





# Introduction to GPUs in HPC

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# Concurrency

# Concurrency

Concurrency is the ability to perform multiple CUDA operations simultaneously, including:

- CUDA kernels;
- Copying from host to device;
- Copying from device to host;
- Operations on the host CPU.

### Concurrency enables

- Both CPU and GPU can work at the same time.
- Multiple tasks can be run on GPU simultaneously.
- Overlapping of communication and computation.





#### Host code kernel\_1 <<< ... >>> (...); kernel\_2 < < < . . . > >> ( . . . ); host\_1(...); host\_2(...);

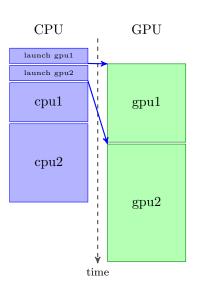
#### The host:

- launches the two CUDA kernels;
- then executes host calls sequentially.

#### The GPU:

- executes asynchronously to host;
- executes kernels sequentially.







The CUDA language and runtime libraries provide mechanisms for coordinating asynchronous GPU execution:

- CUDA streams can concurrently run independent kernels and memory transfers;
- CUDA events can be used to synchronize streams and query the status of kernels and transfers.





### Streams

A CUDA str is a sequence of operations that execute in issue order on the GPU.

 CUDA operations are kernels and copies between host and device memory spaces.

#### Streams and concurrency

- Operations in different streams may run concurrently
  - requires sufficient resources on the GPU (registers, shared memory, SMXs, etc).
- Operations in the same stream are executed sequentially.
- If no stream is specified, all kernels are launched in the default stream.



# Managing streams

cudaStreamDestroy(stream);

A stream is represented using a cudaStream\_t type

- cudaStreamCreate(cudaStream t\* s) and
- cudaStreamDestroy(cudaStream\_t s) can be used to create and free CUDA streams respectively.
  - To launch a kernel on a stream specify the stream id as a fourth parameter to the launch syntax:

```
kernel<<<grid_dim, block_dim, shared_size, stream>>>(...)
```

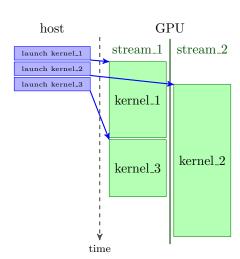
• The default CUDA stream is the NULL stream, or stream ( cudaStream\_t is an integer).



Basic cuda stream usage // create stream cudaStream t stream: cudaStreamCreate(&stream): // launch kernel in stream my\_kernel <<< grid\_dim, block\_dim, shared\_size, stream >>> (..) // release stream when finished

# Host code kernel\_1 <<<\_,\_,\_, stream\_1 >>>(); kernel\_2<<<\_,\_,\_,stream\_2>>>(); kernel\_3<<<\_,\_,\_,stream\_1>>>();

- kernel\_1 and kernel\_3 are serialized in stream\_1
- kernel\_2 can run asynchronously in stream\_2
- Note kernel\_2 will only run concurrently if there are sufficient resources available on the GPU, i.e. if kernel\_1 is not using all of the SMXs.





# Asynchronous copy

#### cudaMemcpyAsync(\*dst, \*src, size, kind, cudaStream\_t stream = 0);

- Takes an additional parameter stream, which is 0 by default.
- Returns immediately after initiating copy:
  - Host can do work while copy is performed;
  - Only if **pinned memory** is used.
- Copies in the same direction (i.e. H2L D2H) are serialized.
  - Copies from host—device and device—host are concurrent if in different streams.





# Pinned memory



Pinned (or page-locked) memory will not be paged out to disk:

- The GPU can safely remotely read/write the memory directly without host involvement;
- Only use for transfers, because it easy to run out of memory.

```
Managing pinned memory
```

```
cudaMallocHost(**ptr, size); and cudaFreeHost(*ptr);
```

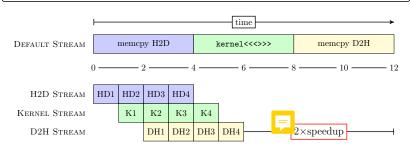
• Allocate and free pinned memory (size is in bytes).



### Asynchronous copy example: streaming workloads

Computations that can be performed independently, e.g. our axpy example:

- Data in host memory has to be copied to the device, and the result copied back after the kernel is computed.
- Overlap copies with kernel calls by breaking the data into chunks.





### CUDA events

To implement the streaming workload we have to coordinate operations on the GPU. CUDA events can be used for this purpose.

- Synchronize tasks in different streams, e.g.:
  - Don't start kernel in kernel stream until data copy stream has finished:
  - Wait until required data has finished copy from host before launching kernel.
- Query status of concurrent tasks:
  - Has kernel finished/started yet?
  - How long did a kernel take to compute?



# Managing events



cudaEventCreate(cudaEvent\_t\*); and cudaEventDestroy(cudaEvent\_t);

Create and free cudaEvent\_t.

cudaEventRecord(cudaEvent\_t, cudaStream\_t);

• Enqueue an event in a stream.

cudaEventSynchronize(cudaEvent\_t);

Make host execution wait for event to occur.

cudaEventQuery(cudaEvent\_t)

• Test if the work before an event in a queue has been completed.

cudaEventElapsedTime(float\*, cudaEvent\_t, cudaEvent\_t);

• Get time between two events.



#### Using events to time kernel execution

```
cudaEvent t start. end:
cudaStream_t stream;
float time_taken;
// initialize the events and streams
cudaEventCreate(&start);
cudaEventCreate(&end);
cudaStreamCreate(&stream):
cudaEventRecord(start, stream); // enqueue start in stream
my_kernel <<< grid_dim, block_dim, 0, stream>>>();
cudaEventRecord(end, stream); // enqueue end in stream
cudaEventSynchronize(end);  // wait for end to be reached
cudaEventElapsedTime(&time_taken, start, end);
std::cout << "kernel took " << 1000*time_taken << " s\n";
// free resources for events and streams
cudaEventDestroy(start);
cudaEventDestroy(end);
cudaStreamDestrov(stream):
```



### Copy→kernel synchronization

```
cudaEvent_t event;
cudaStream_t kernel_stream, h2d_stream;
size_t size = 100*sizeof(double);
double *dptr. *hptr:
// initialize
cudaEventCreate(&event):
cudaStreamCreate(&kernel stream):
cudaStreamCreate(&h2d_stream);
cudaMalloc(&dptr. size):
cudaMallocHost(&hptr, size); // use pinned memory!
// start asynchronous copy in h2d_stream
cudaMemcpyAsync(dptr, hptr, size,
                cudaMemcpyHostToDevice, h2d_stream);
// enqueue event in stream
cudaEventRecord(event, h2d stream):
// make kernel_stream wait for copy to finish
cudaStreamWaitEvent(kernel stream. event. 0):
// enqueue my_kernel to start when event has finished
my_kernel << grid_dim, block_dim, 0, kernel_stream>>>();
// free resources for events and streams
cudaEventDestroy(event);
cudaStreamDestroy(h2d_stream);
cudaStreamDestroy(kernel_stream);
cudaFree(dptr):
cudaFreeHost(hptr);
```



### Exercises

1. Open include/util.hpp and understand

```
copy_to_{host/device}_async() and malloc_pinned()
```

- 2. Open include/cuda\_event.h and include/cuda\_stream.h
  - what is the purpose of these classes?
  - what does cuda\_stream::enqueue\_event() do?
- 3. Open async/memcopy1.cu and run
  - what does the benchmark test?
  - what is the effect of turning on USE\_PINNED? Hint: try small and large values for [n] (8, 16, 20, 24)
- 4. Inspect async/memcopy2.cu and run
  - what effect does changing the number of chunks have?
- 5. Inspect async/memcopy3.cu and run
  - how does it differ from memcopy2.cu? =
  - what effect does changing the number of chunks have?



### Using events to time kernel execution: with helpers

```
CudaStream stream(true);
auto start = stream.enqueue_event();
my_kernel <<< grid_dim , block_dim , 0 , stream .stream() >>> ();
auto end = stream.enqueue event():
end.wait():
auto time_taken = end.time_since(start);
std::cout << "kernel took " << 1000*time taken << " s\n":
```

### Copy—kernel synchronization: with helpers

```
CudaStream kernel_stream(true), h2d_stream(true);
auto size = 100:
auto dptr = device_malloc < double > (size);
auto hptr = pinned_malloc < double > (size);
copy_to_device_async < double > (hptr,dptr,size,h2d_stream.stream());
auto event = h2d_stream.enqueue_event();
kernel_stream.wait_on_event(event);
my_kernel << grid_dim, block_dim, 0, kernel_stream.stream()>>>();
cudaFree(dptr):
cudaFreeHost(hptr):
```





# Rough guidelines for concurrency

Ideally for most workloads you don't want to rely on streams to fill the GPU with work.

- A sign that the working set per GPU is not large enough;
- Full concurrency is difficult in practice;
  - A low-level optimization strategy for the last few %.
- This isn't a hard and fast rule.

Streams come into their own for overlapping communication and computation.

• Possible to transfer data in both directions concurrently with kernel execution.



