PUMPS Summer School

Programming and tUning Massively Parallel Systems

Lecture 3: CUDA Threads

Block IDs and Thread IDs



Matrix Multiplication Using **Multiple Blocks**

- Break-up Pd into tiles
- Each block calculates one • tile
 - Each thread calculates one element

tv

- Block size equal tile size

0

2



bx

0

2

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by 1

A Small Example



A Small Example: Multiplication



Revised Matrix Multiplication Kernel using Multiple Blocks

```
__global__ void MatrixMulKernel(float* Md, float* Nd, float* Pd, int Width)
{
    // Calculate the row index of the Pd element and M
    int Row = blockIdx.y*TILE_WIDTH + threadIdx.y;
    // Calculate the column idenx of Pd and N
    int Col = blockIdx.x*TILE_WIDTH + threadIdx.x;
```

```
float Pvalue = 0;
// each thread computes one element of the block sub-matrix
for (int k = 0; k < Width; ++k)
    Pvalue += Md[Row*Width+k] * Nd[k*Width+Col];
```

```
Pd[Row*Width+Col] = Pvalue;
```

Revised Step 5: Kernel Invocation (Host-side Code)

// Setup the execution configuration dim3 dimGrid(Width/TILE_WIDTH, Width/TILE_WIDTH); dim3 dimBlock(TILE_WIDTH, TILE_WIDTH);

// Launch the device computation threads!
MatrixMulKernel<<<dimGrid, dimBlock>>>(Md, Nd, Pd, Width);

CUDA Thread Block

- All threads in a block execute the same kernel program (SPMD)
- Programmer declares block:
 - Block size 1 to 512 concurrent threads
 - Block shape 1D, 2D, or 3D
 - Block dimensions in threads
- Threads have thread id numbers within block
 - Thread program uses thread id to select work and address shared data
- Threads in the same block share data and synchronize while doing their share of the work
- Threads in different blocks cannot cooperate
 - Each block can execute in any order relative to other blocks!

in threads

CUDA Thread Block



Courtesy: John Nickolls, NVIDIA

Transparent Scalability

- Hardware is free to assigns blocks to any processor at any time
 - A kernel scales across any number of parallel processors



Example: Executing Thread Blocks

t0 t1 t2 ... tm

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SM 0 SM 1

SP

MTIU

SP

t0 t1 t2 ... tm

Blocks

Threads are assigned to Streaming Multiprocessors in block granularity

- Up to 8 (?) blocks to each SM as resource allows
- Fermi SM can take up to **1536** threads
 - Could be 256 (threads/block) * 6 blocks
 - Or 512 (threads/block) * 3 blocks, etc.

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• Threads run concurrently

Blocks

- SM maintains thread/block id #s
 - SM manages/schedules thread execution

Example: Thread Scheduling

- Each Block is executed as 32-thread Warps
 - An implementation decision, not part of the CUDA programming model
 - Warps are scheduling units in SM
- If 3 blocks are assigned to an SM and each block has 256 threads, how many Warps are there in an SM?
 - Each Block is divided into 256/32 = 8 Warps
 - There are 8 * 3 = 24 Warps





Example: Thread Scheduling (Cont.)

- SM implements zero-overhead warp scheduling
 - At any time, 1 or 2 of the warps is executed by SM
 - Warps whose next instruction has its operands ready for consumption are eligible for execution
 - Eligible Warps are selected for execution on a prioritized scheduling policy
 - All threads in a warp execute the same instruction when selected



Block Granularity Considerations

- For Matrix Multiplication using multiple blocks, should I use 8X8, 16X16 or 32X32 blocks?
 - For 8X8, we have 64 threads per Block. Since each SM can take up to 1536 threads, there are 24 Blocks. However, each SM can only take up to 8 Blocks, only 512 threads will go into each SM!
 - For 16X16, we have 256 threads per Block. Since each SM can take up to 1536 threads, it can take up to 6 Blocks and achieve full capacity unless other resource considerations overrule.
 - For 32X32, we would have 1024 threads per Block. Each block can only have up to 512 threads! This configuration will not work.

Some Additional API Features

Application Programming Interface

- The API is an extension to the C programming language
- It consists of:
 - Language extensions
 - To target portions of the code for execution on the device
 - A runtime library split into:
 - A common component providing built-in vector types and a subset of the C runtime library in both host and device codes
 - A host component to control and access one or more devices from the host
 - A device component providing device-specific functions

Language Extensions: Built-in Variables

- dim3 gridDim;
 - Dimensions of the grid in blocks (gridDim.z unused)
- dim3 blockDim;
 - Dimensions of the block in threads
- dim3 blockIdx;
 - Block index within the grid
- dim3 threadIdx;
 - Thread index within the block

Common Runtime Component: Mathematical Functions

- pow, sqrt, cbrt, hypot
- exp, exp2, expm1
- log, log2, log10, log1p
- sin, cos, tan, asin, acos, atan, atan2
- sinh, cosh, tanh, asinh, acosh, atanh
- ceil, floor, trunc, round
- Etc.
 - When executed on the host, a given function uses the C runtime implementation if available
 - These functions are only supported for scalar types, not vector types

Device Runtime Component: Mathematical Functions

 Some mathematical functions (e.g. sin(x)) have a less accurate, but faster device-only version (e.g. __sin(x))



Host Runtime Component

- Provides functions to deal with:
 - Device management (including multi-device systems)
 - Memory management
 - Error handling
- Initializes the first time a runtime function is called
- A host thread can invoke device code on only one device
 - Multiple host threads required to run on multiple devices

Device Runtime Component: Synchronization Function

- void __syncthreads();
- Synchronizes all threads in a block
- Once all threads have reached this point, execution resumes normally
- Used to avoid RAW / WAR / WAW hazards when accessing shared or global memory
- Allowed in conditional constructs only if the conditional is uniform across the entire thread block