





Introduction to GPUs in HPC

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Using GPUs in Your Application

Rule #1: don't develop your own GPU code!



Libraries

There are many open libraries for GPUs.

- cuBLAS: Dense linear algebra primitives.
- Thrust: C++ STL-like algorithms and containers.
- cuRAND and Random123: Random numbers.
- cuFFT: FFT
- Kokkos: Generic performance portable parallel motifs.

... And many more!

Take some time to investigate what is available before starting.



You want to write your own code?

Directives

- OpenACC and OpenMP 4 define directives that can be used to instruct the compiler how to generate GPU code
- in theory the easiest path for porting

GPU-specific Languages

- languages designed for GPU programming
- maximum flexibility and performance
- e.g. CUDA and OpenCL





Things to consider

Before starting on a GPU implementation, it pays to ask some questions and do some preliminary exploration:

- 1. Is my program computationally or bandwidth intensive?
- 2. Does it have enough parallel work to utilize the GPU?
- 3. Must I change algorithms to expose enough parallelelism?
- 4. Are there serial bottlenecks that will limit scaling?
- 5. Is the pain worth the gain?
 - Questions 1, 2 and 3 will be discussed in this course.
 - Question 4 will be considered briefly here.
- Questions 5 requires answers for 1–4.





Limitations to parallel speedup

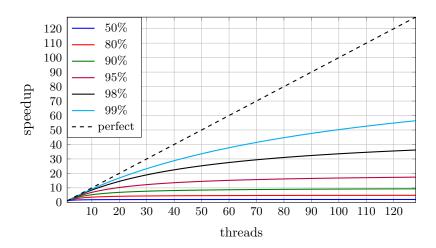
- Parallel speedup is limited by the proportion of serial work in your code.
- Amdahl's law defines the maximum possible speedup when only parts of the code can be parallelised

$$t_n = t_1 \left(p + \frac{(1-p)}{n} \right),\,$$

where t_n is time to solution for n threads and $p \in [0, 1]$ is the proportion of sequential code.

- The limit on time to solution is $\lim_{n\to\infty} = pt_1$
 - $^-$ e.g. 1% of serial code gives a maximum 100× speedup.

Amdahl illustrated





CUDA

CUDA is a parallel computing platform and API

• For CUDA-enabled Nvidia GPUs.

We use CUDA as short hand for CUDA C/C++ and API

- CUDA C++ is a **superset** of C++
- Adds keywords for writing kernels to run on the GPU.
- Adds syntax for launching kernels on the GPU.

The CUDA toolkit is more than a programming language:

- Runtime API for managing GPU resources and execution.
- Tools including profilers and debuggers.





Compiling CUDA

CUDA code is compiled with the **nvcc** compiler driver

- source files have .cu extension
- headers have .h, .hpp, .hcu extension.

CUDA compilation involves multiple splitting, compilation, preprocessing and merging steps

- nvcc hides this complexity from the user.
- It closely mimics the interface of the GNU compiler.
- Behind the scenes it:
 - uses GCC to compile the code that runs on CPU;
 - and compiles the GPU code separately.





Compiling CUDA

Example CUDA compilation

```
> nvcc -arch=sm_60 -lineinfo -02 -std=c++11 -g -o foo foo.cu
```

Some flags are for **device** code generation:

- -arch=sm_60 target GPU architecture (Pascal)
- -lineinfo debug information for device code.

Some are for **host**:

-g debug information for host code.

And some are for both **host and device**:

- -02 optimization level
- -std=c++11 target language
- -o foo name of executable.





Exercise: Getting Started on Piz Daint

In this exercise we will get introduced to Daint and make sure that everybody is set up.

```
# log on to daint with your course username & password
> ssh -X courseXX@daint
# get one node on the course reservation
<<<<<< HEAD
> salloc -C gpu --reservation=summer
> salloc -C gpu --reservation=Pascal1day -p jalet
>>>>> master
# go to scratch and get the course material
> cd $SCRATCH/my_repo
> git pull
# compile and test the demo
> cd topics/cuda/practicals/demos
> cat hello.cu
> module load cudatoolkit gcc
> nvcc -arch=sm_60 hello.cu -o hello
> srun ./hello
```

