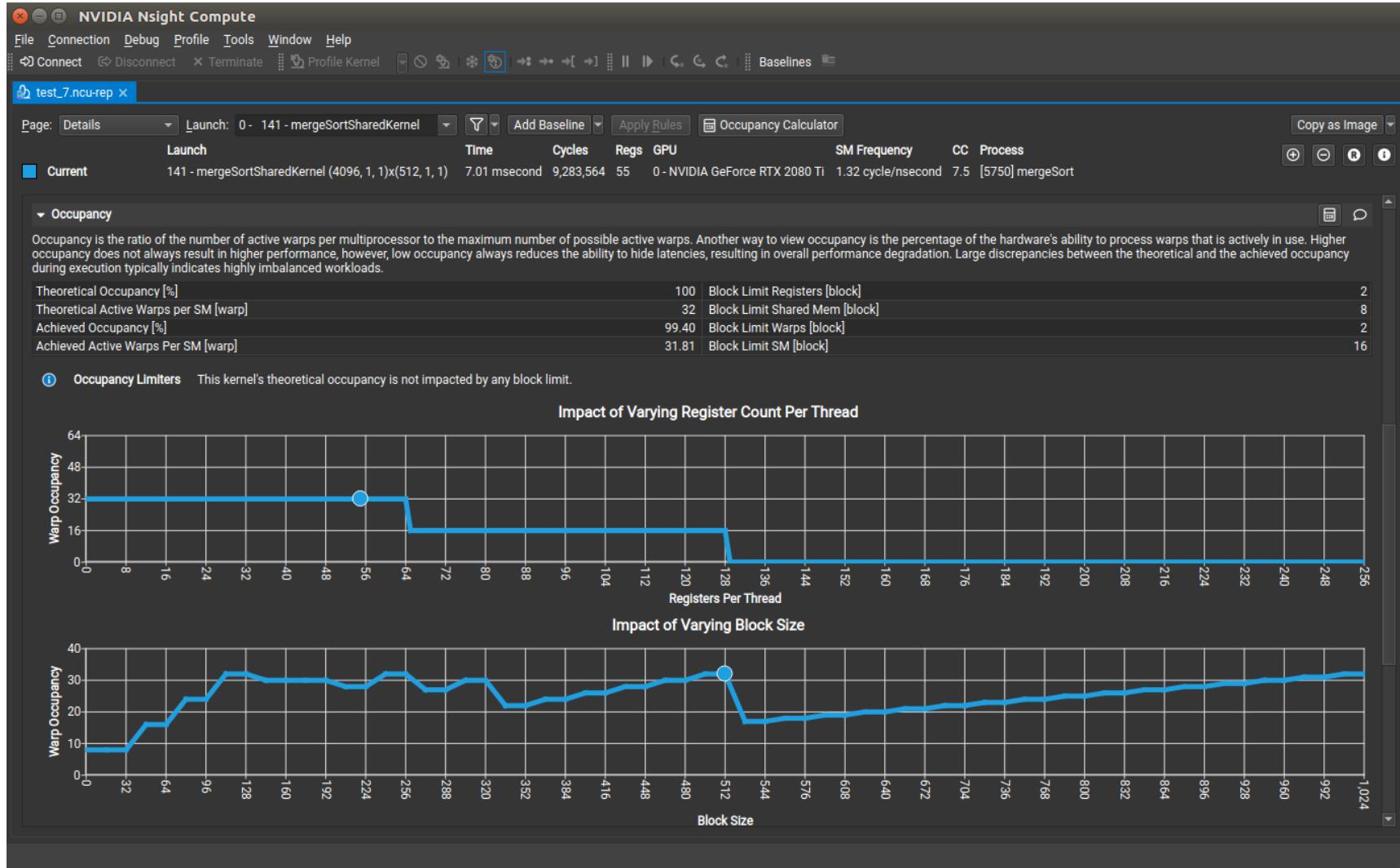
Source/PTX/SASS analysis and correlation

Source metrics per instruction

Metric heatmap to quickly identify hotspots

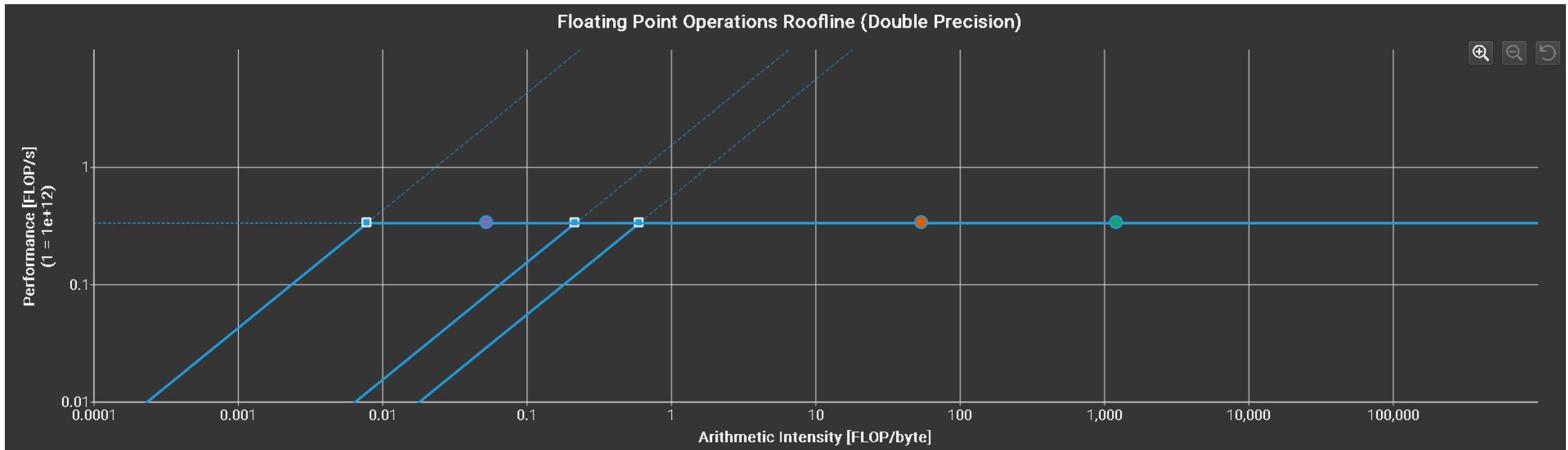
OCCUPANCY CALCULATOR

Model Hardware Usage and Identify Limiters



- Model theoretical hardware usage
- Understand limitations from hardware vs. kernel parameters
- Configure model to vary HW and kernel parameters
- Opened from an existing report or as a new activity

HIERARCHICAL ROOFLINE



- Visualize multiple levels of the memory hierarchy
- Identify bottlenecks caused by memory limitations
- Determine how modifying algorithms may (or may not) impact performance

Sections/Rules Info			
Sections/Rules		Reload	<input checked="" type="checkbox"/> Enable All <input type="checkbox"/> Disable All <input type="checkbox"/> Restore
Enter filter			
Name	Priority	Description	
GPU Speed Of Light Throughput (1)	10	High-level overview of the throughput for computation...	
GPU Speed Of Light Roofline Chart (1)	11	High-level overview of the utilization for computation...	
GPU Speed Of Light Hierarchical Roofline Chart (Double Precision)	12	High-level overview of the utilization for computation...	
GPU Speed Of Light Hierarchical Roofline Chart (Half Precision)	12	High-level overview of the utilization for computation...	
GPU Speed Of Light Hierarchical Roofline Chart (Single Precision)	12	High-level overview of the utilization for computation...	
GPU Speed Of Light Hierarchical Roofline Chart (Tensor Core)	12	High-level overview of the utilization for computation...	
Compute Workload Analysis (2)	20	Detailed analysis of the compute resources of the system...	

KERNEL PROFILES WITH NSIGHT COMPUTE

```
$ ncu -k mykernel -o report ./myapp.exe
```

- Generated file: report.ncu-report
 - Open for viewing in the Nsight Compute UI
- (Without the -k option, Nsight Compute will profile everything and take a long time)

CUDA-GDB

Command-Line and IDE Back-End Debugger

- Unified CPU and CUDA Debugging
- CUDA-C/SASS support
- Built on GDB and uses many of the same CLI commands
- Local/Remote connection support

```
(cuda-gdb) info cuda threads breakpoint all
      BlockIdx ThreadIdx          Virtual PC Dev SM Wp Ln      Filename Line
Kernel 0
(1,0,0)   (0,0,0) 0x0000000000948e58  0 11 0 0 infoCommands.cu 12
(1,0,0)   (1,0,0) 0x0000000000948e58  0 11 0 1 infoCommands.cu 12
(1,0,0)   (2,0,0) 0x0000000000948e58  0 11 0 2 infoCommands.cu 12
(1,0,0)   (3,0,0) 0x0000000000948e58  0 11 0 3 infoCommands.cu 12
(1,0,0)   (4,0,0) 0x0000000000948e58  0 11 0 4 infoCommands.cu 12
(1,0,0)   (5,0,0) 0x0000000000948e58  0 11 0 5 infoCommands.cu 12

(cuda-gdb) info cuda threads breakpoint 2 lane 1
      BlockIdx ThreadIdx          Virtual PC Dev SM Wp Ln      Filename Line
Kernel 0
(1,0,0)   (1,0,0) 0x0000000000948e58  0 11 0 1 infoCommands.cu 12
```

COMPUTE SANITIZER

Automatically Scan for Bugs and Memory Issues

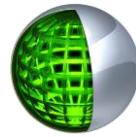
- Compute Sanitizer checks correctness issues via sub-tools:

- **Memcheck** - Memory access error and leak detection tool.
- **Racecheck** - Shared memory data access hazard detection tool.
- **Initcheck** - Uninitialized device global memory access detection tool.
- **Synccheck** - Thread synchronization hazard detection tool.

- <https://github.com/NVIDIA/compute-sanitizer-samples>

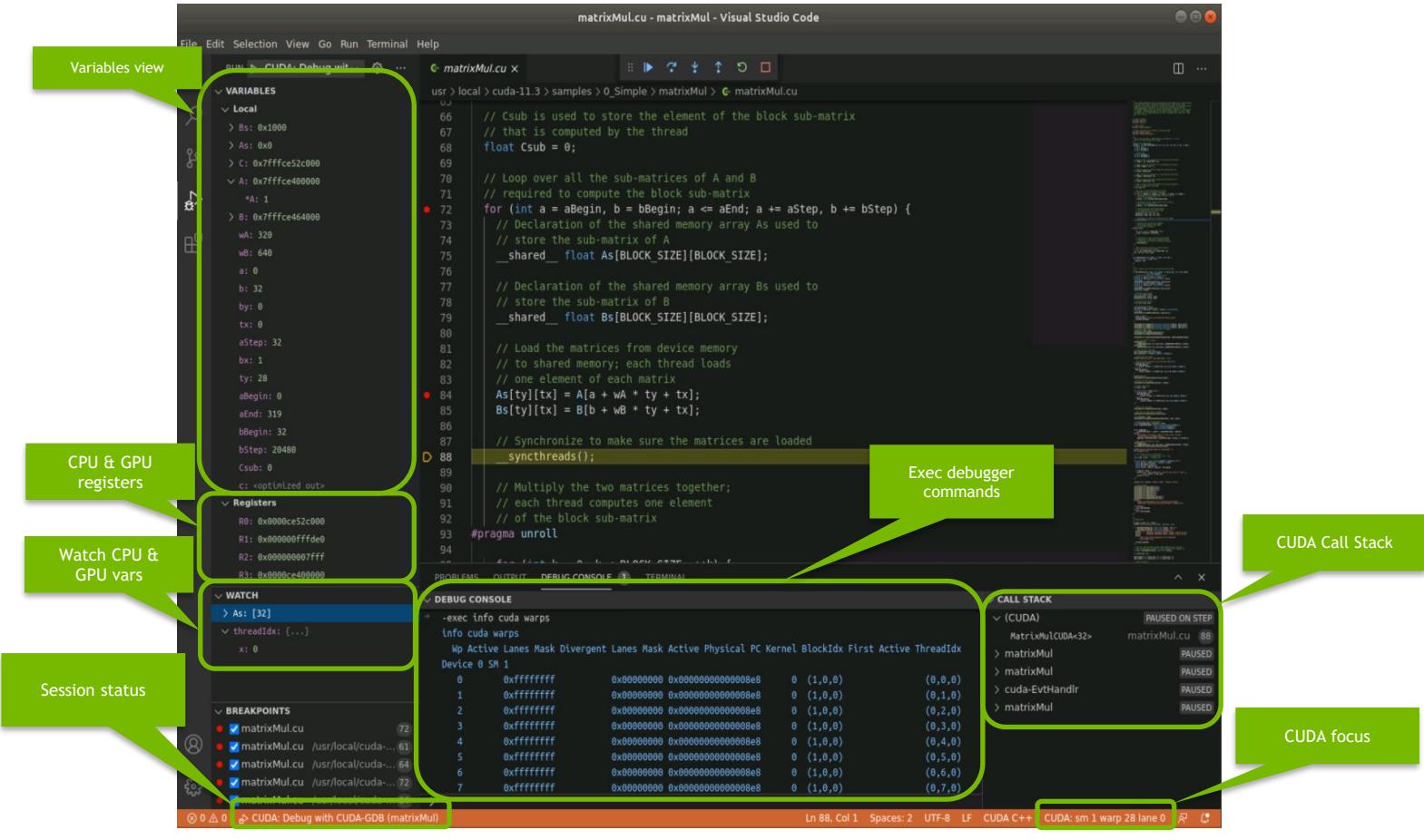
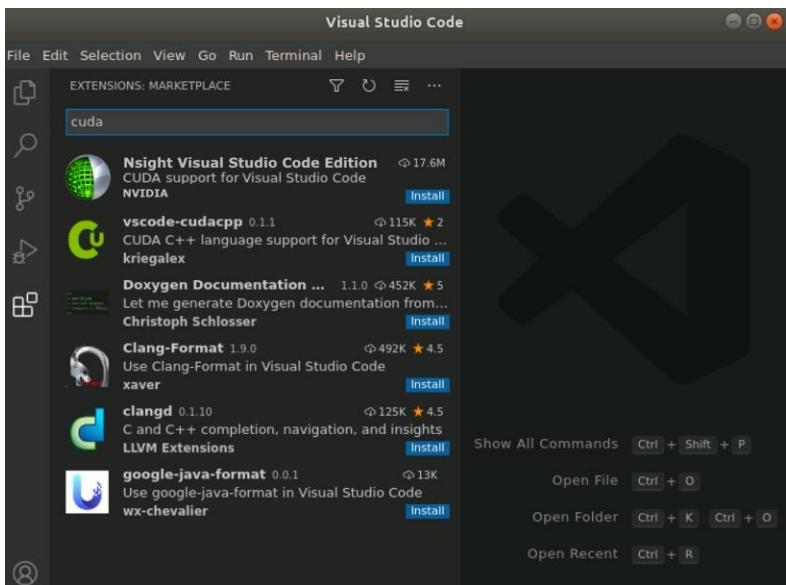
```
$ make run_memcheck
/usr/local/cuda/compute-sanitizer/compute-sanitizer --destroy-on-device-error kernel memcheck_demo
=====
===== COMPUTE-SANITIZER
Mallocing memory
=====
===== Invalid __global__ write of size 4 bytes
===== at 0x70 in unaligned_kernel()
===== by thread (0,0,0) in block (0,0,0)
===== Address 0x7f671ac00001 is misaligned
===== and is inside the nearest allocation at 0x7fb654c00000 of size 4 bytes
===== Saved host backtrace up to driver entry point at kernel launch time
===== Host Frame: [0x2774ec]
=====           in /lib/x86_64-linux-gnu/libcuda.so.1
===== Host Frame: __cudart803 [0xfcdb]
=====           in /home/cuda/github/compute-sanitizer-samples/Memcheck/memcheck_demo
===== Host Frame:cudaLaunchKernel [0x6a578]
=====           in /home/cuda/github/compute-sanitizer-samples/Memcheck/memcheck_demo
===== Host Frame:cudaError cudaLaunchKernel<char>(char const*, dim3, dim3, void**, unsigned
=====           in /home/cuda/github/compute-sanitizer-samples/Memcheck/memcheck_demo
===== Host Frame:__device_stub_Z16unaligned_kernelv() [0xb22e]
=====           in /home/cuda/github/compute-sanitizer-samples/Memcheck/memcheck_demo
===== Host Frame:unaligned_kernel() [0xb28c]
=====           in /home/cuda/github/compute-sanitizer-samples/Memcheck/memcheck_demo
===== Host Frame:run_unaligned() [0xaf55]
=====           in /home/cuda/github/compute-sanitizer-samples/Memcheck/memcheck_demo
===== Host Frame:main [0xb0e2]
=====           in /home/cuda/github/compute-sanitizer-samples/Memcheck/memcheck_demo
===== Host Frame:.../sysdeps/nptl/libc_start_call_main.h:58:__libc_start_call_main [0x2dfd0]
=====           in /lib/x86_64-linux-gnu/libc.so.6
```

NSIGHT VISUAL STUDIO CODE EDITION



Visual Studio Code extensions that provides:

- CUDA code syntax highlighting
- CUDA code completion
- Build warning/errors
- Debug CPU & GPU code
- Remote connection support via SSH
- Available on the VS Code Marketplace now!



<https://developer.nvidia.com/nsight-visual-studio-code-edition>



ADDITIONAL RESOURCES

- Sessions
 - [A41100](#) - CUDA: New Features and Beyond
 - [A41131](#) - Developing Efficient CUDA Kernels for Fourth-Generation Tensor Cores
- Labs
 - [DLIT41277](#) - Optimizing CUDA Machine Learning Codes with Nsight Profiling Tools
 - [DLIT41274](#) - Debugging and Analyzing Correctness of CUDA Applications
 - [DLIT41276](#) - Developer Tools Fundamentals for Ray Tracing using NVIDIA Nsight Graphics and NVIDIA Nsight Systems
- Ampere Architecture Detailed Blog
 - [NVIDIA Ampere Architecture In-Depth](#)
- Developer Tools are free and packaged in the latest version of the CUDA Toolkit
 - <https://developer.nvidia.com/cuda-downloads>
- Support is available via:
 - <https://forums.developer.nvidia.com/c/development-tools/>
- More information at:
 - <https://developer.nvidia.com/tools-overview>

HANDS-ON

/lus/eagle/projects/SDL_Workshop/jacobi

- Solving Laplace Equation with Jacobi Iterations

```
f_{i,j} = (f_{i+1,j} + f_{i-1,j} + f_{i,j+1} + f_{i,j-1}) / 4

// set initial conditions at f[0][:], f[N-1][:], f[:,0], f[:,N-1]

while (error > tolerance):
    error = 0
    for i = 1, N-2:
        for j = 1, N-2:
            f[i][j] = 0.5 * (f_old[i+1][j] + f_old[i-1][j] + f_old[i][j+1] + f_old[i][j-1])
            error += (f[i][j] - f_old[i][j]) * (f[i][j] - f_old[i][j])
    swap(f_old, f)
```

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KMA and NOAA Workshop on AI for Weather and Climate

Date(s): Oct 21, 2022 - Oct 21, 2022

Event Focus: **HPC+AI**

Asia-Pacific



Applications [Open](#)

NVIDIA/HLRs SciML GPU Bootcamp

Date(s): Oct 24, 2022 - Oct 25, 2022

Event Focus: **HPC+AI**

Europe/Middle East/Africa



Applications [Open](#)

NCI GPU Hackathon 2022

Date(s): Oct 25, 2022 - Nov 4, 2022

Event Focus: **HPC+AI**

Asia-Pacific



Applications [Closed](#)

TWCC Open Hackathon 2022

Date(s): Nov 23, 2022 - Dec 2, 2022

Event Focus: **HPC+AI**

Asia-Pacific



Applications [Open](#)

NERSC GPU Hackathon 2022

Date(s): Nov 30, 2022 - Dec 8, 2022

Event Focus: **HPC+AI**

North America/Latin America



Applications [Open](#)

Westlake University GPU Hackathon 2022

Date(s): Nov 29, 2022 - Dec 9, 2022

Event Focus: **HPC+AI**

Asia-Pacific



Applications [Open](#)



