Memory allocation on the GPU Exchanging data with the host Bandwidth measurement Use of page-locked memory

Programming in CUDA

Part II. Interacting with the GPU from the host

Frédéric S. Masset

Instituto de Ciencias Físicas, UNAM

Escuela de Verano GPU 2014



- Memory allocation on the GPU
 - cudaMalloc ()
 - cudaMallocPitch () in two dimensions
 - cudaFree ()
- Exchanging data with the host
 - cudaMemcpy ()
- Bandwidth measurement
- Use of page-locked memory
 - cudaMallocHost ()



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Reminder: allocating memory on the host

```
#define NB 1000
float *myarray;
myarray = malloc (sizeof(float) * NB);
...
myarray[5] = sqrt(3);
...
```

Reminder: allocating memory on the host

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#define NB 1000 vector of 1000 elements
float *myarray;
myarray = malloc (sizeof(float) * NB);
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myarray[5] = sqrt(3);
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Reminder: allocating memory on the host

```
#define NB 1000
float *myarray; start address is myarray
myarray = malloc (sizeof(float) * NB);
...
myarray[5] = sqrt(3);
...
```

```
#define NB 1000
float *myarray;
cudaMalloc (&myarray, sizeof(float) * NB);
...
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```
#define NB 1000
float *myarray;
cudaMalloc (&myarray, sizeof(float) * NB);
floats have same size on CPU and GPU
...
cudaMalloc (myarray, sizeof(float)) cannot work!
```

```
#define NB 1000
float *myarray;
cudaMalloc (&myarray, sizeof(float) * NB);
...
```

Note that a pointer to a float on the CPU and a pointer to a float on the GPU have same type. The compiler is **unable** to know whether a given pointer is CPU or GPU related... This is bug prone!

CPU and GPU pointers

Dereferencing = using what is pointed to by the address represented by a pointer.

```
float *ptr;
...
*ptr = 3; pointer is dereferenced
ptr[1] = 4;
```

- Dereferencing a host pointer on the GPU will crash.
- Dereferencing a device pointer on the host will crash.

So beware! You may use a notation that conveys information about the destination of the pointer (host / device)



Example of notation

So beware! You may use a notation that conveys information about the destination of the pointer (host / device)
For instance:

```
float *ptr; pointer to host memory space
versus
```

```
float *gpu_ptr; pointer to video RAM

It is entirely up to you.
```

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Importance of alignment

As we shall see later, *data alignment* is extremely important on the device memory.

What is memory alignment? Assume that the data travels by chunks of four bytes on the bus.



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Assume that we have a four byte array that starts at 4: it is misaligned: reading this array (e.g. to the cache) requires two transactions.

cudaMalloc and alignment

The function cudaMalloc () always takes care of alignment (the start address is properly aligned).

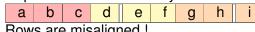
What happens if we have a two-dimensional array?

Bad news: multi-dimensional arrays do not exist in C. You want to consider a 1D array that contains all elements instead.

Intended array

а	b	С	
d	е	f	
g	h	i	

Representation in memory

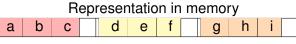


Rows are misaligned!

Alignment for two-dimensional arrays

A better memory scheme would be the following:

Intended array							
а	b	С		а			
d	е	f					
	l_	•					



Solution

We pad the array with extra cells (here in white), whose content is never used, so that each rows starts at an aligned address. cudaMallocPitch () does that automatically for you.

Assume you want to reserve memory for an array of w by h elements (say floats)

```
float *gpu_ptr;
size_t w, h, pitch;
cudaMallocPitch (&gpu_ptr, &pitch, w*sizeof(float), h);
```

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pointer to pitch

Assume you want to reserve memory for an array of w by h elements (say floats)

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width in bytes

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Freeing device memory upon completion

In the same manner as you must free memory mallocated on the host, you must free device memory (video RAM) allocated on the device

```
cudaMalloc (&myarray, sizeof(float) * NB);
...
cudaFree (myarray);
```

Note that once that is done, the data referred to by myarray is no longer accessible, but it persists on the device and may be retrieved by others. You may want to overwrite it prior to calling cudaFree ().

- Until a few years ago, there was no possibility of input/outputs on the device
- There is now a printf () equivalent... One can now do some printf () debugging.
- The endpoint of a computation must be accessible (commonly written on disk).
- Data must travel as fast as possible between device and host.
- communication device ⇒ host, then written by host in a standard manner.



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Usage of cudaMemcpy ()

The function <code>cudaMemcpy</code> () is used to transfer chunks of memory from host to device, or vice-versa. It is a synchronous function: it returns only upon transfer completion.

E.g. transfer NB floats from device to host

Remember that the compiler has no idea that ptr is on the host and gpu_ptr is on the device. You have to help it.

Other example: assume you have initialized an array on the host, and want to perform calculations on the device.

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A bits of hands on practice

Bandwidth measurement

Write a piece of code that evaluates the data transfer rate between host and device.

We need:

- big arrays allocated on host and device
- a chronometer

Use the file bandwidth.cu provided in this lecture. Look up comments that say *UP TO YOU* and add the instructions requested.

Then compile it and run it:

```
nvcc bandwidth.cu
./a.out
```



Playing with bandwidth.cu

The original executable gives the host \Rightarrow device bandwidth. Modify it to check also the device \Rightarrow host and device \Rightarrow device bandwidths.

Conclusions

Compare with other groups' results. How does the bandwidth vary for low-end vs high-end GPUs?

Outline

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- Normally, the memory allocated on the host (via malloc) is "mobile", i.e. it can go to the swap, then go to other segments of the host RAM. It is called pageable memory.
- CUDA provides the possibility to allocate non-pageable memory (pinned memory). It resides in a given part of the RAM and does not move until it is freed. That speeds up transfers with the host.
- Only root can normally do that. Beware of the use of excessive non-pageable memory ⇒ risk of decreasing performance of the host.
- Syntax : same as cudaMalloc ():



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Freeing non-pageable memory on host

This memory MUST be freed after use, obviously.

- Can you do that with free ()?
- Can you do that with CudaFree ()?

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None of these !... It must be freed with CudaFreeHost ()

bandwidth with pinned memory

Modify the bandwidth.cu file in order to allocate (and free!) non-pageable memory.

Check how the bandwidth is improved

Conclusions?

SDK's bandwidthTest

A more sophisticated implementation of this bandwidth test already exists in NVIDIA's SDK.

Check the following, and compare with your own results:

- ./bandwidthTest
- ./bandwidthTest -memory=pinned

More information can be obtained with

./bandwidthTest -help

- We have not written a kernel yet
- Everything we have written runs on the host
- We have only interacted with the GPU for
 - allocations in the global memory
 - transfers involving the global memory
- That's all!

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we have not programmed the GPU yet.

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Programming our first kernel is the purpose of the next lecture.