

CUDA C/C++ BASICS (cont.)

NVIDIA Corporation

COOPERATING THREADS

CONCEPTS

Heterogeneous Computing

Blocks

Threads

Indexing

Shared memory

__syncthreads()

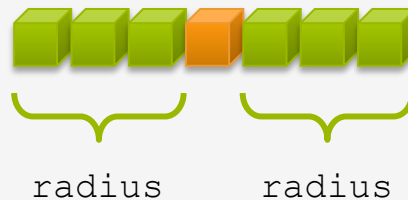
Asynchronous operation

Handling errors

Managing devices

1D Stencil

- Consider applying a 1D stencil to a 1D array of elements
 - Each output element is the sum of input elements within a radius
- If radius is 3, then each output element is the sum of 7 input elements:



Implementing Within a Block

- Each thread processes one output element
 - blockDim.x elements per block
- Input elements are read several times
 - With radius 3, each input element is read seven times

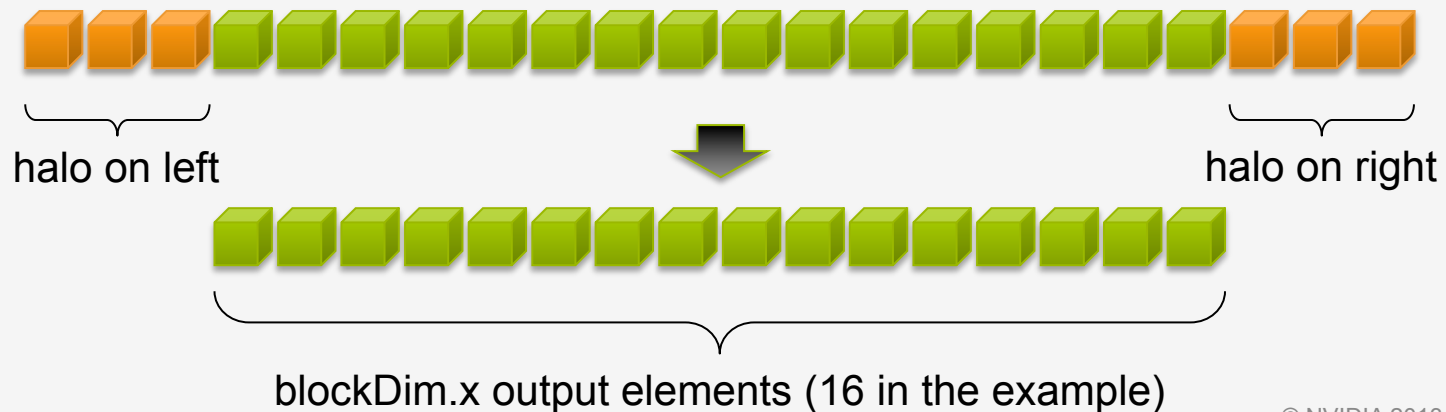


Sharing Data Between Threads

- Terminology: within a block, threads share data via `shared memory`
- Extremely fast on-chip memory, user-managed
- Declare using `__shared__`, allocated per block
- Data is not visible to threads in other blocks

Implementing With Shared Memory

- Cache data in shared memory
 - Read ($\text{blockDim.x} + 2 * \text{radius}$) input elements from global memory to shared memory
 - Compute blockDim.x output elements
 - Write blockDim.x output elements to global memory
- Each block needs a **halo** of radius elements at each boundary



Stencil Kernel

```
__global__ void stencil_1d(int *in, int *out) {  
    __shared__ int temp[BLOCK_SIZE + 2 * RADIUS];  
    int gindex = threadIdx.x + blockIdx.x * blockDim.x;  
    int lindex = threadIdx.x + RADIUS;  
  
    // Read input elements into shared memory  
    temp[lindex] = in[gindex];  
    if (threadIdx.x < RADIUS) {  
        temp[lindex - RADIUS] = in[gindex - RADIUS];  
        temp[lindex + BLOCK_SIZE] =  
            in[gindex + BLOCK_SIZE];  
    }  
}
```





Stencil Kernel

```
// Apply the stencil  
int result = 0;  
for (int offset = -RADIUS ; offset <= RADIUS ; offset++)  
    result += temp[lindex + offset];  
  
// Store the result  
out[gindex] = result;  
}
```


Data Race!

- The stencil example will not work...
- Suppose thread 15 reads the halo before thread 0 has fetched it...

```
temp[lindex] = in[gindex];           Store at temp[18]   
if (threadIdx.x < RADIUS) {  
    temp[lindex - RADIUS] = in[gindex - RADIUS];    Skipped, threadIdx > RADIUS  
    temp[lindex + BLOCK_SIZE] = in[gindex + BLOCK_SIZE];  
}  
  
int result = 0;  
result += temp[lindex + 1];          Load from temp[19] 
```

__syncthreads()

- `void __syncthreads();`
- Synchronizes all threads within a block
 - Used to prevent RAW / WAR / WAW hazards
- All threads must reach the barrier
 - In conditional code, the condition must be uniform across the block

Stencil Kernel

```
__global__ void stencil_1d(int *in, int *out) {  
    __shared__ int temp[BLOCK_SIZE + 2 * RADIUS];  
    int gindex = threadIdx.x + blockIdx.x * blockDim.x;  
    int lindex = threadIdx.x + radius;  
  
    // Read input elements into shared memory  
    temp[lindex] = in[gindex];  
    if (threadIdx.x < RADIUS) {  
        temp[lindex - RADIUS] = in[gindex - RADIUS];  
        temp[lindex + BLOCK_SIZE] = in[gindex + BLOCK_SIZE];  
    }  
  
    // Synchronize (ensure all the data is available)  
    __syncthreads();  
}
```

Stencil Kernel

```
// Apply the stencil
```

```
int result = 0;
```

```
for (int offset = -RADIUS ; offset <= RADIUS ; offset++)  
    result += temp[lindex + offset];
```

```
// Store the result
```

```
out[gindex] = result;
```

```
}
```

Review (1 of 2)

- Launching parallel threads
 - Launch N blocks with M threads per block with
`kernel<<<N,M>>> (...);`
 - Use `blockIdx.x` to access block index within grid
 - Use `threadIdx.x` to access thread index within block

- Allocate elements to threads:

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

Review (2 of 2)

- Use `__shared__` to declare a variable/array in shared memory
 - Data is shared between threads in a block
 - Not visible to threads in other blocks
- Use `__syncthreads()` as a barrier
 - Use to prevent data hazards

MANAGING THE DEVICE

CONCEPTS



- Heterogeneous Computing
- Blocks
- Threads
- Indexing
- Shared memory
- `__syncthreads()`
- Asynchronous operation
- Handling errors
- Managing devices

Coordinating Host & Device

- Kernel launches are **asynchronous**
 - Control returns to the CPU immediately
- CPU needs to synchronize before consuming the results

cudaMemcpy ()

Blocks the CPU until the copy is complete
Copy begins when all preceding CUDA calls have completed

cudaMemcpyAsync ()

Asynchronous, does not block the CPU

cudaDeviceSynchronize ()

Blocks the CPU until all preceding CUDA calls have completed

Reporting Errors

- All CUDA API calls return an error code (`cudaError_t`)
 - Error in the API call itself
 - OR
 - Error in an earlier asynchronous operation (e.g. kernel)

- Get the error code for the last error:

```
cudaError_t cudaGetLastError(void)
```

- Get a string to describe the error:

```
char *cudaGetErrorString(cudaError_t)
```

```
printf("%s\n", cudaGetErrorString(cudaGetLastError()));
```

Device Management

- Application can query and select GPUs

```
cudaGetDeviceCount(int *count)
```

```
cudaSetDevice(int device)
```

```
cudaGetDevice(int *device)
```

```
cudaGetDeviceProperties(cudaDeviceProp *prop, int device)
```

- Multiple threads can share a device
- A single thread can manage multiple devices

```
cudaSetDevice(i) to select current device
```

```
cudaMemcpy(...) for peer-to-peer copies†
```

[†] requires OS and device support

Introduction to CUDA C/C++

- What have we learned?
 - Write and launch CUDA C/C++ kernels
 - `__global__`, `blockIdx.x`, `threadIdx.x`, `<<<>>>`
 - Manage GPU memory
 - `cudaMalloc()`, `cudaMemcpy()`, `cudaFree()`
 - Manage communication and synchronization
 - `__shared__`, `__syncthreads()`
 - `cudaMemcpy()` VS `cudaMemcpyAsync()`,
`cudaDeviceSynchronize()`

Compute Capability

- The **compute capability** of a device describes its architecture, e.g.
 - Number of registers
 - Sizes of memories
 - Features & capabilities

Compute Capability	Selected Features (see CUDA C Programming Guide for complete list)	Tesla models
1.0	Fundamental CUDA support	870
1.3	Double precision, improved memory accesses, atomics	10-series
2.0	Caches, fused multiply-add, 3D grids, surfaces, ECC, P2P, concurrent kernels/copies, function pointers, recursion	20-series

- The following presentations concentrate on Fermi devices
 - Compute Capability ≥ 2.0

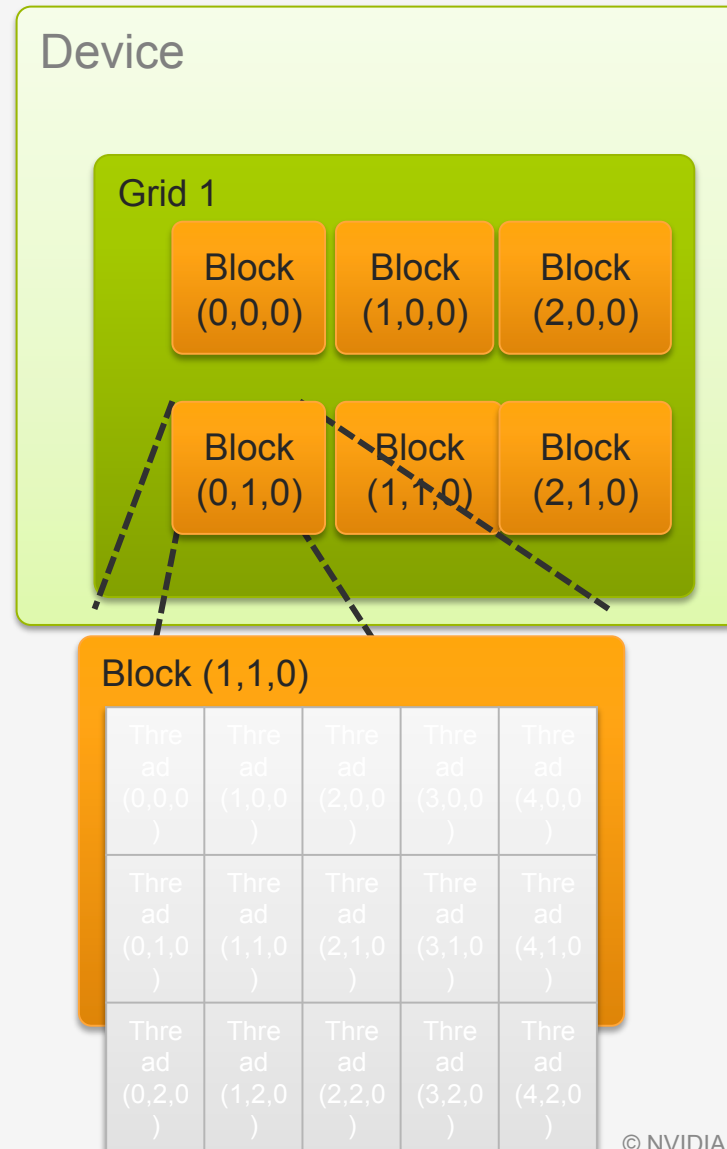
IDs and Dimensions

– A kernel is launched as a grid of blocks of threads

- `blockIdx` and `threadIdx` are 3D
- We showed only one dimension (x)

- **Built-in variables:**

- `threadIdx`
- `blockIdx`
- `blockDim`
- `gridDim`



Topics we skipped

- We skipped some details, you can learn more:
 - CUDA Programming Guide
 - CUDA Zone – tools, training, webinars and more
developer.nvidia.com/cuda