# **Analysis Report**

## gpu5\_kernel(Matrix, Matrix, Matrix)

Duration	211.38781 ms (211,387,808 ns)
Grid Size	[ 128,128,1 ]
Block Size	[ 16,16,1 ]
Registers/Thread	32
Shared Memory/Block	4 KiB
Shared Memory Requested	96 KiB
Shared Memory Executed	96 KiB
Shared Memory Bank Size	4 B

## [0] Tesla V100-PCIE-16GB

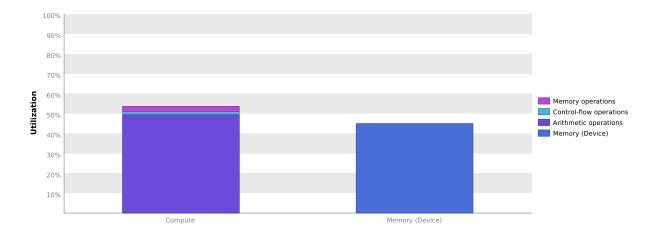
GPU UUID	GPU-94112f20-3db3-b803-e2fe-85e4cdc16a13
Compute Capability	7.0
Max. Threads per Block	1024
Max. Threads per Multiprocessor	2048
Max. Shared Memory per Block	48 KiB
Max. Shared Memory per Multiprocessor	96 KiB
Max. Registers per Block	65536
Max. Registers per Multiprocessor	65536
Max. Grid Dimensions	[ 2147483647, 65535, 65535 ]
Max. Block Dimensions	[ 1024, 1024, 64 ]
Max. Warps per Multiprocessor	64
Max. Blocks per Multiprocessor	32
Half Precision FLOP/s	28.262 TeraFLOP/s
Single Precision FLOP/s	14.131 TeraFLOP/s
Double Precision FLOP/s	7.066 TeraFLOP/s
Number of Multiprocessors	80
Multiprocessor Clock Rate	1.38 GHz
Concurrent Kernel	true
Max IPC	4
Threads per Warp	32
Global Memory Bandwidth	898.048 GB/s
Global Memory Size	15.774 GiB
Constant Memory Size	64 KiB
L2 Cache Size	6 MiB
Memcpy Engines	7
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 "gpu5\_kernel" is most likely limited by instruction and memory latency. You should first examine the information in the "Instruction And Memory Latency" section to determine how it is limiting performance.

#### 1.1. Kernel Performance Is Bound By Instruction And Memory Latency

This kernel exhibits low compute throughput and memory bandwidth utilization relative to the peak performance of "Tesla V100-PCIE-16GB". These utilization levels indicate that the performance of the kernel is most likely limited by the latency of arithmetic or memory operations. Achieved compute throughput and/or memory bandwidth below 60% of peak typically indicates latency issues.



## 2. 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.

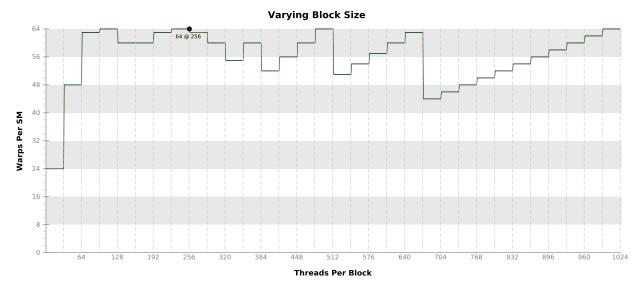
#### 2.1. Occupancy Is Not Limiting Kernel Performance

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

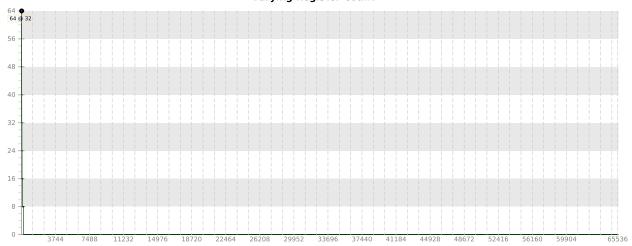
Variable	Achieved	Theoretical	Device Limit	Grid Size	: [ 128,12	8,1]	(16384	bloc	ks) Blo	ck Siz	ze: [ 16
Occupancy Per SM											
Active Blocks		8	32	0 4	4 8	12	16	20	24	28	32
Active Warps	63.36	64	64	0	9 18	2	7 36	ō	45	54	664
Active Threads		2048	2048	0	512		1024		1536		2048
Occupancy	99%	100%	100%	0%	25%		50%		75%	, )	100%
Warps											
Threads/Block		256	1024	0	256		512		768		1024
Warps/Block		8	32	0 4	4 8	12	16	20	24	28	32
Block Limit		8	32	0 4	4 8	12	16	20	24	28	32
Registers											
Registers/Thread		32	65536	0	16384		32768		49152	2	65536
Registers/Block		8192	65536	0	16k		32k		48k		64k
Block Limit		8	32	0 4	4 8	12	16	20	24	28	32
Shared Memory											
Shared Memory/Block		4096	98304	0	32k 64k					96k	
Block Limit		24	32	0 4	4 8	12	16	20	24	28	32

#### 2.2. Occupancy Charts

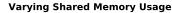
The following charts show how varying different components of the kernel will impact theoretical occupancy.

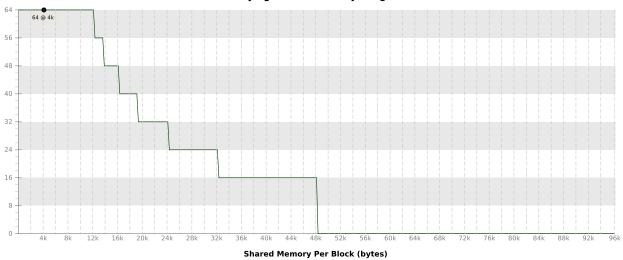


#### **Varying Register Count**



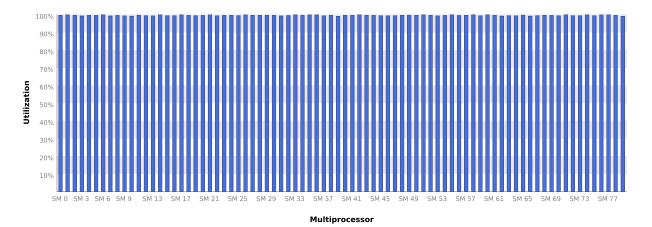
Registers Per Thread





#### 2.3. Multiprocessor Utilization

The kernel's blocks are distributed across the GPU's multiprocessors for execution. Depending on the number of blocks and the execution duration of each block some multiprocessors may be more highly utilized than others during execution of the kernel. The following chart shows the utilization of each multiprocessor during execution of the kernel.



5

## 3. Compute Resources

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

#### 3.1. 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.

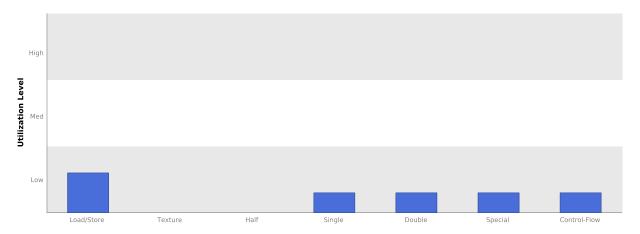
Half - Half-precision floating-point arithmetic instructions.

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

Double - Double-precision floating-point arithmetic instructions.

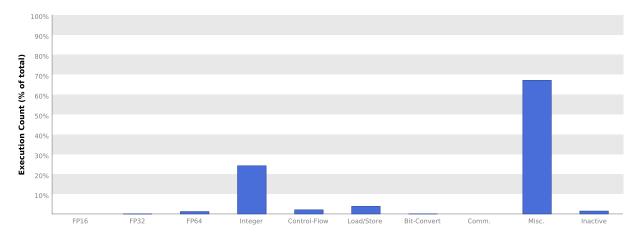
Special - Special arithmetic instructions such as sin, cos, popc, etc.

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



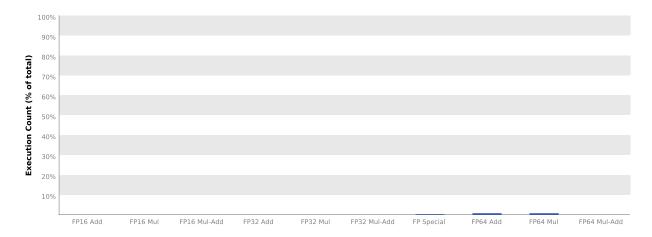
#### 3.2. Instruction Execution Counts

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.



#### 3.3. 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.



## 4. 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.

## 4.1. Memory Bandwidth And Utilization

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.

