CS 220: Introduction to Parallel Computing **CUDA**

Lecture 29

- GPUs vs CPUs
- More Terminology
- Getting Started Lab
- Last 15 minutes:

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GPUs vs CPUs

- We discussed the differences between GPUs and CPUs last class
 - CPU = single paintball gun
 - GPU = lots of paintball guns?!
- To be more fair: the CPU offers us a lot more flexibility, but fewer cores
- Naturally, these differences lead to changes in how we think about implementing our programs

GPUs vs CPUs

- So we need to think about our problems in a different way...
- MPI is similar: many times we might think "we need to do this in a loop"
 - ...but we actually don't always need loops, since all the processes will execute the same thing

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The Kernel

- When your CUDA program starts, it executes the main() function on the **host**
 - Runs on the CPU and uses main memory
- The host is responsible for launching the kernel
 - Runs on the GPU and uses GPU memory
- Kernels are launched with this syntax: kernel_name<<<grid_size, block_size>>>(params); "Triple chevron" syntax <<<>>>

Grids and Blocks

- A grid is a collection of blocks
- A block is a collection of threads
- Total number of threads? grid_size * block_size
- Note: the blocks can also be 2D and 3D: <<<grid_sz, X, Y, Z>>>

Why it Matters

- Grids, blocks, and threads closely follow the hardware and logical design of GPUs
 - Each block gets assigned to a streaming multiprocessor (SM)
 - Memory availability/speed is impacted by thread organization
- By configuring these parameters correctly, we can get better performance
- For instance...

Performance Considerations

- Each block gets assigned to an SM
- The SMs split their blocks into warps
 - CUDA unit of SIMD execution
 - A warp = 32 threads
- If the number of threads in the block isn't evenly divisible by 32, then we'll have inactive threads:
 - 20 threads? 12 are inactive

Hello World

 Let's take a look at doing our usual "hello world" (CUDA style)

Math

- So, CUDA is not great for printing strings!
- More recent hardware and versions of the SDK do support printing directly from the kernel
- cuPrintf is another option
- Let's try something that GPUs are better at: math

Argument Passing

- Arguments to the CUDA kernel are passed by value (copied)
- To return anything back to the host program, we have to use:
 - cudaMalloc cudaMemcpy And our old friend, pointers!

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nvcc

- MPI and pthreads are both libraries built on C
- CUDA requires its own compiler
 - We're not writing standard C/C++ code
- A CUDA program ends with the extension .cu
- Compiling:

nvcc my_program.cu -o my_program

To use nvcc, we need to configure our systems first

Setting up the Environment

- Every process inherits its environment from its parent process
- The environment contains several environment variables that contain configuration information
- Want to see your environment variables? Run 'env' at your shell prompt
- To use nvcc, we need to set up one variable in particular: PATH

PATH

- To check your path, run: echo \$PATH
- If /usr/local/cuda/bin isn't in the path, you can add it with:

export PATH=/usr/local/cuda/bin:\$PATH

- (You can also add this line to your ~/.bash_profile)
- If your path is configured properly, you can run 'nvcc' and will be prompted to provide input files

LD_LIBRARY_PATH

- If you're having problems with CUDA functions not being found, you may also need to modify your LD_LIBRARY_PATH:
- export LD_LIBRARY_PATH=/usr/local/cuda/lib64: \$LD_LIBRARY_PATH

Lab: CUDA

- This lab has three parts:
 - In the first part, you'll compile Nvidia's device query program
 - This lets us find out warp sizes, thread info, etc.
 - Second part: vector addition using CUDA
 - Third part: matrix multiplication using CUDA

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