CUDA Programming

Sum of 2 Arrays

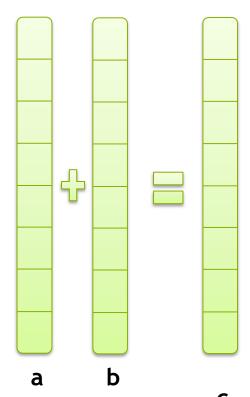
Outline

- **□**Addition on the device
 - ☐ Moving to parallel using blocks
 - ☐ Moving to parallel using threads
 - **□**Combining blocks and threads

Parallel Programming in CUDA C/C++

GPU computing is about massive parallelism!

 We'll start by adding two integers and build up to vector addition



Addition on the Device

A simple kernel to add two integers

```
__global__ void add(int *a, int *b, int *c) {
    *c = *a + *b;
}
```

- As before __global__ is a CUDA C/C++ keyword meaning
 - add() will execute on the device
 - add() will be called from the host

Addition on the Device

Note that we use pointers for the variables

```
__global__ void add(int *a, int *b, int *c) {
    *c = *a + *b;
}
```

- add() runs on the device, so a, b and c must point to device memory
- We need to allocate memory on the GPU

Memory Management

- Host and device memory are separate entities
 - Device pointers point to GPU memory
 May be passed to/from host code
 May not be dereferenced in host code
 - Host pointers point to CPU memory
 May be passed to/from device code
 May not be dereferenced in device code





- Simple CUDA API for handling device memory
 - cudaMalloc(), cudaFree(), cudaMemcpy()
 - Similar to the C equivalents malloc(), free(), memcpy()

Addition on the Device: add()

• Returning to our add() kernel

```
__global__ void add(int *a, int *b, int *c) {
    *c = *a + *b;
}
```

• Let's take a look at main()...

Addition on the Device: main()

```
int main(void) {
                     // host copies of a, b, c
      int a, b, c;
      int *d_a, *d_b, *d_c; // device copies of a, b, c
      int size = sizeof(int);
      // Allocate space for device copies of a, b, c
      cudaMalloc((void **)&d a, size);
      cudaMalloc((void **)&d b, size);
      cudaMalloc((void **)&d c, size);
      // Setup input values
      a = 2;
      b = 7;
```

Addition on the Device: main()

```
// Copy inputs to device
cudaMemcpy(d a, &a, size, cudaMemcpyHostToDevice);
cudaMemcpy(d b, &b, size, cudaMemcpyHostToDevice);
// Launch add() kernel on GPU
add<<<1,1>>>(d a, d b, d c);
// Copy result back to host
cudaMemcpy(&c, d c, size, cudaMemcpyDeviceToHost);
// Cleanup
cudaFree(d a); cudaFree(d b); cudaFree(d c);
return 0;
```

Moving to Parallel

- GPU computing is about massive parallelism
 - So how do we run code in parallel on the device?

```
add<<< 1, 1 >>>();

|
add<<< N, 1 >>>();
```

 Instead of executing add () once, execute N times in parallel

Vector Addition on the Device

- With add() running in parallel we can do vector addition
- Terminology: each parallel invocation of add() is referred to as a block
 - The set of blocks is referred to as a grid
 - Each invocation can refer to its block index using blockidx.x

```
__global__ void add(int *a, int *b, int *c) {
    c[blockIdx.x] = a[blockIdx.x] + b[blockIdx.x];
}
```

By using blockIdx.x to index into the array, each block handles
a different index

Vector Addition on the Device

```
__global__ void add(int *a, int *b, int *c) {
    c[blockIdx.x] = a[blockIdx.x] + b[blockIdx.x];
}
```

On the device, each block can execute in parallel:

Vector Addition on the Device: add()

Returning to our parallelized add() kernel

```
__global__ void add(int *a, int *b, int *c) {
    c[blockIdx.x] = a[blockIdx.x] + b[blockIdx.x];
}
```

Let's take a look at main()...

Vector Addition on the Device: main()

```
#define N 512
int main(void) {
   int *a *b *c
                          // host copies of a, b, c
   int *d a, *d b, *d c; // device copies of a, b, c
   int size = N * sizeof(int);
   // Alloc space for device copies of a, b, c
   cudaMalloc((void **)&d a, size);
   cudaMalloc((void **)&d b, size);
   cudaMalloc((void **)&d c, size);
   // Alloc space for host copies of a, b, c and setup input values
   a = (int *)malloc(size); random ints(a, N);
   b = (int *)malloc(size); random ints(b, N);
   c = (int *)malloc(size);
```

Vector Addition on the Device: main()

```
// Copy inputs to device
cudaMemcpy(d a, a, size, cudaMemcpyHostToDevice);
cudaMemcpy(d b, b, size, cudaMemcpyHostToDevice);
// Launch add() kernel on GPU with N blocks
add <<< N, 1>>> (d a, d b, d c);
// Copy result back to host
cudaMemcpy(c, d c, size, cudaMemcpyDeviceToHost);
// Cleanup
free(a); free(b); free(c);
cudaFree(d a); cudaFree(d b); cudaFree(d c);
return 0;
```

CUDA Threads

Terminology: a block can be split into parallel threads

Let's change add() to use parallel threads instead of parallel blocks

```
__global__ void add(int *a, int *b, int *c) {
    c[threadIdx.x] = a[threadIdx.x] + b[threadIdx.x];
}
```

- We use threadIdx.x instead of blockIdx.x
- Need to make one change in main()...

```
add<<< 1, 1 >>>();
add<<< 1, N >>>();
```

Vector Addition on the Device

- With add() running in parallel we can do vector addition
- Terminology: each parallel invocation of add() is referred to as a thread
 - Each invocation can refer to its thread index using threadIdx.x

```
__global__ void add(int *a, int *b, int *c) {
    c[threadIdx.x] = a[threadIdx.x] + b[threadIdx.x];
}
```

By using threadIdx.x to index into the array, each thread handles a different index

Vector Addition on the Device

```
__global__ void add(int *a, int *b, int *c) {
    c[threadIdx.x] = a[threadIdx.x] + b[threadIdx.x];
}
```

On the device, each thread can execute in parallel:

Vector Addition on the Device: add()

Returning to our parallelized add() kernel

```
__global__ void add(int *a, int *b, int *c) {
    c[threadIdx.x] = a[threadIdx.x] + b[threadIdx.x];
}
```

Let's take a look at main()...

Vector Addition on the Device: main()

```
#define N 512
int main(void) {
   int *a *b *c
                          // host copies of a, b, c
   int *d a, *d b, *d c; // device copies of a, b, c
   int size = N * sizeof(int);
   // Alloc space for device copies of a, b, c
   cudaMalloc((void **)&d a, size);
   cudaMalloc((void **)&d b, size);
   cudaMalloc((void **)&d c, size);
   // Alloc space for host copies of a, b, c and setup input values
   a = (int *)malloc(size); random ints(a, N);
   b = (int *)malloc(size); random ints(b, N);
   c = (int *)malloc(size);
```

Vector Addition on the Device: main()

```
// Copy inputs to device
cudaMemcpy(d_a, a, size, cudaMemcpyHostToDevice);
cudaMemcpy(d b, b, size, cudaMemcpyHostToDevice);
// Launch add() kernel on GPU with N blocks
add <<<1, N>>> (d a, d b, d c);
// Copy result back to host
cudaMemcpy(c, d c, size, cudaMemcpyDeviceToHost);
// Cleanup
free(a); free(b); free(c);
cudaFree(d a); cudaFree(d b); cudaFree(d c);
return 0;
```

Combining Blocks and Threads

- We've seen parallel vector addition using:
 - Many blocks with one thread each
 - One block with many threads
- Let's adapt vector addition to use both blocks and threads
- Why? We'll come to that...
- First let's discuss data indexing...

Indexing Arrays with Blocks and Threads

- No longer as simple as using blockIdx.x and threadIdx.x
 - Consider indexing an array with one element per thread (8 threads/block)

```
threadIdx.x threadIdx.x threadIdx.x

0 1 2 3 4 5 6 7 0 1 2 3 4 5 6 7 0 1 2 3 4 5 6 7

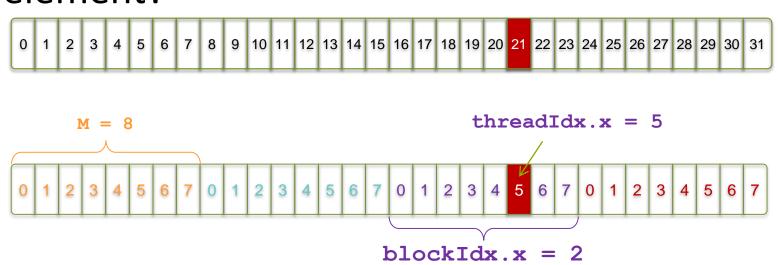
blockIdx.x = 0 blockIdx.x = 1 blockIdx.x = 2 blockIdx.x = 3
```

 With M threads/block a unique index for each thread is given by:

```
int index = threadIdx.x + blockIdx.x * M;
```

Indexing Arrays: Example

Which thread will operate on the red element?



```
int index = threadIdx.x + blockIdx.x * M;
= 5 + 2 * 8;
= 21;
```

Vector Addition with Blocks and Threads

• Use the built-in variable blockDim.x for threads per block

```
int index = threadIdx.x + blockIdx.x * blockDim.x;
```

 Combined version of add() to use parallel threads and parallel blocks

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

What changes need to be made in main()?

Addition with Blocks and Threads: main()

```
#define N (2048*2048)
#define THREADS PER BLOCK 512
int main(void) {
   int *a, *b, *c;
                                         // host copies of a, b, c
    int *d a, *d b, *d c; // device copies of a, b, c
    int size = N * sizeof(int);
    // Alloc space for device copies of a, b, c
    cudaMalloc((void **)&d a, size);
    cudaMalloc((void **)&d b, size);
    cudaMalloc((void **)&d c, size);
    // Alloc space for host copies of a, b, c and setup input values
   a = (int *)malloc(size); random ints(a, N);
   b = (int *)malloc(size); random ints(b, N);
   c = (int *)malloc(size);
```

Addition with Blocks and Threads: main()

```
// Copy inputs to device
cudaMemcpy(d a, a, size, cudaMemcpyHostToDevice);
cudaMemcpy(d b, b, size, cudaMemcpyHostToDevice);
// Launch add() kernel on GPU
add<<<N/THREADS PER BLOCK, THREADS PER BLOCK>>>(d a, d b, d c);
// Copy result back to host
cudaMemcpy(c, d_c, size, cudaMemcpyDeviceToHost);
// Cleanup
free(a); free(b); free(c);
cudaFree(d a); cudaFree(d b); cudaFree(d c);
return 0;
```

Handling Arbitrary Vector Sizes

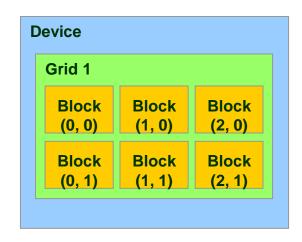
- Typical problems are not friendly multiples of
- Avoid accessing beyond the end of the arrays:

```
__global___ void add(int *a, int *b, int *c, int n) {
    int index = threadIdx.x + blockIdx.x * blockDim.x;
    if (index < n)
        c[index] = a[index] + b[index];
}</pre>
```

Update the kernel launch:

```
add <<<(N + M-1) / M,M>>>(d_a, d_b, d_c, N);
```

- dim3 grid(3,2);
- kernel<<grid, 1>>(...);



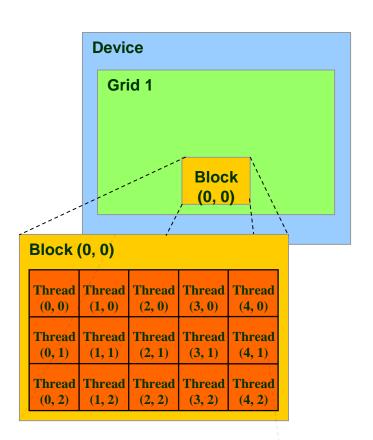
- dim3 grid (3,2);
- kernel<<<grid, 1>>>(...);

```
        Block (0, 0)
        Block (1, 0)
        Block (2, 0)

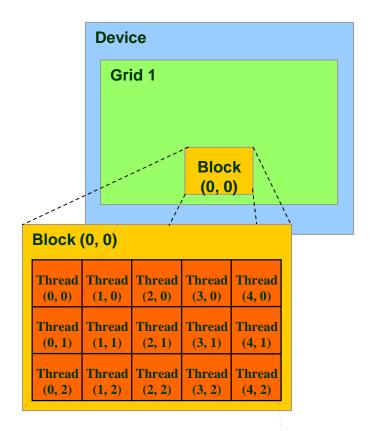
        Block (0, 1)
        Block (1, 1)
        Block (2, 1)
```

```
int index = blockIdx.x + blockIdx.y * gridDim.x;
```

- dim3 threads(5,3);
- kernel<<<1, threads>>>(...);

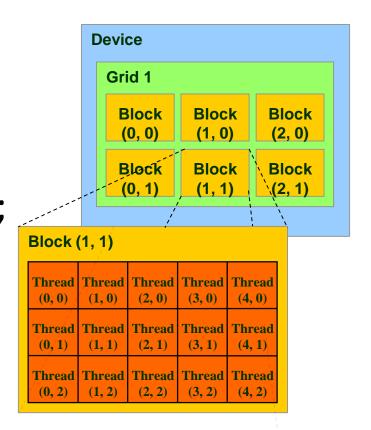


- dim3 threads(5,3);
- kernel<<1, threads>>(...);



```
• int index = threadIdx.x + threadIdx.y * blockDim.x;
```

- dim3 grid (3,2);
- dim3 block(5,3);
- kernel<<<grid, block>>>(...);



- dim3 grid(3,2);
- dim3 block(5,3);
- kernel<<<grid, block>>>(...);

