

INTRODUCTION TO PROGRAMMING IN CUDA C

Will Landau, Prof. Jarad Niemi

BASIC C PROGRAM

```
#include <iostream>

int main ( void ) {
    printf( "Hello, World!\n" );
    return 0;
}
```

BASIC CUDA C PROGRAM

```
#include <iostream>

__global__ void kernel ( void ) {
}

int main ( void ) {
    kernel<<<1,1>>>();
    printf( "Hello, World!\n" );
    return 0;
}
```

```
#include <iostream>

__global__ void kernel ( void ) {
}
```

Prefix that says: "only run this function on the GPU"

```
int main ( void ) {
    kernel<<<1,1>>>();
    printf( "Hello, World!\n" );
    return 0;
}
```

`__global__`: Call from CPU and run only on GPU.

`__device__`: Call from GPU and run only on GPU.

(More specifically, call only from within
a `__global__` or another `__device__` function.)

`__host__`: Call from CPU and run only on CPU.

(i.e., a traditional C function.)

	Executed on the:	Only callable from the:
<code>__device__ float DeviceFunc()</code>	device	device
<code>__global__ void KernelFunc()</code>	device	host
<code>__host__ float HostFunc()</code>	host	host

```
__device__ int dev1( void ){
}

__device__ int dev2( void ){
}

__global__ void kernel ( void ) {
    dev1();
    dev2();
}

int main ( void ) {
    kernel<<<1, 1>>>();
    printf( "Hello, World!\n" );
    return 0;
}
```

```
#include <iostream>

__global__ void kernel ( void ) {
}

int main ( void ) { ← Started on CPU
    kernel<<<1,1>>>(); ← Sent to GPU
    printf( "Hello, World!\n" ); ← Done
                                on CPU
    return 0; ← Done on CPU
}
```

```
#include <iostream>

__global__ void kernel ( void ) {
}

int main ( void ) {
    kernel<<<1,1>>>(); ?
    printf( "Hello, World!\n" );
    return 0;
}
```

REVIEW OF PARALLELIZATION

Parallelization: Running different calculations simultaneously.

Kernel: An instruction set executed on the GPU. (All others are executed on the CPU.)

In CUDA C, a kernel is any function prefixed with the keyword, `--global__`. For example:

```
--global__ void kernel ( void ) {  
}
```

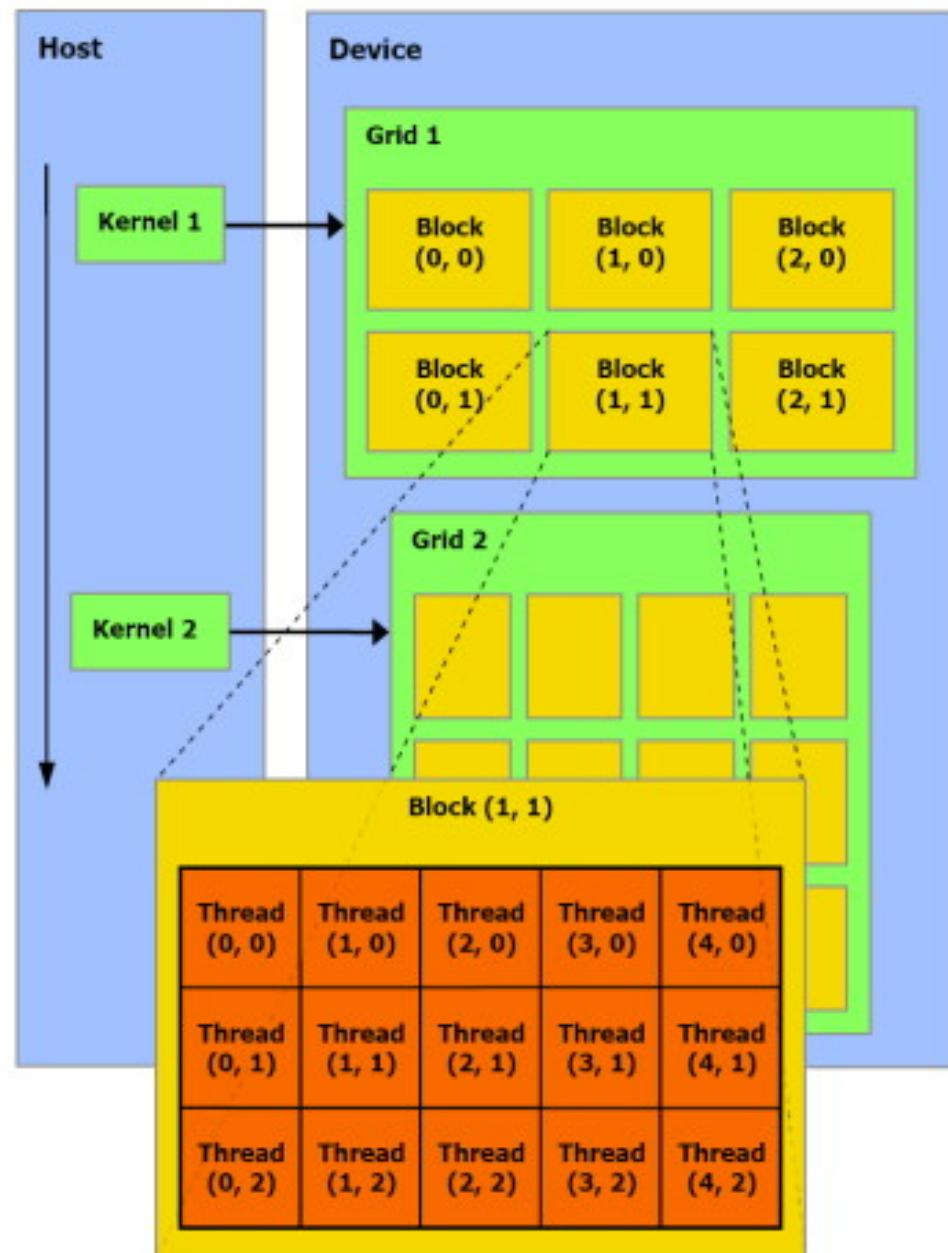
PARALLELIZATION AND THE SIMD PARADIGM

SIMD: Single Instruction Multiple Data. The paradigm works like this:

1. The CPU sends a kernel (a SINGLE INSTRUCTION set) to the GPU.
2. The GPU executes the kernel MULTIPLE times simultaneously in **PARALLEL**. Each such execution of the kernel is called a **thread**, and each thread is executed on different parts of a giant data set.

PARALLELIZING A WORKLOAD ON A GPU

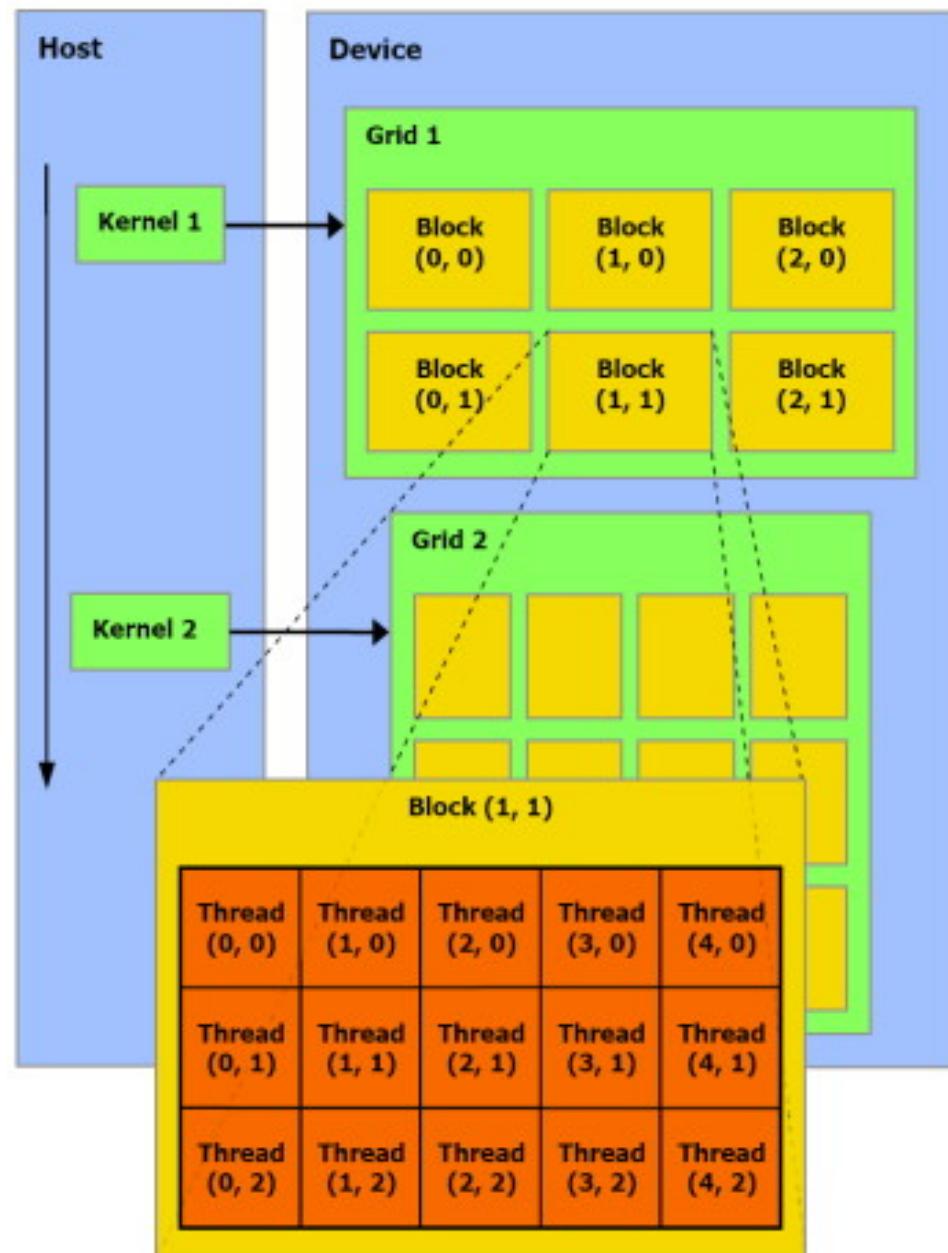
1. The CPU sends a CPU-to-GPU command called a **kernel** to a single GPU core.
2. The GPU core multitasks to execute the command:
 - a. The GPU makes $N \cdot M$ **copies** of the kernel's code, and then runs all those copies simultaneously. Those parallel copies are called **threads**.
 - b. The $N \cdot M$ threads are partitioned into N groups, called **blocks**, of M threads each.
 - c. The sum total of all the threads from a kernel call is a **grid**.



ORGANIZATION OF THREADS

Grid; The collection of all the threads that are spawned when the CPU sends a kernel to the GPU.

Block: A collection of threads within a grid that share memory quickly and easily.



NOW BACK TO THE STUFF IN ANGLE BRACKTES...

```
#include <iostream>

__global__ void kernel ( void ) {
} Uses one grid.      Number of blocks
int main ( void ) {
    kernel<<<1,1>>>();
    printf( "Hello, World!\n" );
    return 0;
}
```

Number of threads per block

The code demonstrates a CUDA kernel launch with a single grid and one block. The kernel has no parameters. The main function launches the kernel with a grid size of 1 and a block size of 1, containing 1 thread per block.

Here, the GPU runs kernel() one time:

```
#include <iostream>

__global__ void kernel ( void ) {
}

int main ( void ) {
    kernel<<<1,1>>>();
    printf( "Hello, World!" \n" )
    return 0;
}
```

Here, the GPU runs `kernel()` 5 times:

```
#include <iostream>

__global__ void kernel ( void ) {
}

int main ( void ) {
    kernel<<<5,1>>>();
    printf( "Hello, World!" \n" )
    return 0;
}
```

Here, the GPU runs `kernel()` 5 times:

```
#include <iostream>

__global__ void kernel ( void ) {
}

int main ( void ) {
    kernel<<<1,5>>>();
    printf( "Hello, World!" \n" )
    return 0;
}
```

Here, the GPU runs kernel() 20 times:

```
#include <iostream>

__global__ void kernel ( void ) {
}

int main ( void ) {
    kernel<<<4,5>>>();
    printf( "Hello, World!" \n" )
    return 0;
}
```

PASSING DATA TO AND FROM THE GPU: simple1.cu

```
#include <stdio.h>
#include <stdlib.h>

__global__ void colonel(int *dev_a){
    *dev_a = 1;
}

int main(){

    // Declare variables and allocate memory on the GPU.
    int a[1], *dev_a;
    cudaMalloc((void**) &dev_a, sizeof(int));

    // Execute kernel and copy the result to CPU memory.
    colonel<<<1,1>>>(dev_a);
    cudaMemcpy(a, dev_a, sizeof(int), cudaMemcpyDeviceToHost);

    // Print result and free dynamically allocated memory.
    printf("a[0] = %d\n", a[0]); // REMEMBER: INDEXING IN C STARTS FROM 0.
    cudaFree(dev_a);

}
```

```
[landau@impact1 simple]$ nvcc simple1.cu -o simple1.out
[landau@impact1 simple]$ ./simple1.out
a[0] = 1
[landau@impact1 simple]$ |
```

PASSING DATA TO AND FROM THE GPU: simple2.cu

```
#include <stdio.h>
#include <stdlib.h>

__global__ void colonel(int *dev_a){
    *dev_a = *dev_a + 1;
}

int main(){
    // Declare variables and allocate memory on the GPU.
    int a[1], *dev_a;
    cudaMalloc((void**) &dev_a, sizeof(int));

    // Initialize argument a, executed kernel, and store result back in a.
    a[0] = 1; // REMEMBER: INDEXING IN C STARTS FROM 0.
    cudaMemcpy(dev_a, a, sizeof(int), cudaMemcpyHostToDevice);
    colonel<<<1,1>>>(dev_a);
    cudaMemcpy(a, dev_a, sizeof(int), cudaMemcpyDeviceToHost);

    // Print result and free dynamically allocated memory.
    printf("a[0] = %d\n", a[0]); // REMEMBER: INDEXING IN C STARTS FROM 0.
    cudaFree(dev_a);
}
```

```
[landau@impact1 simple]$ nvcc simple2.cu -o simple2.out
[landau@impact1 simple]$ ./simple2.out
b[0] = 2
[landau@impact1 simple]$
```

threadIdx.x AND blockIdx.x

Say the CPU sends a kernel to the GPU with B blocks and T threads per block.

The following CUDA C variables are available for use in the kernel:

blockIdx.x: the block ID corresponding to the current thread, an integer from 0 to $B - 1$ inclusive.

threadIdx.x: the thread ID of the current thread within its block, an integer from 0 to $T - 1$ inclusive.

NOTE: there are also variables like `blockIdx.y` and `threadIdx.y`, but those are only useful if you deliberately organize your blocks and threads in two-dimensional arrays (which can be done).

USING blockIdx.x AND threadIdx.x: identify.cu

```
#include <stdio.h>
#include <stdlib.h>

__global__ void isExecuted(int *dev_a, int blockid, int threadid){

    if(blockIdx.x == blockid && threadIdx.x == threadid)
        *dev_a = 1;

}

int main(){

    // Declare variables and allocate memory on the GPU.
    int init[1], a[1], *dev_a;
```

```
cudaMalloc((void**) &dev_a, sizeof(int));

// Initialize dev_a, execute kernel, and copy the result to CPU memory.
init[1] = 0;
cudaMemcpy(dev_a, init, sizeof(int), cudaMemcpyHostToDevice);
isExecuted<<<100,100>>>(dev_a, 2, 4);
    // NOTE: INDEXING OF THREADS AND BLOCKS STARTS FROM 0.
cudaMemcpy(a, dev_a, sizeof(int), cudaMemcpyDeviceToHost);

// Print result and free dynamically allocated memory.
printf("a[0] = %d\n", a[0]); // REMEMBER: INDEXING IN C STARTS FROM 0.
cudaFree(dev_a);

}
```

```
[landau@impact1 simple]$ nvcc identify.cu -o identify.out
[landau@impact1 simple]$ ./identify.out
a[0] = 1
[landau@impact1 simple]$ |
```

```
isExecuted<<<2,3>>>(dev_a, blockid = 1, threadid = 1);  
  
    ┌─────────────────────────────────────────────────────────────────────────┐  
    | isExecuted(dev_a, blockid = 1, threadid = 1, blockIdx.x = 0, threadIdx.x = 0, ... ); → *dev_a is unchanged.  
    | isExecuted(dev_a, blockid = 1, threadid = 1, blockIdx.x = 0, threadIdx.x = 1, ... ); → *dev_a is unchanged.  
    | isExecuted(dev_a, blockid = 1, threadid = 1, blockIdx.x = 0, threadIdx.x = 2, ... ); → *dev_a is unchanged.  
    | isExecuted(dev_a, blockid = 1, threadid = 1, blockIdx.x = 1, threadIdx.x = 0, ... ); → *dev_a is unchanged.  
    | isExecuted(dev_a, blockid = 1, threadid = 1, blockIdx.x = 1, threadIdx.x = 1, ... ); → *dev_a is set to 1.  
    | isExecuted(dev_a, blockid = 1, threadid = 1, blockIdx.x = 1, threadIdx.x = 2, ... ); → *dev_a is unchanged.  
    └─────────────────────────────────────────────────────────────────────────┘
```

You can verify that when the line,

```
isExecuted<<<1,1>>>(dev_a, 0, 0);
```

is changed to:

```
isExecuted<<<1,1>>>(dev_a, 2, 4);
```

the output of the program:

```
a[0] = 0
```

which makes sense because there was no block number 2 or thread number 4.

ASIDE: AVOIDING PITFALLS IN PARALLELIZATION

What problem would arise if I used the function, isExecuted_gooofup(), instead of Executed():

```
__global__ void isExecuted_gooofup(int *dev_a, int blockid,
                                    int threadid){

    if(blockIdx.x == blockid && threadIdx.x == threadid)
        *dev_a = 1;
    else
        *dev_a = 0;

}
```

```
isExecuted_goofup<<<2,3>>>(dev_a, blockid = 1, threadid = 1);  
  
    └─ isExecuted_goofup(dev_a, blockid = 1, threadid = 1, blockIdx.x = 0, threadIdx.x = 0, ... ); → *dev_a is set to 0.  
    └─ isExecuted_goofup(dev_a, blockid = 1, threadid = 1, blockIdx.x = 0, threadIdx.x = 1, ... ); → *dev_a is set to 0.  
    └─ isExecuted_goofup(dev_a, blockid = 1, threadid = 1, blockIdx.x = 0, threadIdx.x = 2, ... ); → *dev_a is set to 0.  
    └─ isExecuted_goofup(dev_a, blockid = 1, threadid = 1, blockIdx.x = 1, threadIdx.x = 0, ... ); → *dev_a is set to 0.  
    └─ isExecuted_goofup(dev_a, blockid = 1, threadid = 1, blockIdx.x = 1, threadIdx.x = 1, ... ); → *dev_a is set to 1.  
    └─ isExecuted_goofup(dev_a, blockid = 1, threadid = 1, blockIdx.x = 1, threadIdx.x = 2, ... ); → *dev_a is set to 0.
```

If block 1 thread 1 finishes last, then $*\text{dev_a} = 1$.

If block 1 thread 1 does not finish last, $*\text{dev_a} = 0$.

We can't know or control which thread finishes last!

LECTURE SERIES MATERIALS

These lecture slides, a tentative syllabus for the whole lecture series, and code are available at:

<https://github.com/wlandau/gpu>.

After logging into you home directory on impact1, type:

```
git clone https://github.com/wlandau/gpu
```

into the command line to download all the course materials.

REFERENCES

- David B. Kirk and Wen-mei W. Hwu. “Programming Massively Parallel Processors: a Hands-on Approach.” Morgan Kaufman, 2010.
- J. Sanders and E. Kandrot. *CUDA by Example*. Addison-Wesley, 2010.