$10_HandsOnNsight_ncu$

June 21, 2022







#

Hands-On Session with Nsight Systems and Nsight Compute

By: Daniel Howard, Consulting Services Group, CISL & NCAR

dhoward@ucar.edu

Date: June 16th, 2022

In this notebook we explore profiling of the mini-app MiniWeather to present profiling techniques and code examples. We will cover:

- Overview of Profiling and Performance Sampling Tools
 - Typical development workflows with profiling tools
- NSight Compute for Individual GPU Kernel Performance Analysis
 - How to generate ncu reports and command line parameters
 - Overview of GPU kernel profiling data and source code timing heatmaps
 - External resources for interpreting neu reports data

Head to the NCAR JupyterHub portal and start a JupyterHub session on Casper login (or batch nodes using 1 CPU, no GPUs) and open the notebook in 10_HandsOnNsight/ncu/10_HandsOnNsight_ncu.ipynb. Be sure to clone (if needed) and update/pull the NCAR GPU_workshop directory.

Use the JupyterHub GitHub GUI on the left panel or the below shell commands git clone git@github.com:NCAR/GPU_workshop.git git pull

1 Workshop Etiquette

- Please mute yourself and turn off video during the session.
- Questions may be submitted in the chat and will be answered when appropriate. You may also raise your hand, unmute, and ask questions during Q&A at the end of the presentation.
- By participating, you are agreeing to UCAR's Code of Conduct
- Recordings & other material will be archived & shared publicly.
- Feel free to follow up with the GPU workshop team via Slack or submit support requests to support.ucar.edu
 - Office Hours: Asynchronous support via Slack or schedule a time with an organizer

1.1 Notebook Setup

Set the PROJECT code to a currently active project, ie UCIS0004 for the GPU workshop, and QUEUE to the appropriate routing queue depending on if during a live workshop session (gpuworkshop), during weekday 8am to 5:30pm MT (gpudev), or all other times (casper). Due to limited shared GPU resources, please use GPU_TYPE=gp100 during the workshop. Otherwise, set GPU_TYPE=v100 (required for gpudev) for independent work. See Casper queue documentation for more info.

1.2 What is a Profiler?

Profilers are tools that **samples** and measure performance characteristics of an executable across its runtime. This information is intended to aid program optimization and performance engineering.

Profiler software that are supported at NCAR include **Arm Map**, **Nsight Systems**, and **Nsight Compute**. All of these tools are able to analyze GPU code. Other profilers you may be aware of include TAU, Intel VTune Advisor, HPC Toolkit, and Vampir.

Today, we will focus on the NVIDIA Nsight profiling tools and usage techniques of these tools.

- **Nsight Systems** Provides a high level runtime and trace analysis of the program runtime via a measured timeline of various metrics and GPU kernels across a program.
- **Nsight Compute** Provides an in depth level assessment of individual GPU kernel performance and how various GPU resources are utilized across many different metrics.

▶ CPU (12)	A REAL PROPERTY AND A REAL	Shmem 2363 - 1	kernel_apsm_detect (1920), 1, 1)x Time	6.08 msecond Cyc	les: 3,557,30	5 Regs: 5
 Processes (2) 							
• • [26247] deepstream-app	COLORED D. C.	GPU Speed Of Ligh	٦C			63.34 1.6	3 019.)
 Threads (72) 		UL SM [%]				63 24 (+0	03.01%) Du
▶ ✓ [26270] dsexample_que +		L1/TEX Cache [%]	1			94.37 (- 1.29 (-6	4.47%) SM 52.82%) SM
▼ ✔ [26283] deepstream-ap -		DRAM [%]				0.15 (+14	(5.52%) DR GPU Utili
OS runtime libraries	Waiting (with callchain) Time: 9.21413s	. [pt					
NVTX	Call stack at 9.209s:	r(Batch=					
CUDA API	libc-2.23.so!syscall libglib-2.0.so.0.4800.2!g cond wait	.[trt] [%]					
CuDNN	libglib-2.0.so.0.4800.2!0x7f0d567e6						
CuBLAS	libnvdsgst infer.solgst nvinfer infe	rence_thread	10.0	20.0	30.0	40.0	Groot
Profiler overhead	libpthread-2.23.so!start_thread	3655				_	speed
70 threads hidden 🗕 🕂	libc-2.23.so!_clone					Re	commen
 CUDA (Quadro GV100, 0000:0 	D		Compute and Memory an	e well-balanced: 1	o reduce runtime, bo	th computatio	n and mem
 0.9% Context 2 		Heck	Analysis report sections.				
 100.0% Memory 	(a)						
100.0% HtoD memcpy		Ana	iysis The ratio of peak f	float (fp32) to dou	ible (fp64) performan	ce on this dev	ICE IS 32:1.
 99.1% Context 1 		loa	d Analysis				
 91.0% gstdsdsexample 		se	d [inst/cycle]			1.42 (+13	87.97%) SM
 100.0% Kernels 	so so so so so so	so	[inst/cycle]			1.43 (+13	8.67%) Is
94.4% sobelCalcKernel	so so so so so so so	o so so so so so	inst/cycle]			1.43 (+13	8.67%)
5.6% sobelApplyKernel		1.1.1.1.1.1	Analysis				
NVTX	gst_dsexample_tran gst_dsexample_tran So So So So	ple_tra gst_dsexample_trans gst_d o So So So So So So	byte/second]			467.71 (+14 94.72 (+	5.37%) Me 1.51%) Ma
6 streams hidden 💻	-Menty -	واليفيس يودر والمط	4			98.92 (-	0.90%) Me
	4		4				
Bottom-Up View V Process [2624	47] deepstream-app (21.9%, 72 of 72 thre	ads)	ler [warp]			7.99 (+3	87.74%) No
			uler [warn]			1 82 (+38	12 57%) On

Nsight Systems (left) shows a timeline of code runtime.

Nsight Compute (right) records and presents extensive performance statistics for individual kernels.

1.3 Profiling Documentation Resources

NVIDIA provides extensive documentation for each of these profilers. We will go over basic usage of these tools but to learn more and get the most out of Nsight, consult the below resources:

- Nsight Systems Main Documentation
- Nsight Compute Main Documentation
- Nsight Compute Profiling Guide
- Nsight Compute Training Resources Forum, Videos, and Blog Posts curated by NVIDIA

An excellent interactive step-by-step tutorial given by Max Katz (NVIDIA) using Nsight Compute to optimize an OpenACC kernel in the BerkeleyGW many-body perturbation theory software can be found at this Gitlab repository. A recorded video on this material is here.

Additionally, the CLI help pages via the -h flag for each profiler is a useful quick reference point. Run the below cells to view them.

[]: ncu -h

1.4 Profiling Workflow



When assessing performance of software, first profile the overall program with Nsight Systems. Then, expensive kernels can be identified and profiled using Nsight Compute.

Iteratively analyze and modify code to optimize performance, up to the amount of effort is worth-while.

1.5 Nsight Compute

After getting a sense of the overall performance of your program with Nsight Systems, use Nsight Compute to dive deeper into the performance of individual GPU kernels.

- CUDA kernel profiler (or CUDA kernels generated by OpenACC/OpenMP/Kokkos code)
- Curates **performance statistics** into targeted metrics sections
- Able to select amount of data to collect and how it's presented
 More detailed analysis has greater overhead with profiler usage
- Fully featured Command Line and User Friendly GUI interfaces
- Regularly updated and customizable Python based rules for **guided analysis** and post-processing



1.6 Preparing Code for Nsight Compute

When preparing code for Nsight Compute, an important compile option to add is -gpu=lineinfo. **DON'T USE -pg, -g, or -G flags**. The lineinfo flag allows the Source/SASS analysis section of Nsight Compute correlate performance information with specific lines of CUDA and/or Ope-nACC/OpenMP code.

Use the below cell to compile and re-compile MiniWeather after code changes are made. You may also modify the runtime parameters, grid size, and simulation time to investigate how different problem sizes impact performance. Review the generated GPU kernel specifications from the -Minfo=acc output.

Notably, only a short simulation time (enough to cover a few timesteps) is required for us to effectively analyze and optimize model performance.

1.7 Nsight Compute CLI Options

- -o <report-name> Writes output to a *.ncu-rep file to analyze via GUI

 Without -o, analysis is summarized in stdout.
- -f Force overwrite of output files
- -c or --launch-count Specifies the number of kernel launches to profile. Otherwise, all launched kernels are profiled
- -s or --launch-skip Skips a specified number of kernel launches. Useful for letting the GPU "warm-up"

- --set <arg> Sets the amount of data collected and kernel metrics measured, i.e. detailed, full, or others given from --list-sets flag
 - More data collected requires more redundant runs of GPU kernels and increases profiler overhead
- -k or --kernel-name Specifies the exact name (see nsys) of kernels to be profiled - Use -k regex:<expression> to filter kernels by a regex expression
- --nvtx Enables support for NVTX ranges
- --nvtx-include arg Filters profiled kernels based on NVTX ranges
- --import-source on Imports CUDA/source code directly into the report.

Generate Nsight Compute Report 1.8

Start with the final version of miniWeather_mpi_openacc.F90. As we analyze performance, use the generated report to inform code optimizations to experiment with.

First, use the submit script ncu_bash.sh to run Nsight Compute on MiniWeather by running command ncu <ncu options> <exec> <exec arguments>. Useful ncu options are listed above but also may be reviewed via ncu -h.

The first profile run of MiniWeather will profile all kernels using --sets full in order to make a baseline (requires redundantly running kernels 73-74 times). When changing code, modify the report filename when you re-run the below cell to help you keep track of reports between different code versions.

[]: gsub -g \$QUEUE -l gpu type=\$GPU TYPE -A \$PROJECT -v NCU REPORT="MW DivToMult" →ncu bash.sh

SHIFT + right click MW_baseline.ncu-rep in order to save the Nsight Compute report to your personal machine (or download the file from the left pane explorer). Install a free local Nsight Compute client to open the file. Alternatively, after setting module load nvhpc, you can run ncu-ui <report-name> over a terminal X session or VNC/FastX session on Casper.

Analysis of Nsight Compute Profiles 1.9

Depending on the option chosen for --set and number of metrics measured, the kernel profiling report will contain a selection of different sections for review covering performance metrics of each kernel profiled.

When using the GUI, guided analysis as alerted via exclamation point warning signs will suggest specific issues the profiler identifies and tries to suggest solutions. These are automatically triggered Python rules written by Nsight Compute maintainers and experts, which can be further customized or added to. If you need help interpretting this information, hover your mouse over a piece of information and an informative text box will appear to explain.

Below, we review a few important sections.



1.10 Nsight Compute - GPU Speed of Light

Mouse over each to see the associated metric

The GPU Speed of Light section highlights to what percentage is this kernel using the full capability of the GPU, both in terms of Streaming Multiprocesser (SM) occupancy and Memory Throughput.

1.11 Nsight Compute - Roofline Analysis



With at least --set detailed, a roofline plot is generated, used to determine if the kernel is compute bound or memory bound. Measurements are plotted as points, where distance from the roofline indicates potential for improvement at a given kerne;'s arithmentic intensity, calculations per total size of data loads and stores.

Memory bound kernels can perhaps benefit by assigning more compute operations per thread if possible. Compute bound kernels will likely require further analysis for optimization, typically by checking for warp stalls or coallesced memory issues.

1.12 Nsight Compute - Memory Workload Analysis

This section provides a detailed analysis of the memory resources of the GPU. In this case, Nsight Compute identifies that there is an imbalance of data movement between the L1 and L2 caches due to uncoalesced memory. To improve this, memory access patterns need to be re-designed within the source code and OpenACC kernel.

1.13 Nsight Compute - Source/SASS and Instruction Hotspots



Navigated to via the **Source Counters** section, a heatmap of resource usage and other metrics can be correlated to specific lines of code within the source files. This can more easily identify which specific areas of your program are causing poor performance.

1.14 Nsight Compute - Add a Baseline

Whenever profiling a program or specific kernel, it is vitally important to record and **set a baseline** to reference performance changes against. In Nsight Compute, set a baseline by clicking **Add Baseline** near the top of the main window within the Nsight Compute GUI. Note, you can add multiple "baselines" from multiple reports.

A MW_Baseline nou-rep									
Page: Details	Seesis * Besut 2- 1629-compute tendencies, 27 * 🗙 Add Baseline * Apply Belen 🗮 Company: Calculator								
	Result Time Croles Rea OPU SM Frequency CC Poscess								
Current	Carrent 1659-compute_sendencies_x,229,gpu (65555,1,1)v(128,1,1) 6.19 msecond 8,024031 64 0-Teals V10050M25208 1.30 cycle/insecond 7.0 [21121] operance								
- GPU Speed Of List		AI TO							
High generation of the Impact of the UML for each unit, the Employment the active of presting of the UML for each unit, the Employment the Active of the Act									
Compute (SM) Throu	Compute (34) Throughput [3] 70.32 Dutation [mescent]								
Memory Throughput									
L1/TEX Cache Throug	L1/TEX Cache Throughput (h) 19.62 SM Active Cycles (sycle)								
L2 Cache Throughout	t[b] 17.18 SM Frequency (cycle/intecond)								
DRAM Throughput [%	DRAM Throughput (%) 44.57 (DRAM Frequency (system)								
A High Company Throughout Company to Among Market Among Look at the Company particle according Among									
A revice base of the state of t									
A F94.32 Difference - The actives of pds performance in 1% issues than the fpds pipeline utilization. Check the "inclusion" citations' and function to ase if using function to ase if using functions can benefit this issues.									
GPU Throughput									
Compute (SM) Put									
Memory (b)									
c.									
Speed Of Light (SOL) [h]									

Rename a baseline by clicking the **Baseline** # text label.



Now, **open the new profile report** or switch to the other tab referencing this report. The baseline performance metrics will now be displayed and compared to the new current report's performance metrics.



1.15 Experiment with a Proposed Optimization - Replace Divide with Multiply

Noting the **hotspot at line 288**, we can assess if there's a way to re-formulate this line to either reduce redundant operations or refactor the overall algorithm. The metrics provided may be able to provide a hint towards why this line is a bottleneck for MiniWeather.

In this case, there are a significant number of warp stalls (see here for descriptions of types of warp stalls) as well as a much higher number of instructions executed compared to other lines in this kernel. Looking at this line, we see multiple divisions by 12 that could be simplified. Additionally, division is typically more expensive than multiplication within IEEE computational arithmetic.



1.16 EXERCISE - Adjust MiniWeather Problem Size and Other Optimizations

Adjust MiniWeather's problem size using the values nx=128,512,1024,2048,4096,9192 with nz=nx/2. Try more problem sizes if interested. Generate ncu reports for each of these problem sizes.

Then, open up all the reports and add each one as a named baseline for that problem size. Compare performance between problem sizes.

- 1. Describe the performance for small problem sizes? What is the SM utilization and memory throughput for small problems?
- 2. Is there an optimal problem size?
- 3. Do performance or other metrics stop changing after a certain order of magnitude for the problem size?
- 4. Experiment with and attempt other optimizations/code changes to improve MiniWeather's performance. What other ways or styles of refactoring might you try to improve performance?

1.17 Resources

- Nsight Systems Main Documentation
- Nsight Compute Main Documentation
- Nsight Compute Profiling Guide
- Nsight Compute Training Resources Forum, Videos, and Blog Posts curated by NVIDIA
- Introduction to Kernel Performance Analysis with NVIDIA Nsight Compute, Max Katz (NVIDIA invited to Argonne/NERSC)
 - GitLab repo and video

1.18 Return to Nsight Systems Profiler Tool

Nsight Systems Profiler