Analysis Report

matrixMultDevice(float*, float*, float*, int)

Duration	93.339 ms (93,339,293 ns)
Grid Size	[63,63,1]
Block Size	[32,32,1]
Registers/Thread	25
Shared Memory/Block	0 B
Shared Memory Requested	96 KiB
Shared Memory Executed	96 KiB
Shared Memory Bank Size	4 B

[0] GeForce GTX 980

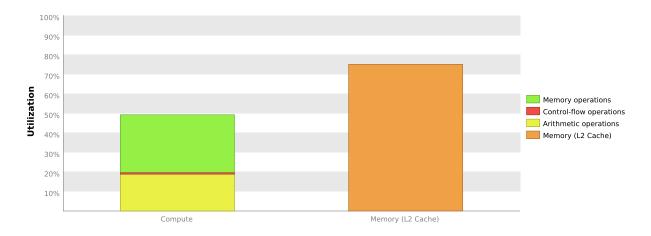
[U] GeForce G1X 980						
GPU UUID	GPU-76f08495-5b58-98e5-fb8d-af6a8ca423be					
Compute Capability	5.2					
Max. Threads per Block	1024					
Max. Threads per Multiprocessor	2048					
Max. Shared Memory per Block	48 KiB					
Max. Shared Memory per Multiprocessor	96 KiB					
Max. Registers per Block	65536					
Max. Registers per Multiprocessor	65536					
Max. Grid Dimensions	[2147483647, 65535, 65535]					
Max. Block Dimensions	[1024, 1024, 64]					
Max. Warps per Multiprocessor	64					
Max. Blocks per Multiprocessor	32					
Single Precision FLOP/s	5.186 TeraFLOP/s					
Double Precision FLOP/s	162.048 GigaFLOP/s					
Number of Multiprocessors	16					
Multiprocessor Clock Rate	1.266 GHz					
Concurrent Kernel	true					
Max IPC	6					
Threads per Warp	32					
Global Memory Bandwidth	224.32 GB/s					
Global Memory Size	3.94 GiB					
Constant Memory Size	64 KiB					
L2 Cache Size	2 MiB					
Memcpy Engines	2					
PCIe Generation	3					
PCIe Link Rate	8 Gbit/s					
PCIe Link Width	16					

1. Compute, Bandwidth, or Latency Bound

The first step in analyzing an individual kernel is to determine if the performance of the kernel is bounded by computation, memory bandwidth, or instruction/memory latency. The results below indicate that the performance of kernel "matrixMultDevice" is most likely limited by memory bandwidth. You should first examine the information in the "Memory Bandwidth" section to determine how it is limiting performance.

1.1. Kernel Performance Is Bound By Memory Bandwidth

For device "GeForce GTX 980" the kernel's compute utilization is significantly lower than its memory utilization. These utilization levels indicate that the performance of the kernel is most likely being limited by the memory system. For this kernel the limiting factor in the memory system is the bandwidth of the L2 Cache memory.



2. Memory Bandwidth

Memory bandwidth limits the performance of a kernel when one or more memories in the GPU cannot provide data at the rate requested by the kernel. The results below indicate that the kernel is limited by the bandwidth available to the L2 cache.

2.1. GPU Utilization Is Limited By Memory Bandwidth

The following table shows the memory bandwidth used by this kernel for the various types of memory on the device. The table also shows the utilization of each memory type relative to the maximum throughput supported by the memory. The results show that the kernel's performance is potentially limited by the bandwidth available from one or more of the memories on the device.

Optimization: Try the following optimizations for the memory with high bandwidth utilization.

Shared Memory - If possible use 64-bit accesses to shared memory and 8-byte bank mode to achieved 2x throughput.

L2 Cache - Align and block kernel data to maximize L2 cache efficiency.

Unified Cache - Reallocate texture data to shared or global memory. Resolve alignment and access pattern issues for global loads and stores.

Device Memory - Resolve alignment and access pattern issues for global loads and stores.

System Memory (via PCIe) - Make sure performance critical data is placed in device or shared memory.

Transactions	Bandwidth	Utilization					
Shared Memory							
Shared Loads	0	0 B/s					
Shared Stores	0	0 B/s					
Shared Total	0	0 B/s	Idle	Low	Medium	High	Max
L2 Cache	ı		luic	LOW	Mediam	riigii	MUX
Reads	1254347867	473.402 GB/s					
Writes	500006	188.707 MB/s					
Total	1254847873	473.591 GB/s	Idle	Low	Medium	High	Max
Unified Cache	Į.		IGIC	LOW	ricalani	Tilgii	HUX
Local Loads	0	0 B/s					
Local Stores	0	0 B/s					
Global Loads	3000000000	472.516 GB/s					
Global Stores	500000	188.704 MB/s					
Texture Reads	2000000000	754.818 GB/s					
Unified Total	5000500000	1,227.523 GB/s	Idle	Low	Medium	High	Max
Device Memory	!		1010	2011	ricalani	111911	TIGA
Reads	35234494	13.298 GB/s					
Writes	527588	199.116 MB/s					
Total	35762082	13.497 GB/s	Idle	Low	Medium	High	Max
System Memory	ı		laie	LOVV	Medium	Tilgii	IMAX
[PCle configuration: Gen3 x16	6, 8 Gbit/s]						
Reads	0	0 B/s	Idle	Low	Medium	High	Max
Writes	5 1.887 kB/s			LO VV	Picalum -	111911	
WITECS	3	1.007 KD/S	Idle	Low	Medium	High	Max

3. Instruction and Memory Latency

Instruction and memory latency limit the performance of a kernel when the GPU does not have enough work to keep busy. The performance of latency-limited kernels can often be improved by increasing occupancy. Occupancy is a measure of how many warps the kernel has active on the GPU, relative to the maximum number of warps supported by the GPU. Theoretical occupancy provides an upper bound while achieved occupancy indicates the kernel's actual occupancy.

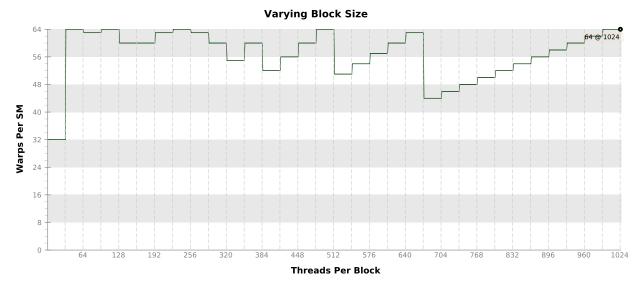
3.1. Occupancy Is Not Limiting Kernel Performance

The kernel's block size, register usage, and shared memory usage allow it to fully utilize all warps on the GPU.

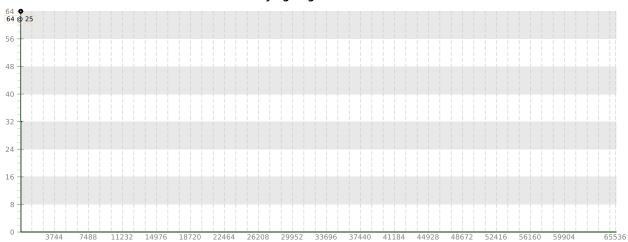
Variable	Achieved	Theoretical	Device Limit	Grid Size: [63,63,1] (3969 blocks) Block Size: [32,32,1] (1024
Occupancy Per SM				
Active Blocks		2	32	0 3 6 9 12 15 18 21 24 27 30 32
Active Warps	62.02	64	64	0 7 14 21 28 35 42 49 56 664
Active Threads		2048	2048	0 256 512 768 1024 1280 1536 1792 204
Occupancy	96.9%	100%	100%	0% 25% 50% 75% 100
Warps				
Threads/Block		1024	1024	0 128 256 384 512 640 768 896 102
Warps/Block		32	32	0 3 6 9 12 15 18 21 24 27 30 32
Block Limit		2	32	0 3 6 9 12 15 18 21 24 27 30 32
Registers				
Registers/Thread		25	65536	0 8192 16384 24576 32768 40960 49152 57344 6553
Registers/Block		32768	65536	0 16k 32k 48k 64
Block Limit		2	32	0 3 6 9 12 15 18 21 24 27 30 32
Shared Memory				
Shared Memory/Block		0	98304	0 32k 64k 96
Block Limit			32	

3.2. Occupancy Charts

The following charts show how varying different components of the kernel will impact theoretical occupancy.

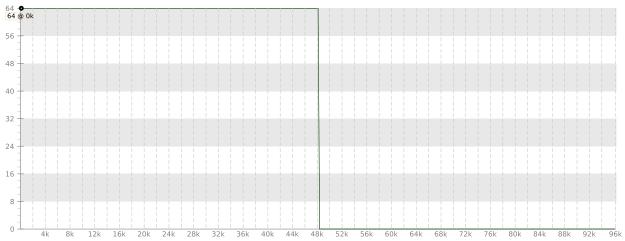






Registers Per Thread

Varying Shared Memory Usage



4. Compute Resources

GPU compute resources limit the performance of a kernel when those resources are insufficient or poorly utilized.

4.1. Function Unit Utilization

Different types of instructions are executed on different function units within each SM. Performance can be limited if a function unit is over-used by the instructions executed by the kernel. The following results show that the kernel's performance is not limited by overuse of any function unit.

Load/Store - Load and store instructions for shared and constant memory.

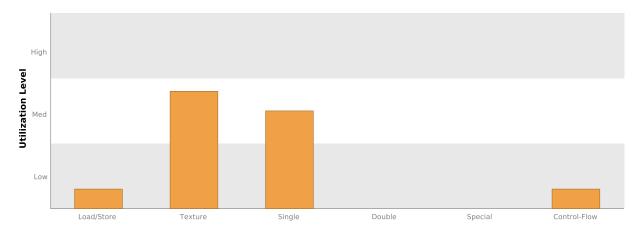
Texture - Load and store instructions for local, global, and texture memory.

Single - Single-precision integer and floating-point arithmetic instructions.

Double - Double-precision floating-point arithmetic instructions.

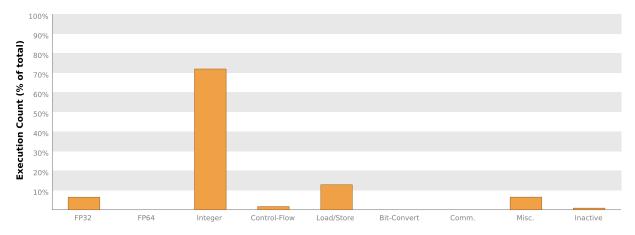
Special - Special arithmetic instructions such as sin, cos, popc, etc.

Control-Flow - Direct and indirect branches, jumps, and calls.



4.2. Instruction Execution Counts

The following chart shows the mix of instructions executed by the kernel. The instructions are grouped into classes and for each class the chart shows the percentage of thread execution cycles that were devoted to executing instructions in that class. The "Inactive" result shows the thread executions that did not execute any instruction because the thread was predicated or inactive due to divergence.



4.3. Floating-Point Operation Counts

The following chart shows the mix of floating-point operations executed by the kernel. The operations are grouped into classes and for each class the chart shows the percentage of thread execution cycles that were devoted to executing operations in that class. The results do not sum to 100% because non-floating-point operations executed by the kernel are not shown in this chart.

