CS 179: GPU Computing

Lecture 2: The Basics

Recap

- Can use GPU to solve highly parallelizable problems
 - Performance benefits vs. CPU
- Straightforward extension to C language



Disclaimer

- Goal for Week 1:
 - Fast-paced introduction
 - "Know enough to be dangerous"
- We will fill in details later!

Our original problem...

- Add two arrays
 - A[] + B[] -> C[]
- Goal: Understand what's going on

CUDA code (first part)

```
void cudaAddVectors(const float* a, const float* b, float* c, size){
   //For now, suppose a and b were created before calling this function
   // dev a, dev b (for inputs) and dev c (for outputs) will be
   // arrays on the GPU.
    float * dev a;
    float * dev b;
    float * dev c;
    cudaMalloc((void **) &dev a, size*sizeof(float));
    cudaMemcpy(dev a, a, size*sizeof(float), cudaMemcpyHostToDevice);
    cudaMalloc((void **) &dev b, size*sizeof(float)); // and dev b
    cudaMemcpy(dev b, b, size*sizeof(float), cudaMemcpyHostToDevice);
   // Allocate memory on the GPu for our outputs:
    cudaMalloc((void **) &dev c, size*sizeof(float));
```

Basic "formula"

- Setup inputs on the host (CPU-accessible memory)
- Allocate memory for inputs on the GPU
- Copy inputs from host to GPU
- Allocate memory for outputs on the host
- Allocate memory for outputs on the GPU
- Start GPU kernel
- Copy output from GPU to host



"Classic" Memory Hierarchy

Memory Hierarchy



The GPU



The GPU





• Difference between CPU and GPU pointers?

• Difference between CPU and GPU pointers?

– None – pointers are just addresses!

- Difference between CPU and GPU pointers?
 - None pointers are just addresses!
 - Up to the programmer to keep track!

• Good practice:

Special naming conventions, e.g. "dev_" prefix





Memory allocational

• With the CPU (host memory)...

float *c = malloc(N * sizeof(float));

Attempts to allocate #bytes in argument

Memory allocation

• On the GPU (global memory):

```
float *dev_c;
cudaMalloc(&dev_c, N * sizeof(float));
```

• Signature:

cudaError_t cudaMalloc (void ** devPtr, size_t size)

- Attempts to allocate #bytes in arg2
- arg1 is the *pointer* to the pointer in GPU memory!
 - Passed into function for modification
 - Result after successful call: Memory allocated in location given by dev_c on GPU
- Return value is error code, can be checked





• With the CPU (host memory)...

// pointers source,destination to memory regions
memcpy(destination, source, N);

• Signature:

void * memcpy (void * destination, const void * source, size_t num);

 Copies *num* bytes from (area pointed to by) source to (area pointed to by) destination

- Versatile cudaMemcpy() equivalent
 - CPU -> GPU
 - GPU -> CPU
 - GPU -> GPU
 - CPU -> CPU

• Signature:

cudaError_t cudaMemcpy(void *destination, void *src, size_t count, enum cudaMemcpyKind kind)

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• Values:

- cudaMemcpyHostToHost
- cudaMemcpyHostToDevice
- cudaMemcpyDeviceToHost
- cudaMemcpyDeviceToDevice

• Signature:



- cudaMemcpyHostToHost
- cudaMemcpyHostToDevice
- cudaMemcpyDeviceToHost
- cudaMemcpyDeviceToDevice



Summary of memory

- CPU vs. GPU pointers
- cudaMalloc()
- cudaMemcpy()

```
//At lowest, should be 32
//Limit of 512 (Tesla), 1024 (newer)
const unsigned int threadsPerBlock = 512;
```

```
//How many blocks we'll end up needing
const unsigned int blocks = ceil(size/float(threadsPerBlock));
```

```
//Call the kernel!
cudaAddVectorsKernel<<<blocks, threadsPerBlock>>>
      (dev_a, dev_b, dev_c);
```

//Copy output from device to host (assume here that host memory //for the output has been calculated)

cudaMemcpy(c, dev_c, size*sizeof(float), cudaMemcpyDeviceToHost);

```
//Free GPU memory
cudaFree(dev_a);
cudaFree(dev_b);
cudaFree(dev_c);
```

}

Part 2

```
//At lowest, should be 32
//Limit of 512 (Tesla), 1024 (newer)
const unsigned int threadsPerBlock = 512;
//How many blocks we'll end up needing
const unsigned int blocks = ceil(size/float(threadsPerBlock));
//Call the kernel!
cudaAddVectorsKernel<<<blocks, threadsPerBlock>>>
        (dev_a, dev_b, dev_c);
//Copy output from device to host (assume here that host memory
//for the output has been calculated)
cudaMemcpy(c, dev_c, size*sizeof(float), cudaMemcpyDeviceToHost);
//Free GPU memory
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```

Recall...

- GPUs...
 - Have lots of cores
 - Are suited toward "parallel problems"









Device Multiprocessor N	
Multiprocessor 2	
Multiprocessor 1	
Shared Memory Registers Registers Re Processor 1 Processor 2 • Pro	gisters Decessor M Constant Cache
	Texture Cache
+ +	↓ IT
Device Memory	

One instruction unit for multiple cores!

Warps

- Groups of threads simultaneously execute same instructions!
 - Called a "warp"
 - (32 threads in a warp under current standards)



Blocks

- Group of threads scheduled to a multiprocessor
 - Contain multiple warps
 - Has a max. number (varies by GPU, e.g. 512 or 1024)

Multiprocessor execution timeline



Thread groups

- A grid (all the threads started...):
 - ...contains *blocks* <- assigned to multiprocessors</p>
 - Each *block* contains *warps* <- executed simultaneously
 - Each warp contains individual threads

Part 2

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cudaMemcpy(c, dev_c, size*sizeof(float), cudaMemcpyDeviceToHost);
//Free GPU memory
cudaFree(dev_a);
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```

- Moral 1: (from Lecture 1)
 - Start lots of threads!
 - Recall: Low context switch penalty
 - Hide latency
 - Start enough blocks!
 - Occupy SMs
 - e.g. Don't call:

kernel<<<1,1>>>(); // 1 block, 1 thread per block

– Call:

```
kernel<<<50,512>>>();// 50 blocks, 512 threads per block
```

- Moral 2:
 - Multiprocessors execute warps (of 32 threads)
 - Block sizes of 32*n (integer n) are best
 - e.g. Don't call:

kernel<<<50,97>>>(); // 50 blocks, 97 threads per block

– Call:

kernel<<<50,128>>>();// 50 blocks, 128 threads per block

Summary (processor internals)

- Key parameters on kernel call:
 - Threads per block
 - Number of blocks
- Choose carefully!

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```
//Call the kernel!
```

//Copy output from device to host (assume here that host memory //for the output has been calculated)

cudaMemcpy(c, dev_c, size*sizeof(float), cudaMemcpyDeviceToHost);

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```

}

Kernel argument passing

- Similar to arg-passing in C functions
- Some rules:
 - Don't pass host-memory pointers
 - Small variables (e.g. individual ints) are fine
 - No pass-by-reference

Kernel function

- Executed by *many* threads
- Threads have unique ID mechanism:
 - Thread index within block
 - Block index

```
__global__ void
cudaAddVectorsKernel(float * a, float * b, float * c) {
    unsigned int index = blockIdx.x * blockDim.x + threadIdx.x;
    c[index] = a[index] + b[index];
}
```

- Out of bounds issue:
 - If index > (#elements), illegal access!



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- #Threads issue:
 - Cannot start e.g. 1e9 threads!
 - Threads should handle arbitrary # of elements

```
__global___void
cudaAddVectorsKernel(float * a, float * b, float * c, int size) {
    unsigned int index = blockIdx.x * blockDim.x + threadIdx.x;
    if (index < size) {
        c[index] = a[index] + b[index];
    }
}
```

- #Threads issue:
 - Cannot start e.g. 1e9 threads!
 - Threads should handle arbitrary # of elements

global <i>void</i>
<pre>cudaAddVectorsKernel(float * a, float * b, float * c, int size) {</pre>
<pre>unsigned int index = blockIdx.x * blockDim.x + threadIdx.x;</pre>
<pre>while (index < size) ?</pre>
<pre>c[index] = a[index] + b[index];</pre>
<pre>index += blockDim.x * gridDim.x;</pre>
}
}

- #Threads issue:
 - Cannot start e.g. 1e9 threads!
 - Threads should handle arbitrary # of elements



Total number of blocks



Host memory pointer (copy *to* here)

Device memory pointer (copy *from* here)

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//Copy output from device to host (assume here that host memory
//for the output has been calculated)

cudaMemcpy(c, dev_c, size*sizeof(float), cudaMemcpyDeviceToHost);

//Free GPU memory
cudaFree(dev_a);
cudaFree(dev_b);
cudaFree(dev_c);

- cudaFree()
 - Equivalent to host memory's free() function
 - (As on host) Free memory after completion!

Summary

- GPU global memory:
 - Pointers (CPU vs GPU)
 - cudaMalloc() and cudaMemcpy()
- GPU processor details:
 - Thread group hierarchy
 - Launch parameters
- Threads in kernel