# Your first CUDA program

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## Introduction

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## Purpose

In this lab, you will learn how to adapt a simple code such that it uses the GPU.

It has the purpose of negating an array of integers. We introduce the important concepts of device-memory management and kernel invocation. The final version should copy an array of integers from the host to device, multiply each element by -1 on the device, and then copy the array back to the host.

Choose the C or Fortran version.

#### Source code

You can get this from GitHub:

git clone https://github.com/rupertnash/pages-expt.git
cd pages-expt/exercises/cuda-intro/

#### Note

The template source file is clearly marked with the sections to be edited, e.g.

```
/* Part 1A: allocate device memory */
```

Where necessary, you should refer to the CUDA C Programming Guide and Reference Manual documents available from http://developer.nvidia.com/nvidia-gpu-computing-documentation.

## Part 1

## Copying Between Host and Device

## $\mathbf{C}$

Start from the intro.cu template.

#### 1A

Allocate memory for the array on the device: use the existing pointer d\_a and the variable sz (which has already been assigned the size of the array in bytes).

#### 1B

Copy the array h\_a on the host to d\_a on the device.

## 1C

Copy d\_a on the device back to h\_out on the host.

## 1D

Free d\_a.

## **Fortran**

Start from the intro.cuf template.

## **1A**

Allocate memory for the array on the device: use the existing pointer d\_a and ARRAY\_SIZE (which has already been assigned the size of the array in elements)

#### 1B

Copy the array  $h_a$  on the host to  $d_a$  on the device, using an appropriate assignment operation.

#### 1C

Copy  ${\tt d\_a}$  on the device back to  ${\tt h\_out}$  on the host, using another assignment operation.

## 1D

Deallocate d\_a.

# Compilation

## Load modules

```
For C/C++
module load gcc cuda
Fortran also required the PGI compilers
module load gcc cuda pgi
```

#### Use make

Compile the code using make. Note that the compute capability of the CUDA device is specified with the -arch flag for C and with -March= for Fortran.

## Running

#### On Cirrus

You can only run on the backend nodes, so you can submit the job to the batch system with qsub submit.sh

During tutorials we have reserved one node (4 GPUs) for the use of the class. You can access this by editing the PBS script to use the reserved queue. This will be given in the class.

The output (the contents of the h\_out array) or any error messages will be printed. So far the code simply copies from h\_a on the host to d\_a on the device, then copies d\_a back to h\_out, so the output should be the initial content of h\_a - the numbers 0 to 255.

## Part 2

## Launching Kernels

Now we will actually run a kernel on the GPU device.

## $\mathbf{C}$

## 2A

Configure and launch the kernel using a 1D grid and a single thread block (NUM\_BLOCKS and THREADS\_PER\_BLOCK are already defined for this case).

#### 2B

Implement the actual kernel function to negate an array element as follows:

```
int idx = threadIdx.x;
d_a[idx] = -1 * d_a[idx];
```

Compile and run the code as before. This time the output should contain the result of negating each element of the input array. Because the array is initialised to the numbers 0 to 255, you should see the numbers 0 down to -255 printed.

This kernel works, but since it only uses one thread block, it will only be utilising one of the multiple SMs available on the GPU. Multiple thread blocks are needed to fully utilize the available resources.

#### 2C

Implement the kernel again, this time allowing multiple thread blocks. It will be very similar to the previous kernel implementation except that the array index will be computed differently:

```
int idx = threadIdx.x + (blockIdx.x * blockDim.x);
```

Remember to also change the kernel invocation to invoke negate\_multiblock this time. With this version you can change NUM\_BLOCKS and THREADS\_PER\_BLOCK to have different values - so long as they still multiply to give the array size.

## **Fortran**

#### 2A

Configure and launch the kernel using a 1D grid and a single thread block (NUM\_BLOCKS and THREADS\_PER\_BLOCK are already defined for this case).

#### 2B

Implement the actual kernel function to negate an array element as follows:

```
integer :: idx
idx = threadidx%x
aa(idx) = -1*aa(idx)
```

Compile and run the code as before. This time the output should contain the result of negating each element of the input array. Because the array is initialised to the numbers 0 to 255, you should see the numbers 0 down to -255 printed.

This kernel works, but since it only uses one thread block, it will only be utilising one of the multiple SMs available on the GPU. Multiple thread blocks are needed to fully utilize the available resources.

#### 2C

Implement the kernel again, this time allowing multiple thread blocks. It will be very similar to the previous kernel implementation except that the array index will be computed differently:

idx = threadidx%x + ((blockidx%x-1) \* blockdim%x)

Remember to also change the kernel invocation to invoke g\_negate\_multiblock this time. With this version you can change NUM\_BLOCKS and THREADS\_PER\_BLOCK to have different values - so long as they still multiply to give the array size.

## Part 3

## Handling any size array

Currently we are insisting that the array size be an exact multiple of the block size. In general we should handle any size that will fit in GPU memory!

Let the total number of elements be N and the block size be B.

Recall that in integer division we discard the fractional part so we can write:

$$N = k * B + r$$

i.e. N can divided into k (an integer) number of blocks, plus a remainder, r. If r is zero, then we need k blocks, else we need k+1.

This can be expressed in a simple formula:

$$nBlocks = \frac{N-1}{B} + 1$$

Convince yourself this is correct!

#### 3A

Update the kernel launch code to compute the number of blocks using this formula.

What will happen in the last block with the current kernel?

## 3B

Implement a condition in the kernel to protect against this. Try changing ARRAY\_SIZE to a non-multiple of 256 (e.g. 500).