

CUDA Thread Model

~ Threading and Scheduling ~

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Objectives

In this talk I will cover:

- 1. Differences between CPU and GPU threads.
- 2. CUDA threads and indexes.
- 3. CUDA kernel launch & thread scheduling.
- 4. CUDA thread synchronization.

Terminology

New terminology introduced in this talk:

- 1. A warp is a group of 32 thread. (Can anyone tell me why 32 threads?)
- 2. A block is a 1, 2 or 3-dimensional array of threads.
- 3. A grid is a 1 or 2-dimensional array of blocks.
- 4. dim2, dim3 are (x,y) and (x,y,z) data types, respectively.



~ 1. CPU versus GPU ~

GPU threads are different from CPU threads:

- CPUs typically have 16 or less threads in-flight.
- GPUs have tens of thousands of threads in-flight.
- GPU threads are "lightweight" whereas CPU threads are "heavyweight". CPU context switching is expensive. GPU context switching is very lightweight and fast.



~ 2. Threads and Indexes ~

CUDA threads form a hierarchy:

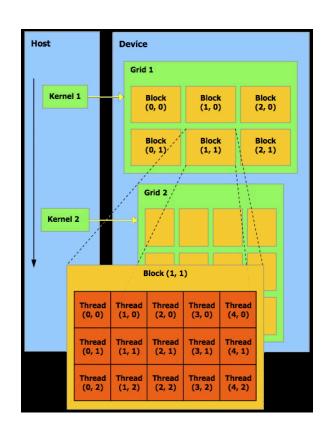
- Threads are grouped into blocks.
- Blocks are grouped into a grid.
- Each thread has a unique index within it's block.
- Each block has a unique index within it's grid.
- Note, there are many threads per block, and many blocks in a grid, but there is only one grid in a kernel.



Fountainhead

~ Threads and Indexes (cont.) ~

CUDA Thread Hierarchy





Threads in a single block:

- Run on a single multiprocessor.
- Share data held in shared memory.
- A warp will always be a subset of threads in a block.
- Threads in different blocks may be assigned to different multiprocessors concurrently, to the same multiprocessor concurrently (multithreading), or a mix at different times.



Threads in a single block (cont.):

- Hard limit on a thread block: 512 threads (16 warps) on Tesla, 1024 threads (32 warps) on Fermi.
- Thread blocks always created in warp units, so don't bother defining a block that is not a multiple of 32 threads (number of threads in a warp).
- Thread blocks within a grid have the same size & shape.



Threads:

- Tesla supports 32 active warps per multiprocessor. Fermi supports 48.
- Memory latency is hidden by swapping warps in and out, mixing computation with memory operations.



Threads running on a single multiprocessor:

- A Fermi GPU can have up to 1024 threads in a block, equivalent to 32 warps.
- However, the 1024 threads (32 warps) can be configured in a number of ways (# blocks, #warps): (2, 16), (3, 10), (4,8) ... (8,4). Note: Max. of 8 blocks per multiprocessor.
- Fermi → 48 active warps → 1536 threads in-flight.

Indexes are exposed through built-in variables:

- dim3 threadIdx identifies a thread in a block.
- dim3 blockIdx identifies a block in a grid.
- dim3 blockDim block dimension.
- dim3 gridDim grid dimension.
- int warpSize warp size in threads.

Grid dimensions:

gridDim is specified at kernel launch.

Block indexes:

```
0 <= blockIdx.x <= gridDim.x - 1
0 <= blockIdx.y <= gridDim.y - 1
    blockIdx.z = 0</pre>
```

Thread indexes:

```
0 <= threadIdx.x <= blockDim.x - 1
0 <= threadIdx.y <= blockDim.y - 1
0 <= threadIdx.z <= blockDim.z - 1</pre>
```

blockDim is specified at kernel launch. Hard limit on total threads in a block: 512 for Tesla, 1024 for Fermi.

Some recurring access patterns:

~ 3. Kernel Launch ~

Kernel launch:

```
dim3 dimBlock(4, 2, 2);
dim3 dimGrid(2, 1, 1);
Kernel<<<dimGrid, dimBlock>>>( ... );
```

~ Kernel Launch (cont.) ~

Scheduling unit is the warp:

- Zero overhead warp scheduling on multiprocessors.
- All threads in a warp execute the same instruction.
- A warp whose next instruction has its data ready is ready for execution.
- Warps eligible to execute are scheduled on a prioritized basis.



~ Kernel Launch (cont.) ~

Scheduling unit is the warp (cont.):

 4 clock cycles needed to dispatch the same instructions for all threads in a block.



~ 4. Thread Synchronization ~

Thread synchronization occurs at different levels:

- Threads within a block using shared memory.
- Threads in different blocks and different grids must use global memory.
- Implicit __syncthreads() (barrier) between kernels.
- Explicit __synchthreads() between threads in the same block. (All threads in a block must reach sync.)



~ Thread Synchronization (cont.) ~

Race conditions:

Race conditions arise when 2+ threads attempt to access the same memory location concurrently and at least one access is a write.

A word of advice:

It's more efficient to design programs that require as little synchronization between threads as possible.



~ Thread Synchronization (cont.) ~

Atomics:

- Atomic memory operations enforce atomic access to shared variables that can be accessed by multiple threads.
- You can synthesize various coordination objects and synchronization methods using atomics.