# **Analysis Report**

## run\_lbm(double[27]\*, double[27]\*, unsigned long, double, IndexBlock)

Duration	5.90161 ms (5,901,607 ns)
Grid Size	[ 30,1,1 ]
Block Size	[ 30,30,1 ]
Registers/Thread	60
Shared Memory/Block	0 B
Shared Memory Requested	48 KiB
Shared Memory Executed	48 KiB
Shared Memory Bank Size	4 B

## [0] GeForce GTX 765M

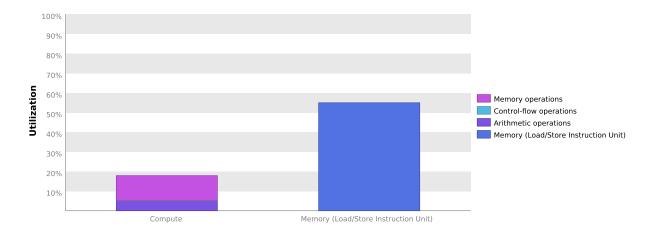
[0] GeForce GTX 765M				
GPU UUID	GPU-fe6c21ea-b7dc-9c6c-991c-33e89452d625			
Compute Capability	3.0			
Max. Threads per Block	1024			
Max. Threads per Multiprocessor	2048			
Max. Shared Memory per Block	48 KiB			
Max. Shared Memory per Multiprocessor	48 KiB			
Max. Registers per Block	65536			
Max. 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	16			
Single Precision FLOP/s	1.325 TeraFLOP/s			
Double Precision FLOP/s	55.2 GigaFLOP/s			
Number of Multiprocessors	4			
Multiprocessor Clock Rate	862.5 MHz			
Concurrent Kernel	true			
Max IPC	7			
Threads per Warp	32			
Global Memory Bandwidth	64.128 GB/s			
Global Memory Size	1.952 GiB			
Constant Memory Size	64 KiB			
L2 Cache Size	256 KiB			
Memcpy Engines	1			
PCIe Generation	2			
PCIe Link Rate	5 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 "run\_lbm" 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 "GeForce GTX 765M". 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. The results below indicate that occupancy can be improved by reducing the number of registers used by the kernel.

#### 2.1. GPU Utilization Is Limited By Register Usage

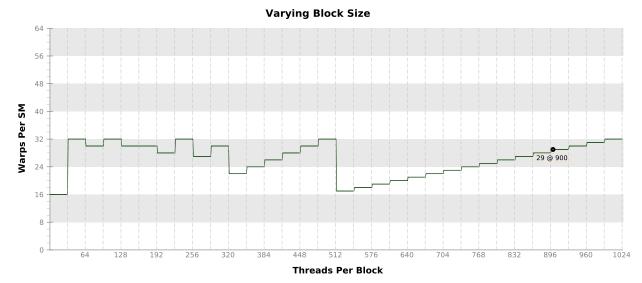
The kernel uses 60 registers for each thread (54000 registers for each block). This register usage is likely preventing the kernel from fully utilizing the GPU. Device "GeForce GTX 765M" provides up to 65536 registers for each block. Because the kernel uses 54000 registers for each block each SM is limited to simultaneously executing 1 block (29 warps). Chart "Varying Register Count" below shows how changing register usage will change the number of blocks that can execute on each SM.

Optimization: Use the -maxrregcount flag or the \_\_launch\_bounds\_\_ qualifier to decrease the number of registers used by each thread. This will increase the number of blocks that can execute on each SM. On devices with Compute Capability 5.2 turning global cache off can increase the occupancy limited by register usage.

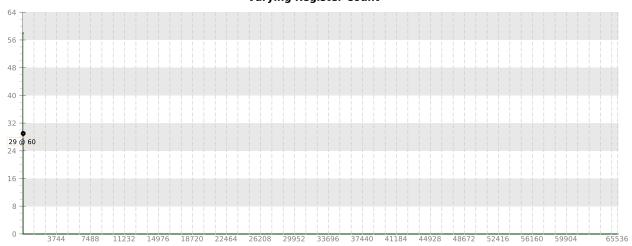
Variable	Achieved	Theoretical	Device Limit	Grid Size: [ 30,1,1 ] (30 blocks) Block Size: [ 30,30,1 ] (900 thread
Occupancy Per SM				
Active Blocks		1	16	0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
Active Warps	27.38	29	64	0 7 14 21 28 35 42 49 56 664
Active Threads		928	2048	0 256 512 768 1024 1280 1536 1792 2048
Occupancy	42.8%	45.3%	100%	0% 25% 50% 75% 100%
Warps				
Threads/Block		900	1024	0 128 256 384 512 640 768 896 1024
Warps/Block		29	32	0 3 6 9 12 15 18 21 24 27 30 32
Block Limit		2	16	0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
Registers				
Registers/Thread		60	65536	0 8192 16384 24576 32768 40960 49152 57344 65536
Registers/Block		59392	65536	0 16k 32k 48k 64k
Block Limit		1	16	0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
Shared Memory				
Shared Memory/Block		0	49152	0 16k 32k 48k
Block Limit		0	16	0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16

#### 2.2. Occupancy Charts

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

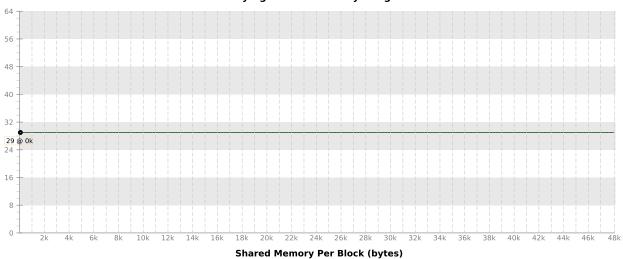


#### **Varying Register Count**



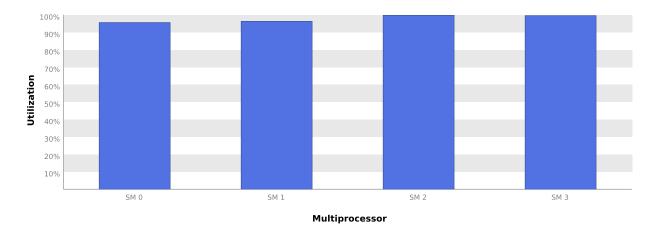
Registers Per Thread





#### 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.



### 3. Compute Resources

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

#### 3.1. Kernel Profile - Instruction Execution

The Kernel Profile - Instruction Execution shows the execution count, inactive threads, and predicated threads for each source and assembly line of the kernel. Using this information you can pinpoint portions of your kernel that are making inefficient use of compute resource due to divergence and predication.

Examine portions of the kernel that have high execution counts and inactive or predicated threads to identify optimization opportunities.

#### Cuda Fuctions :

run\_lbm(double[27]\*, double[27]\*, unsigned long, double, IndexBlock)

Maximum instruction execution count in assembly: 7830

Average instruction execution count in assembly: 3007

Instructions executed for the kernel: 2947560

Thread instructions executed for the kernel: 91476000

Non-predicated thread instructions executed for the kernel: 84960900

Warp non-predicated execution efficiency of the kernel: 90.1%

Warp execution efficiency of the kernel: 97.0%

Source files:

/home/egoh6/final/parallel/lbm\_cuda.cu

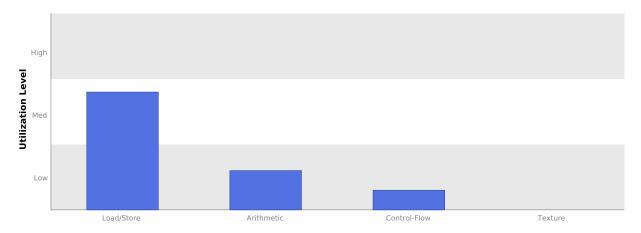
#### 3.2. Function Unit Utilization

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 local, shared, global, constant, etc. memory.

Arithmetic - All arithmetic instructions including integer and floating-point add and multiply, logical and binary operations, etc. Control-Flow - Direct and indirect branches, jumps, and calls.

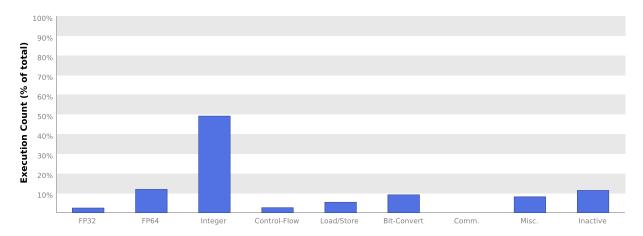
Texture - Texture operations.



#### 3.3. Instruction Execution Counts

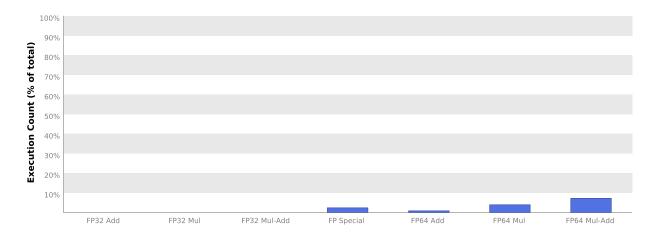
The following chart shows the mix of instructions executed by the kernel. The instructions are grouped into classes and for each class the chart shows the percentage of thread execution cycles that were devoted to executing instructions in that class. The

"Inactive" result shows the thread executions that did not execute any instruction because the thread was predicated or inactive due to divergence.



## 3.4. 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.

