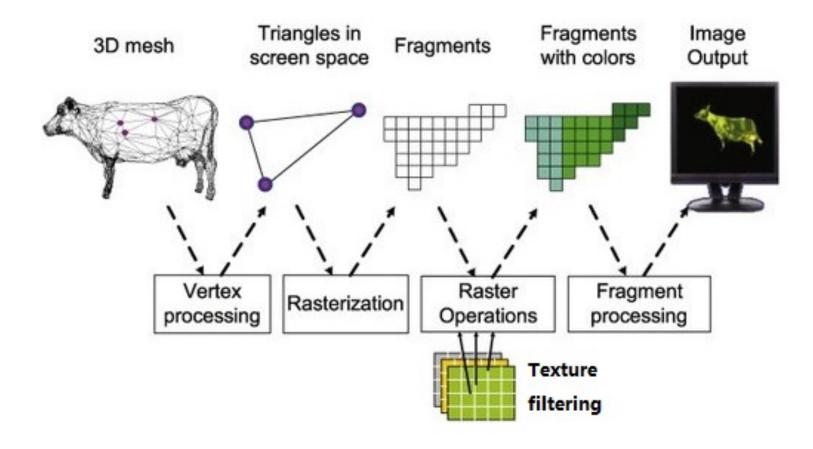
# Fundamentals of CUDA 1

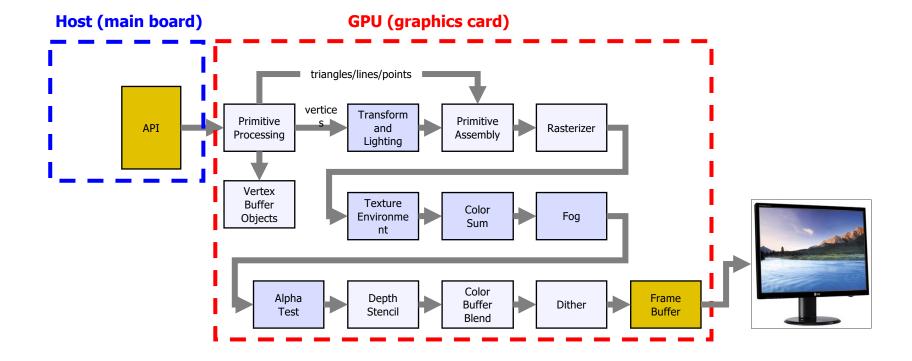
Prof. Seokin Hong

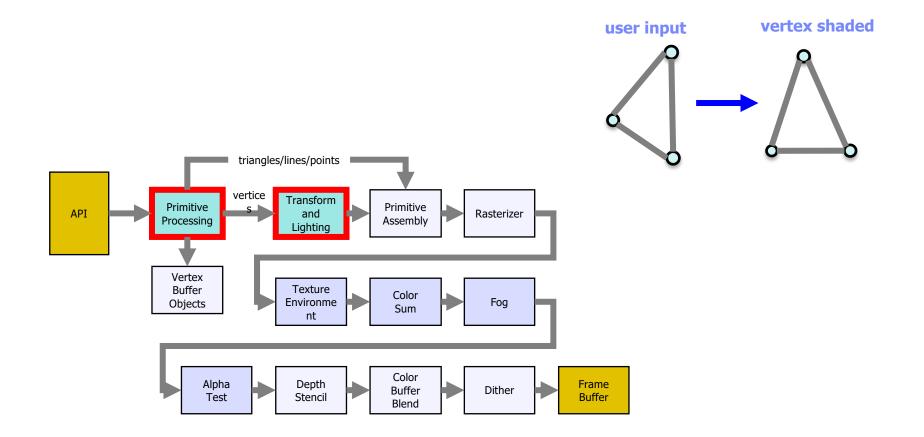
## **Agenda**

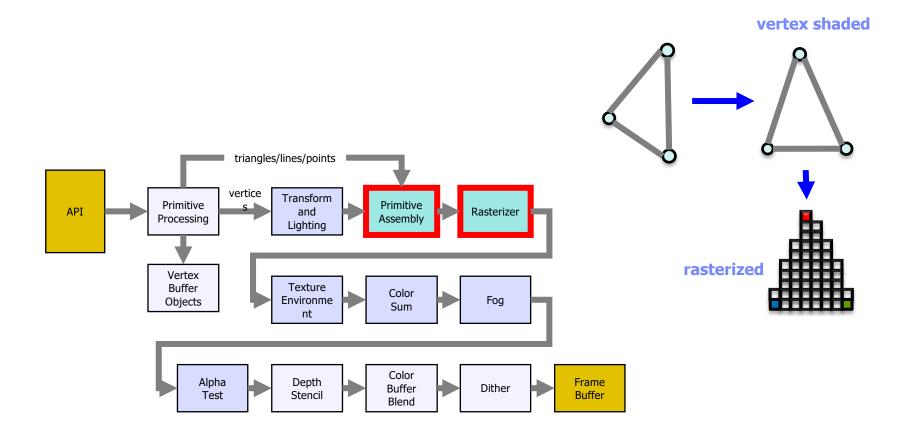
- History
- What is CUDA?
- Device Global Memory and Data Transfer
- Error Checking
- A Vector Addition Kernel
- Kernel Functions and Threading
- Kernel Launch

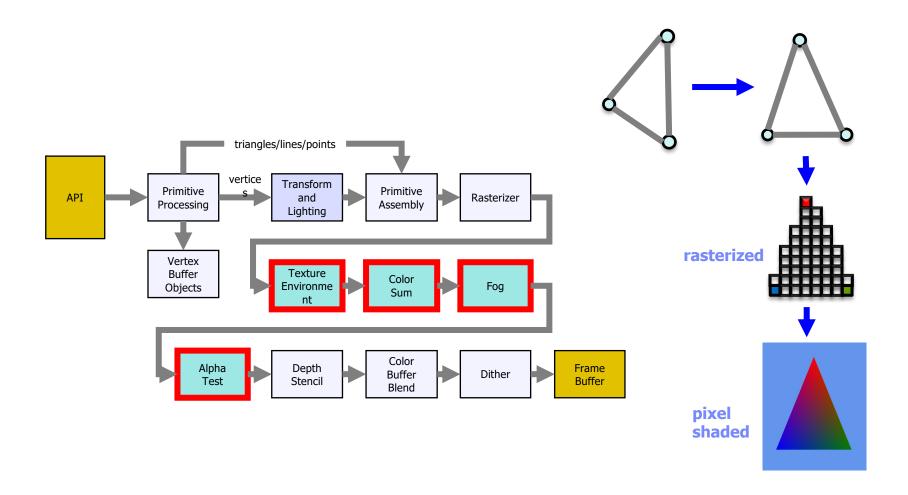








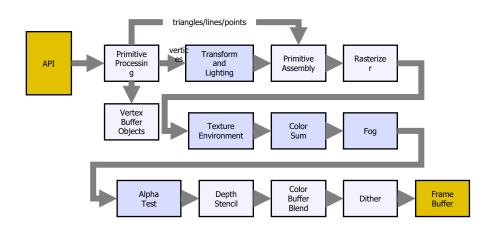




# The Graphics Pipeline – 1st Gen.

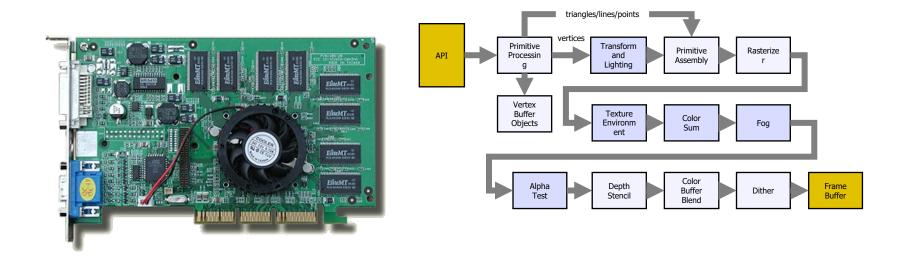
- One chip/board per stage
- Fixed data flow through pipeline





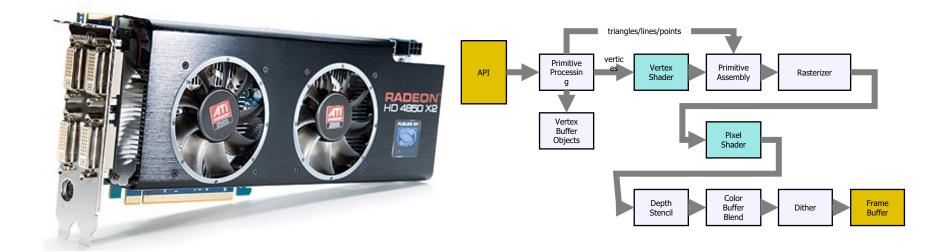
## The Graphics Pipeline – 2nd Gen.

- Everything fixed function,
   with a certain number of modes
- Number of modes for each stage grew over time
- Hard to optimize HW
- Developers always wanted more flexibility



# The Graphics Pipeline – 3rd Gen.

- Vertex & pixel processing became programmable
- GPU architecture increasingly centers around `shader' execution



### **Before CUDA**

- Use the GPU for general-purpose
- Computing by casting problem as graphics
  - Turn data into images ("texture maps")
  - Turn algorithms into image synthesis ("rending passes")

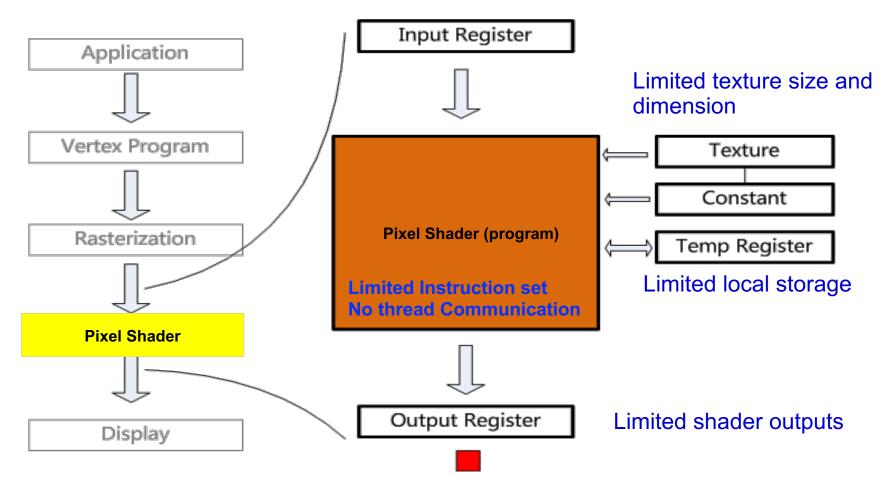
#### Drawback:

- Tough learning curve
- Potentially high overhead of graphics API
- Highly constrained memory layout & access model

### **Before CUDA**

What's wrong with the old GPGPU programming model 1

#### APIs are specific to Graphics

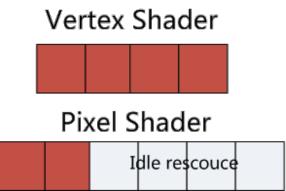


### **Before CUDA**

■ What's wrong with the old GPGPU programming model 2











### What is CUDA?

### What is CUDA?: Compute Unified Device Architecture

- Parallel computing platform and application programming interface (API) model
- A powerful parallel programming model for issuing and managing computations on the GPU <u>without mapping them to a graphics API</u>

### Targeted Software stack

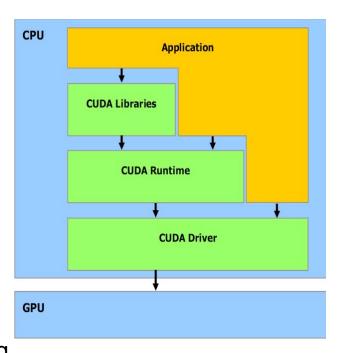
Library, Runtime, Driver

### Advantages

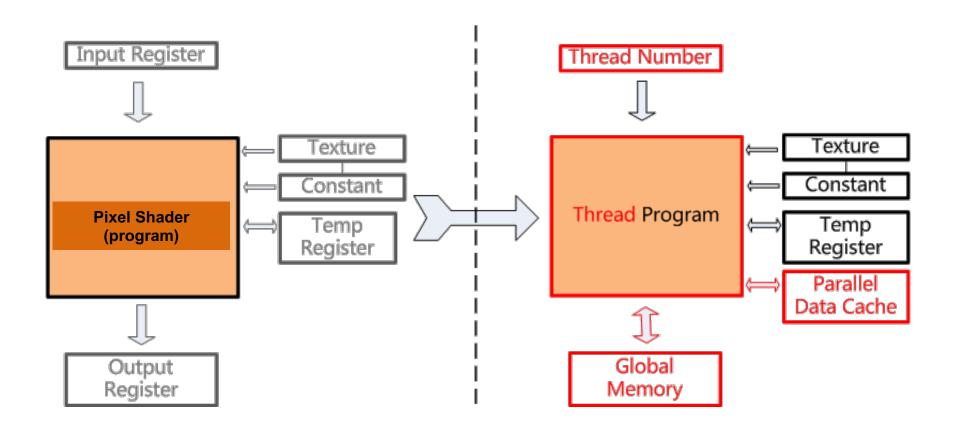
- SW: program the GPU in C
  - Scalable data parallel execution/memory model
  - C with minimal yet powerful extensions
- HW: fully general data-parallel architecture

#### Features

- Heterogenous mixed serial-parallel programming
- Scalable hierarchical thread execution model
- Accessible minimal but expressive changes to C



### What is CUDA?

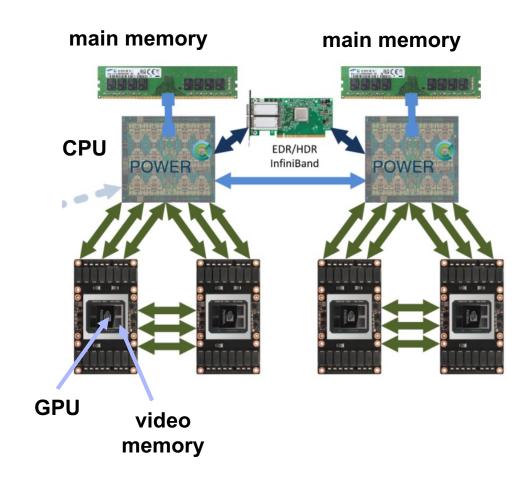


## **Review: Heterogeneous Computing**

- Use more than one kind of processor or cores
  - CPUs for sequential parts
  - GPUs for parallel parts

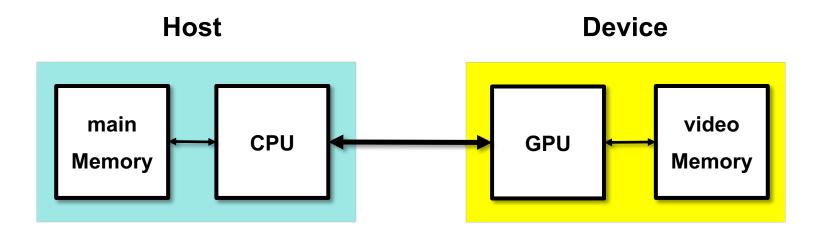






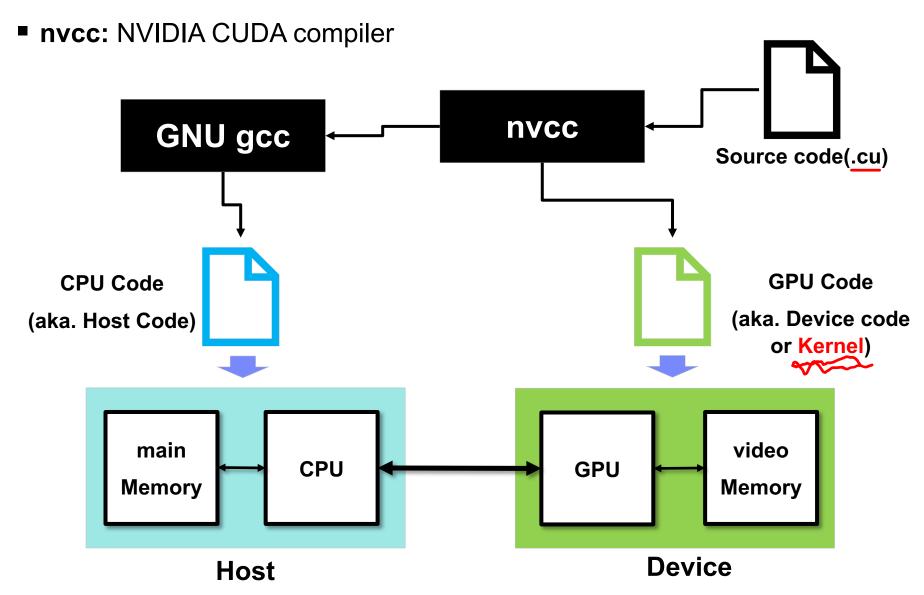
## **Simple CUDA Model**

- Host : CPU + main memory (host memory)
- Device : GPU + video memory (device memory)



## **Simple CUDA Model**

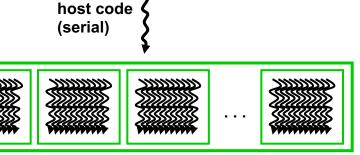
■ GNU gcc : linux c compiler



## **CUDA Program Execution Scenario**

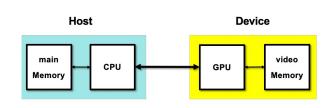
- Integrated "host+device" app C program
  - Serial or modestly parallel parts in host code
  - Highly parallel parts in device code

Device code (parallel)



#### Execution Scenario

- Step 1: host code
  - Serial execution: read data
  - Prepare parallel execution
  - Copy data from host memory to device memory
- Step 2: device code (kernel)
  - Parallel processing
  - Read/write data from device memory to device memory
- Step 3: host code
  - Copy data from device memory to host memory
  - Serial execution: print data

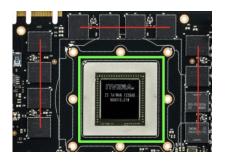


host code (serial)

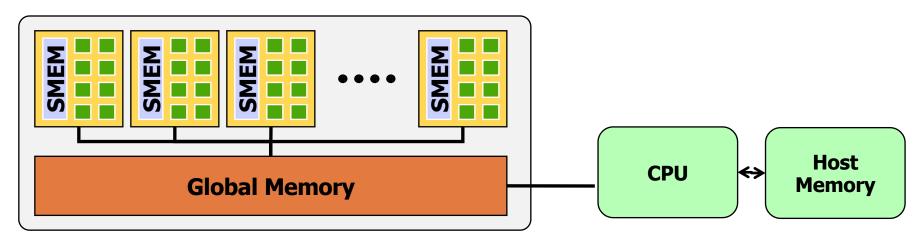
## **Device Global Memory and Data Transfer**

## **CUDA** program uses CUDA memory

- GPU cores share the "global memory" (device memory)
  - o DRAM (e.g., GDDR, HBM) is used as global memory
- To execute a kernel on a device,
  - allocate global memory on the device
  - transfer data from the host memory to allocated device memory
  - transfer result data from the device memory back to the host memory
  - o release global memory



Red lines are global memory



## **Memory Spaces**

### CPU and GPU have separate memory spaces

- Data is moved across data bus
- Use functions to allocate/set/copy memory on GPU
- Very similar to corresponding C functions

- Pointers are just addresses
  - Use pointer to access CPU and GPU memory
  - Can't tell from the pointer value whether the address is on CPU or GPU
  - Dereferencing CPU pointer on GPU will likely crash
  - Same for vice versa

### **CPU Memory Allocation / Release**

Host (CPU) manages host (CPU) memory:

```
o void* malloc (size_t nbytes)
void* memset (void* pointer, int value, size_t count)
void free (void* pointer)
int n = 1024;
int nbytes = 1024*sizeof(int);
int* ptr = 0;
ptr = malloc( nbytes );
memset( ptr, 0, nbytes);
free( ptr );
```

## **GPU Memory Allocation / Release**

Host (CPU) manages device (GPU) memory:

```
    cudaMalloc (void** pointer, size_t nbytes)

    cudaMemset (void* pointer, int value, size_t count)

cudaFree (void* pointer)
int n = 1024;
int nbytes = 1024*sizeof(int);
int* dev a = 0;
cudaMalloc( (void**)&dev_a, nbytes );
cudaMemset( dev_a, 0, nbytes);
cudaFree(dev a);
```

### **CUDA** function rules

- Every library function starts with "cuda"
- Most of them returns error code (or cudaSuccess).

```
cudaError_t cudaMalloc(void** devPtr, size_t size);
```

- cudaError\_t cudaFree(void\* devPtr);
- cudaError\_t cudaMemcpy(void\* dst, const void\* src, size\_t size);

### Example:

```
o if (cudaMalloc(&devPtr, SIZE) != cudaSuccess) {
     exit(1);
}
```

### **CUDA Malloc**

- cudaError\_t cudaMalloc(void\*\* devPtr, size\_t nbytes);
  - allocates nbytes bytes of linear memory on the device
  - The start address is stored into "devPtr"
  - The memory is not cleared.
  - returns cudaSuccess or cudaErrorMemoryAllocation
- cudaError\_t cudaFree( void\* devPtr );
  - frees the memory space pointed by devPtr
  - if devPtr == 0, no operation
  - returns cudaSuccess or cudaErrorInvalidDevicePointer

### **CUDA** mem set

- cudaError\_t cudaMemset(void\* devPtr, int value, size\_t nbytes);
  - o fills the first *nbytes* byte of the memory area pointed by *devPtr* with the *value*
  - o returns cudaSuccess, cudaErrorInvalidValue, cudaErrorInvalidDevicePointer

## **Data Copy**

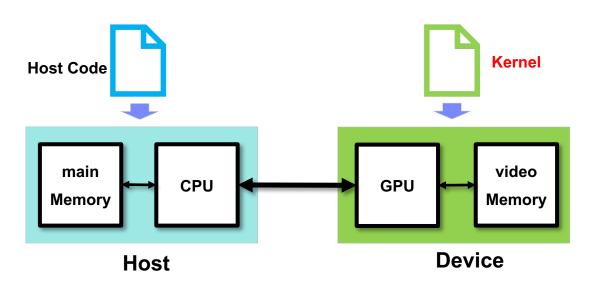
- returns after the copy is complete
- blocks CPU thread until all bytes have been copied
- doesn't start copying until previous CUDA calls complete

### enum cudaMemcpyKind

- cudaMemcpyHostToDevice
- cudaMemcpyDeviceToHost
- cudaMemcpyDeviceToDevice
- cudaMemcpyHostToHost

## **Data Copy**

- host → host : memcpy (in C/C++)
- host → device, device → device, device → host : cudaMemcpy (CUDA)



## **Example: Host-Device Mem copy**

### step 1.

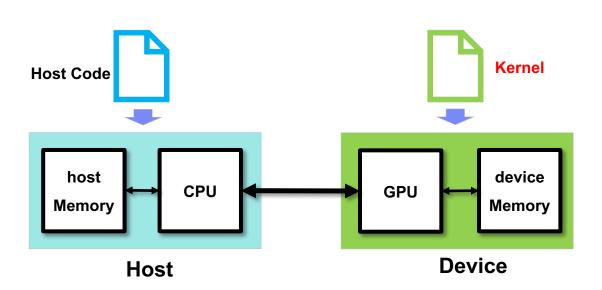
- make a block of data
- print out the source data

### ■ step 2.

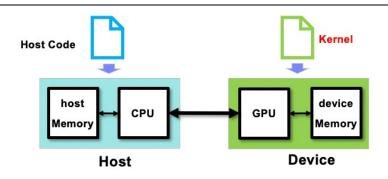
- copy from host memory to device memory
- copy from device memory to device memory
- copy from device memory to host memory

### step 3.

print out the result



## Code: cuda memcpy (1/4)

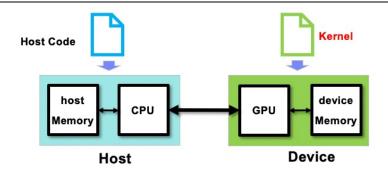


```
#include <iostream>
int main(void) {
  // host-side data
  const int SIZE = 5;
  const int a[SIZE] = { 1, 2, 3, 4, 5 }; // source data
  int b[SIZE] = { 0, 0, 0, 0, 0 }; // final destination
  // print source
  printf("a = \{\%d,\%d,\%d,\%d,\%d\}\n", a[0], a[1], a[2], a[3], a[4]);
```

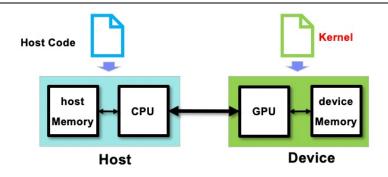
## Code: cuda memcpy (2/4)

```
Kernel
                                                    Host Code
// device-side data
                                                         host
                                                                                     device
int* dev a = 0;
                                                        Memory
                                                                                     Memory
int^* dev_b = 0;
                                                             Host
                                                                                  Device
// allocate device memory
cudaMalloc((void**)&dev a, SIZE * sizeof(int));
cudaMalloc((void**)&dev b, SIZE * sizeof(int));
// copy from host to device
cudaMemcpy(dev_a, a, SIZE * sizeof(int), cudaMemcpyHostToDevice);
```

## Code: cuda memcpy (3/4)



## Code: cuda memcpy (4/4)



```
// free device memory
cudaFree(dev_a);
cudaFree(dev_b);
// print the result
printf("b = \{\%d,\%d,\%d,\%d,\%d\}\n", b[0], b[1], b[2], b[3], b[4]);
// done
return 0;
```

#### **Execution Result**

- Compile the source code
  - o nvcc memcpy.cu -o ./memcpy

executing ./memcpy

$$a = \{1,2,3,4,5\}$$

$$b = \{1,2,3,4,5\}$$



## **Error Checking and Handling in CUDA**

- It is important for a program to check and handle errors
- CUDA API functions return flags that indicate whether an error has occurred

```
Most of them returns error code (or cudaSuccess).
```

```
    cudaError_t cudaMalloc(void** devPtr, size_t size);
    cudaError_t cudaFree(void* devPtr);
    cudaError_t cudaMemcpy(void* dst, const void* src, size_t size);
```

#### Example:

```
o if (cudaMalloc(&devPtr, SIZE) != cudaSuccess) {
     exit(1);
}
```

#### cudaError\_t : data type

- typedef enum cudaError cudaError\_t
- possible values:
  - cudaSuccess, cudaErrorMissingConfiguration, cudaErrorMemoryAllocation, cudaErrorInitializationError, cudaErrorLaunchFailure, cudaErrorLaunchTimeout, cudaErrorLaunchOutOfResources, cudaErrorInvalidDeviceFunction, cudaErrorInvalidConfiguration, cudaErrorInvalidDevice, cudaErrorInvalidValue, cudaErrorInvalidPitchValue, cudaErrorInvalidSymbol, cudaErrorUnmapBufferObjectFailed, cudaErrorInvalidHostPointer, cudaErrorInvalidDevicePointer, cudaErrorInvalidTexture, cudaErrorInvalidTextureBinding, cudaErrorInvalidChannelDescriptor, cudaErrorInvalidMemcpyDirection, cudaErrorInvalidFilterSetting, cudaErrorInvalidNormSetting, cudaErrorUnknown, cudaErrorNotYetImplemented, cudaErrorInvalidResourceHandle, cudaErrorInsufficientDriver, cudaErrorSetOnActiveProcess, cudaErrorStartupFailure, cudaErrorApiFailureBase

## cudaGetErrorName( err )

- const char\* cudaGetErrorName( cudaError\_t err )
  - err : error code to convert to string
  - o returns:
    - char\* to a NULL-terminated string
    - NULL if the error code is not valid
- - o shows:

cudaErrorMemoryAllocation cudaErrorInvalidValue

### cudaGetErrorString( err )

- const char\* cudaGetErrorString( cudaError\_t err )
  - err : error code to convert to string
  - o returns:
    - char\* to a NULL-terminated string
    - NULL if the error code is not valid
- - o shows:

out of memory invalid argument

## cudaGetLastError( void )

- cudaError\_t cudaGetLastError(void)
  - returns the last error due to CUDA runtime calls in the same host thread
  - and resets it to cudaSuccess
  - So, if no CUDA error since the last call, it returns cudaSuccess
  - For multiple errors, it contains the last error only.

- cudaError t cudaPeekAtLastError(void)
  - returns the last error due to CUDA runtime calls in the same host thread
  - Note that this call does NOT reset
  - So, the last error code is still available

## A simple CUDA error check code

```
cudaMemcpy( ... );
cudaError_t e = cudaGetLastError();
if (e != cudaSuccess) {
        printf("cuda failure %s:%d: '%s'\n",
          __FILE___, __LINE___,
          cudaGetErrorString(e) );
        exit(0);
```

# A simple CUDA error check macro

```
#define cudaCheckError() do { \
        cudaError_t e = cudaGetLastError(); \
        if (e != cudaSuccess) { \
                printf("cuda failure %s:%d: '%s'\n", \
                   ___FILE___, __LINE___, \
                   cudaGetErrorString(e) ); \
                exit(0); \
        } \
} while (0)
```

# **Example**

code segment

```
// allocate device memory
cudaMalloc((void**)&dev_a, sizeof(int));
cudaMalloc((void**)&dev_b, sizeof(int));
cudaMalloc((void**)&dev_c, sizeof(int));
cudaCheckError( );
```

#### More advanced macro

```
#ifdef DEBUG // debug mode
#define CUDA_CHECK(x)
                               do {\
          (x); \
          cudaError_t e = cudaGetLastError(); \
          if (cudaSuccess != e) { \
                     printf("cuda failure %s at %s:%d\n", \
                        cudaGetErrorString(e), \
                        __FILE__, __LINE___); \
                     exit(1); \
          } \
} while (0)
#else
#define CUDA CHECK(x)
                                         // release mode
                           (x)
#endif
```

### error\_check.cu

```
#include <iostream>
#ifdef DEBUG
#define CUDA_CHECK(x) do \{\
    (x); \
    cudaError_t e = cudaGetLastError(); \
    if (cudaSuccess != e) { \
       printf("cuda failure \"%s\" at %s:%d\n", \
           cudaGetErrorString(e), \
           __FILE__, __LINE__); \
       exit(1); \
    } \
  } while (0)
#else
#define CUDA_CHECK(x) (x)
#endif
```

#### error\_check.cu

```
// main program for the CPU
int main(void) {
   // host-side data
   const int SIZE = 5;
   const int a[SIZE] = \{ 1, 2, 3, 4, 5 \};
   int b[SIZE] = \{ 0, 0, 0, 0, 0, 0 \};
  // print source
   printf("a = \{\%d,\%d,\%d,\%d,\%d\}\n", a[0], a[1], a[2], a[3], a[4]);
  // device-side data
   int *dev a = 0;
   int *dev b = 0:
  // allocate device memory
   CUDA CHECK( cudaMalloc((void**)&dev_a, SIZE * sizeof(int)) );
   CUDA CHECK( cudaMalloc((void**)&dev b, SIZE * sizeof(int)) ):
   // copy from host to device
   CUDA CHECK( cudaMemcpy(dev a, a, SIZE * sizeof(int), cudaMemcpyDeviceToDevice)); // BOMB here!
  // copy from device to device
   CUDA CHECK( cudaMemcpy(dev b, dev a, SIZE * sizeof(int), cudaMemcpyDeviceToDevice) );
   // copy from device to host
   CUDA CHECK( cudaMemcpy(b, dev b, SIZE * sizeof(int), cudaMemcpyDeviceToHost) );
   // free device memory
   CUDA CHECK( cudaFree(dev a) );
   CUDA CHECK( cudaFree(dev b) );
   // print the result
   printf("b = \{\%d,\%d,\%d,\%d,\%d\}\n", b[0], b[1], b[2], b[3], b[4]);
  // done
   return 0;
```

#### **Execution Result**

- Compile the source code
  - o nvcc error\_check.cu -DDEBUG -o ./error\_check
  - nvcc error\_check.cu -o ./error\_check

executing ./error\_check

$$a = \{1,2,3,4,5\}$$

$$b = \{.....\}$$

## **Agenda**

- What is CUDA?
- Device Global Memory and Data Transfer
- Error Checking
- A Vector Addition Kernel
- Kernel Functions and Threading
- Kernel Launch

#### **CUDA Resources**

#### CUDA API reference:

- http://docs.nvidia.com/cuda/index.html
- http://docs.nvidia.com/cuda/cuda-runtime-api/index.html

#### CUDA course:

- https://developer.nvidia.com/cuda-education-training
- https://developer.nvidia.com/cuda-training

