# CS 677: Parallel Programming for Many-core Processors Lecture 9

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## Outline

- Pinned Memory
- Streams
- Thrust

## Pinned Memory

- Page-locked or pinned memory transfers attain the highest bandwidth between host and device
  - Ensures that host buffer does not get moved to virtual memory
- Allocated using the cudaMallocHost()
- Pinned memory should not be overused
  - Excessive use can reduce overall system performance
  - How much is too much is difficult to tell in advance

## Asynchronous Transfers and Overlapping Transfers with Computation

- Data transfers between host and device using cudaMemcpy() are blocking transfers
  - Control is returned to the host thread only after the data transfer is complete.
- The cudaMemcpyAsync() function is a nonblocking variant of cudaMemcpy()
  - Unlike cudaMemcpy() the asynchronous transfer version requires pinned host memory

## Asynchronous Transfers and Overlapping Transfers with Computation

```
cudaMemcpyAsync(a_d, a_h, size,
  cudaMemcpyHostToDevice, stream);
kernel<<<grid, block>>>(a_d);
cpuFunction();
```

- Memory transfer and device execution are performed in parallel with host execution
- Last argument of cudaMemcpyAsync() specifies stream
  - 0 is the default only nonzero streams are asynchronous (more details soon)
  - Kernel does not begin execution until memory transfer is complete

#### **CUDA Streams**

Patrick Cozzi
University of Pennsylvania
CIS 565 - Spring 2011

Steve Rennich NVIDIA

- Stream: Sequence of commands that execute in order
- Streams may execute their commands out-oforder or concurrently with respect to other streams

Stream A

Command 0

Command 1

Command 2

Stream B

Command 0

Command 1

Command 2

Stream A Stream B

Command 0 Command 0

Command 1 Command 1

Command 2 Command 2

Is this a possible order?

Command 0
Command 1
Command 2
Command 0
Command 1
Command 1

Stream A Stream B

Command 0 Command 0

Command 1 Command 1

Command 2 Command 2

Is this a possible order?

Time

Command 0

Command 0

Command 1

Command 1

Command 2

Command 2

Stream A Stream B

Command 0 Command 0

Command 1 Command 1

Command 2 Command 2

Is this a possible order?

Command 0
Command 2
Command 2
Command 1
Command 1

Stream A Stream B

Command 0 Command 0

Command 1 Command 1

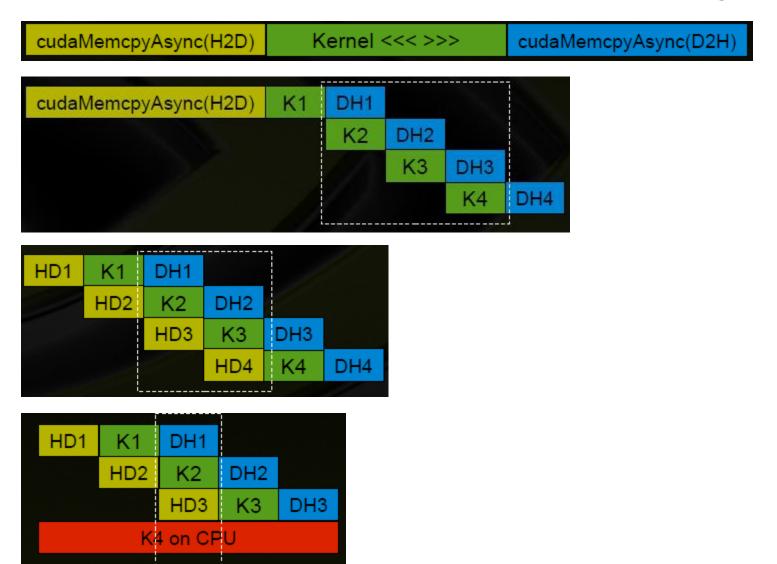
Command 2 Command 2

Is this a possible order?

Command 0 Command 0
Command 1 Command 1
Command 2 Command 2

- In CUDA, what commands go in a stream?
  - Kernel launches
  - Host device memory transfers

## **Amount of Concurrency**



# Default Stream (Stream '0')

- Stream used when no stream is specified
- Completely synchronous w.r.t. host and device
  - As if cudaDeviceSynchronize() inserted before and after every CUDA operation
- Exceptions asynchronous w.r.t. host
  - Kernel launches in the default stream
  - cudaMemcpy\*Async
  - cudaMemset\*Async
  - cudaMemcpy within the same device
  - H2D cudaMemcpy of 64kB or less

## Requirements for Concurrency

- CUDA operations must be in different, non-0, streams
- cudaMemcpyAsync with host from 'pinned' memory
  - Page-locked memory
  - Allocated using cudaMallocHost() or cudaHostAlloc() (\*)
- Sufficient resources must be available
  - cudaMemcpyAsyncs in different directions
  - Device resources (SMEM, registers, blocks, etc.)

<sup>\*</sup> Both commands are roughly equivalent, but not in all versions of CUDA

- Code Example
  - 1. Create two streams
  - 2. Each stream:
    - 1. Copy page-locked memory to device
    - 2. Launch kernel
    - 3. Copy memory back to host
  - 3. Destroy streams

# Stream Example (Step 1 of 3)

```
cudaStream t stream[2];
for (int i = 0; i < 2; ++i)
    cudaStreamCreate(&stream[i]);
                         Create two streams
float *hostPtr;
cudaMallocHost(&hostPtr, 2 * size);
```

# Stream Example (Step 1 of 3)

```
cudaStream t stream[2];
for (int i = 0; i < 2; ++i)
    cudaStreamCreate(&stream[i]);
float *hostPtr;
cudaMallocHost(&hostPtr, 2 * size);
```

Allocate two buffers in page-locked memory

## Stream Example (Step 2 of 3)

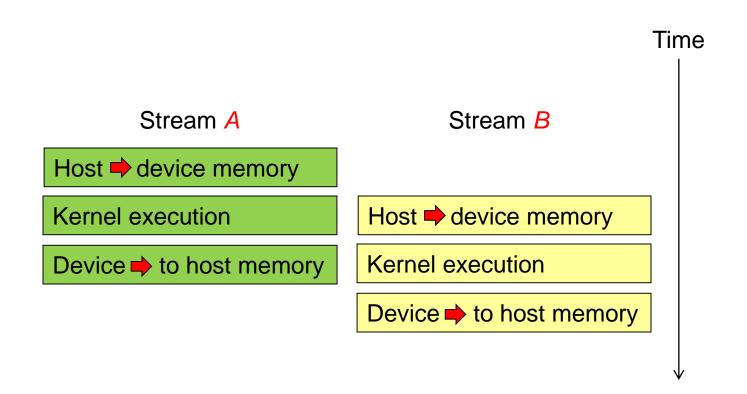
```
for (int i = 0; i < 2; ++i)
  cudaMemcpyAsync(/* ... */,
    cudaMemcpyHostToDevice, stream[i]);
  kernel << 100, 512, 0, stream[i]>>>
    (/* ... */);
  cudaMemcpyAsync(/* ... */,
    cudaMemcpyDeviceToHost, stream[i]);
```

## Stream Example (Step 3 of 3)

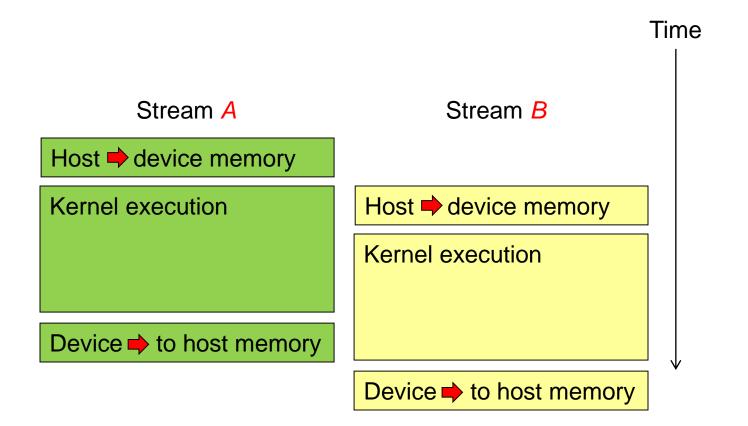
```
for (int i = 0; i < 2; ++i)
{
    // Blocks until commands complete
    cudaStreamDestroy(stream[i]);
}</pre>
```

- Assume compute capability 1.1 and above:
  - Overlap of data transfer and kernel execution
  - Concurrent kernel execution
  - Concurrent data transfer
- How can the streams overlap?

Can we have more overlap than this?



Can we have this?



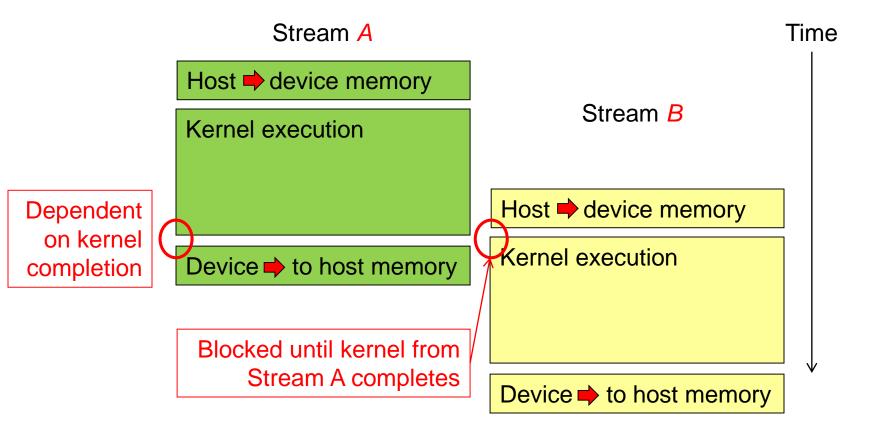
- Implicit Synchronization
  - An operation that requires a dependency check to see if a kernel finished executing:
    - Blocks all kernel launches from any stream until the checked kernel is finished

 cudaStreamQuery() can be used to test if a stream has completed all operations

## Implicit Synchronization

- These operations implicitly synchronize all other CUDA operations
  - Page-locked memory allocation
    - cudaMallocHost
    - cudaHostAlloc
  - Device memory allocation
    - cudaMalloc
  - Non-Async version of memory operations
    - cudaMemcpy\* (no Async suffix)
    - cudaMemset\* (no Async suffix)
  - Change to L1/shared memory configuration
    - cudaDeviceSetCacheConfig

Can we have this?



- Performance Advice
  - Issue all independent commands before dependent ones
  - Delay synchronization (implicit or explicit) as long as possible

 Rewrite this to allow concurrent kernel execution

```
for (int i = 0; i < 2; ++i)
{
    cudaMemcpyAsync(/* ... */, stream[i]);
    kernel <<< /*... */ stream[i]>>>();
    cudaMemcpyAsync(/* ... */, stream[i]);
}
```

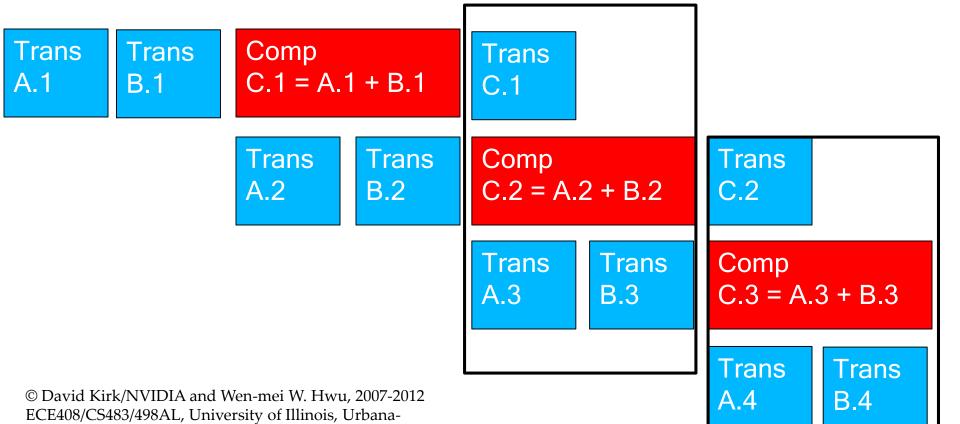
```
for (int i = 0; i < 2; ++i) // to device
cudaMemcpyAsync(/* ... */, stream[i]);
for (int i = 0; i < 2; ++i)
  kernel <<< /*... */ stream[i]>>>();
for (int i = 0; i < 2; ++i) // to host
  cudaMemcpyAsync(/* ... */, stream[i]);
```

# Overlapped (Pipelined) Timing

Divide large vectors into segments

Champaign

Overlap transfer and compute of adjacent segments



## **Explicit Syncrhonization**

- cudaDeviceSynchronize()
  - Blocks until commands in all streams finish
  - cudaThreadSyncrhonize() has been deprecated
- cudaStreamSynchronize(streamid)
  - Blocks until commands in a specific stream finish

## Synchronization using Events

- Create specific 'Events', within streams, to use for synchronization
- cudaEventRecord ( event, streamid )
- cudaEventSynchronize ( event )
- cudaStreamWaitEvent ( stream, event )
- cudaEventQuery ( event )

## **Explicit Synchronization Example**

```
cudaEvent t event;
cudaEventCreate (&event); // create event
// 1) H2D copy of new input
cudaMemcpyAsync ( d_in, in, size, H2D, stream1 );
cudaEventRecord (event, stream1);
                                                    // record event
// 2) D2H copy of previous result
cudaMemcpyAsync ( out, d_out, size, D2H, stream2 );
cudaStreamWaitEvent ( stream2, event );
                                                   // wait for event in stream1
kernel <<< , , , stream2 >>> ( d_in, d_out );
                                                  // 3) must wait for 1 and 2
asynchronousCPUmethod ( ... );
```

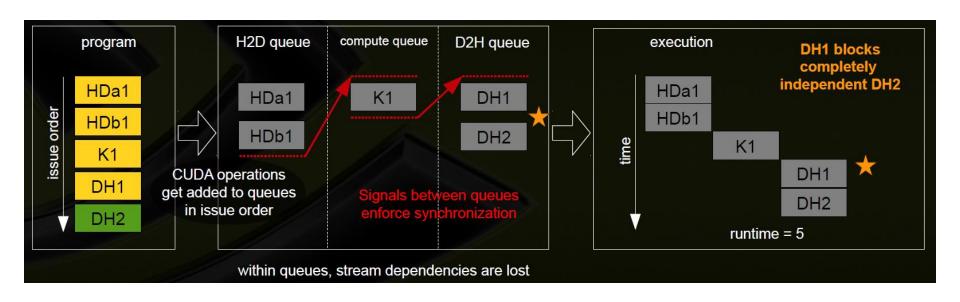
## Stream Scheduling

- Fermi hardware has 3 queues
  - 1 Compute Engine queue
  - 2 Copy Engine queues one for H2D and one for D2H
- CUDA operations are dispatched to HW in the sequence they were issued
  - Placed in the relevant queue
  - Stream dependencies between engine queues are maintained, but lost within an engine queue
- A CUDA operation is dispatched from the engine queue if:
  - Preceding calls in the same stream have completed,
  - Preceding calls in the same queue have been dispatched, and
  - Resources are available
- CUDA kernels may be executed concurrently if they are in different streams
  - Threadblocks for a given kernel are scheduled if all threadblocks for preceding kernels have been scheduled and there still are SM resources available
- Note that a blocked operation blocks all other operations in the queue, even in other streams

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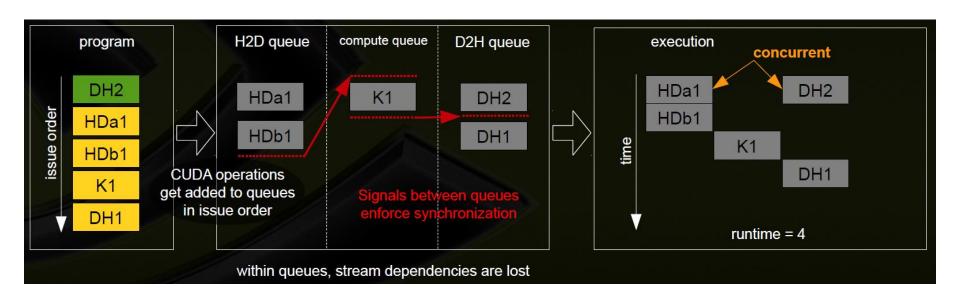
## Example - Blocked Queue

- Two streams, stream 1 is issued first
  - Stream 1: HDa1, HDb1, K1, DH1 (issued first)
  - Stream 2 : DH2 (completely independent of stream 1)



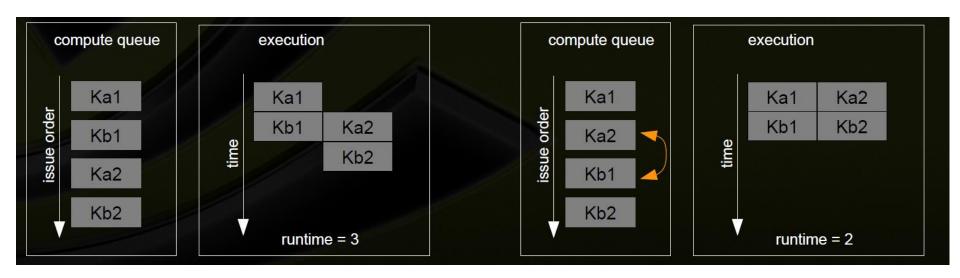
## Example - Blocked Queue

- Two streams, stream 1 is issued first
  - Stream 1 : HDa1, HDb1, K1, DH1
  - Stream 2 : DH2 (issued first)



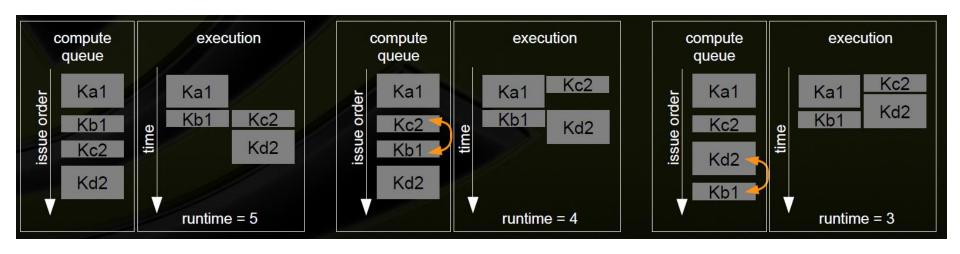
## Example - Blocked Kernel

- Two streams just issuing CUDA kernels
  - Stream 1 : Ka1, Kb1
  - Stream 2 : Ka2, Kb2
  - Kernels are similar size, fill ½ of the SM resources



# Example - Optimal Concurrency can Depend on Kernel Execution Time

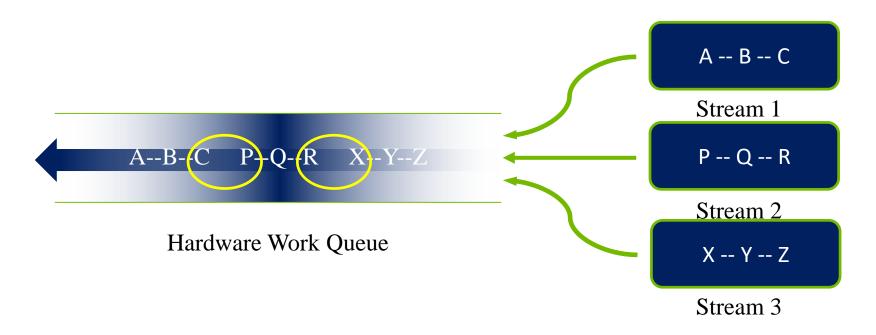
- Two streams just issuing CUDA kernels but kernels are different 'sizes'
  - Stream 1 : Ka1 {2}, Kb1 {1}
  - Stream 2 : Kc2 {1}, Kd2 {2}
  - Kernels fill ½ of the SM resources



## Hyper Queue

- Provide multiple real queues for each engine
- Allow much more concurrency by allowing some streams to make progress for an engine while others are blocked

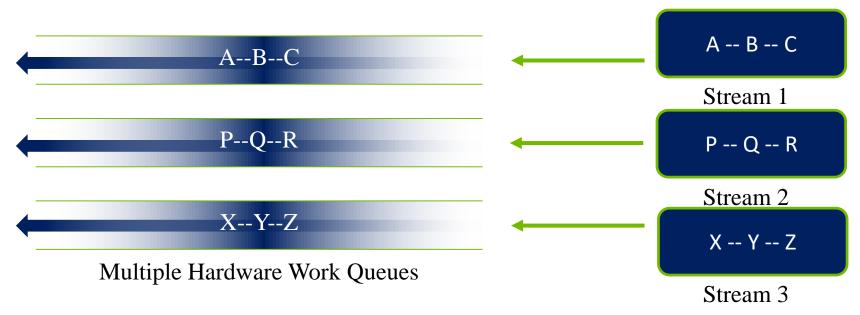
### Fermi (and older) Concurrency



### Fermi allows 16-way concurrency

- Up to 16 grids can run at once
- But CUDA streams multiplex into a single queue
- Overlap only at stream edges

### Improved Concurrency

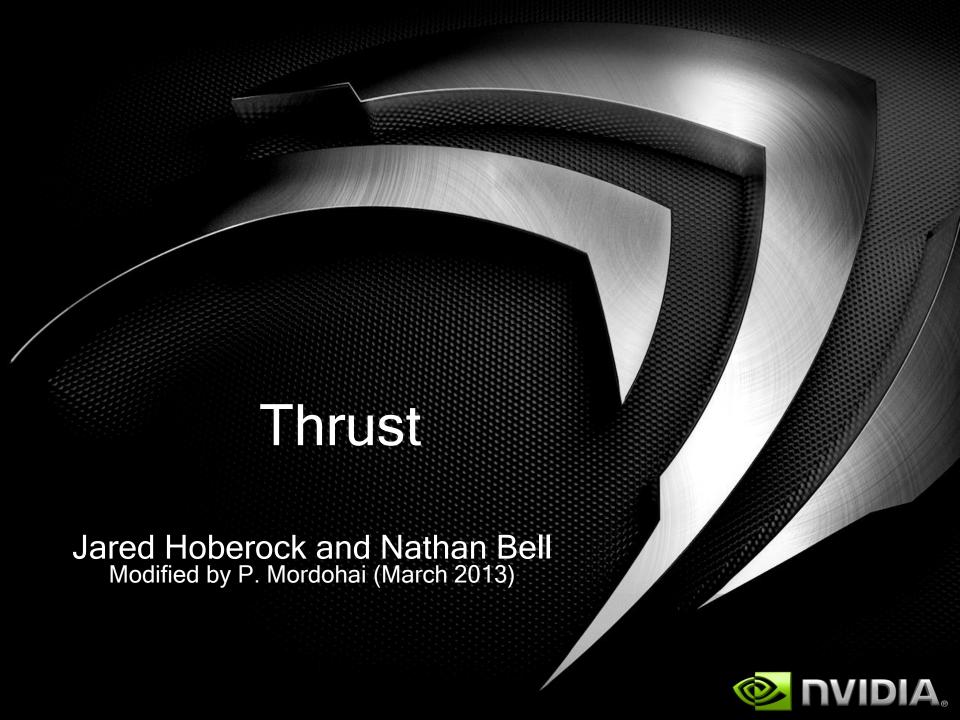


#### **Kepler allows 32-way concurrency**

- One work queue per stream
- Concurrency at full-stream level
- No inter-stream dependencies

#### Pascal also supports 32-way concurrency

- One work queue per stream
- Dynamic scheduling



## A Simple Example

```
#include <thrust/host vector.h>
#include <thrust/device vector.h>
#include <iostream>
int main(void)
    // H has storage for 4 integers
    thrust::host vector<int> H(4);
    // initialize individual elements
    H[0] = 14;
   H[1] = 20;
   H[2] = 38;
   H[3] = 46;
```

```
// H.size() returns the size of vector H
std::cout << "H has size " << H.size() << std::endl;</pre>
// print contents of H
for (int i = 0; i < H.size(); i++)
    std::cout << "H[" << i << "] = " << H[i] << std::endl;
// resize H
H.resize(2);
std::cout << "H now has size " << H.size() << std::endl;</pre>
// Copy host vector H to device vector D
thrust::device vector<int> D = H;
```

```
// elements of D can be modified
 D[0] = 99;
 D[1] = 88;
 // print contents of D
 for (int i = 0; i < D.size(); i++)
     std::cout << "D[" << i << "] = " << D[i] << std::endl;
 // H and D are automatically deleted when the function
returns
 return 0;
```

### Diving In

```
#include <thrust/host vector.h>
#include <thrust/device vector.h>
#include <thrust/sort.h>
int main (void)
    // generate 16M random numbers on the host
    thrust::host vector<int> h vec(1 << 24);</pre>
    thrust::generate(h vec.begin(), h vec.end(), rand);
    // transfer data to the device
    thrust::device vector<int> d vec = h vec;
    // sort data on the device
    thrust::sort(d vec.begin(), d vec.end());
    // transfer data back to host
    thrust::copy(d vec.begin(), d vec.end(), h vec.begin());
```

## Objectives

- Programmer productivity
  - Rapidly develop complex applications
  - Leverage parallel primitives
- Encourage generic programming
  - Don't reinvent the wheel
  - E.g. one reduction to rule them all
- High performance
  - With minimal programmer effort
- Interoperability
  - Integrates with CUDA C/C++ code

### What is Thrust?

- C++ template library for CUDA
  - Mimics Standard Template Library (STL)
- Containers

```
- thrust::host_vector<T>
- thrust::device vector<T>
```

### Algorithms

```
- thrust::sort()
- thrust::reduce()
- thrust::inclusive_scan()
- Etc.
```

### Namespaces

- C++ supports namespaces
  - Thrust uses thrust namespace

```
thrust::device_vectorthrust::copy
```

STL uses std namespace

```
std::vectorstd::list
```

Avoids collisions

```
- thrust::sort()
- std::sort()
```

For brevity

- using namespace thrust;

### Containers

- Make common operations concise and readable
  - Hides cudaMalloc, cudaMemcpy and cudaFree

```
// allocate host vector with two elements
thrust::host_vector<int> h_vec(2);

// copy host vector to device
thrust::device_vector<int> d_vec = h_vec;

// manipulate device values from the host
d_vec[0] = 13;
d_vec[1] = 27;

std::cout << "sum: " << d_vec[0] + d_vec[1] << std::endl;

// vector memory automatically released w/ free() or cudaFree()</pre>
```

### Containers

- Compatible with STL containers
  - Eases integration

```
- vector, list, map, ...
```

```
// list container on host
std::list<int> h_list;
h_list.push_back(13);
h_list.push_back(27);

// copy list to device vector
thrust::device_vector<int> d_vec(h_list.size());
thrust::copy(h_list.begin(), h_list.end(), d_vec.begin());

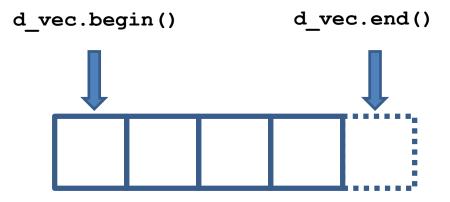
// alternative method
thrust::device_vector<int> d_vec(h_list.begin(), h_list.end());
```

Note: initializing an STL container with a device\_vector works, but results in one cudaMemcpy() for each element instead of a single cudaMemcpy for the entire vector.

Sequences defined by pair of iterators

```
// allocate device vector
thrust::device_vector<int> d_vec(4);

d_vec.begin(); // returns iterator at first element of d_vec
d_vec.end() // returns iterator one past the last element of d_vec
// [begin, end) pair defines a sequence of 4 elements
```



Iterators act like pointers

```
// allocate device vector
thrust::device vector<int> d vec(4);
thrust::device vector<int>::iterator begin = d vec.begin();
thrust::device vector<int>::iterator end = d vec.end();
int length = end - begin; // compute size of sequence [begin, end)
end = d_vec.begin() + 3; // define a sequence of 3 elements
                begin
                                  end
```

Use iterators like pointers

- Track memory space (host/device)
  - Guides algorithm dispatch

```
// initialize random values on host
thrust::host_vector<int> h_vec(1000);
thrust::generate(h_vec.begin(), h_vec.end(), rand);

// copy values to device
thrust::device_vector<int> d_vec = h_vec;

// compute sum on host
int h_sum = thrust::reduce(h_vec.begin(), h_vec.end());

// compute sum on device
int d_sum = thrust::reduce(d_vec.begin(), d_vec.end());
```

Convertible to raw pointers

```
// allocate device vector
thrust::device_vector<int> d_vec(4);

// obtain raw pointer to device vector's memory
int * ptr = thrust::raw_pointer_cast(&d_vec[0]);

// use ptr in a CUDA C kernel
my_kernel<<<N/256, 256>>>(N, ptr);

// Note: ptr cannot be dereferenced on the host!
// raw pointers do not know where they live
// Thrust iterators do
```

Wrap raw pointers with device ptr

```
int N = 10;
// raw pointer to device memory
int * raw ptr;
cudaMalloc((void **) &raw_ptr, N * sizeof(int));
// wrap raw pointer with a device ptr
thrust::device ptr<int> dev ptr(raw ptr);
// use device ptr in thrust algorithms
thrust::fill(dev_ptr, dev_ptr + N, (int) 0);
// access device memory through device ptr
dev ptr[0] = 1;
// extract raw pointer from device ptr
int * raw ptr2 = thrust::raw pointer cast(dev ptr);
// free memory
cudaFree(raw ptr);
```

## Recap

#### Containers

- Manage host & device memory
- Automatic allocation and deallocation
- Simplify data transfers

#### Iterators

- Behave like pointers
- Keep track of memory spaces
- Convertible to raw pointers

#### Namespaces

Avoid collisions

# C++ Background

### Function templates

```
// function template to add numbers (type of T is variable)
template< typename T >
T add(T a, T b)
  return a + b;
}
// add integers
int x = 10; int y = 20; int z;
z = add<int>(x,y);  // type of T explicitly specified
z = add(x,y); // type of T determined automatically
// add floats
float x = 10.0f; float y = 20.0f; float z;
z = add<float>(x,y); // type of T explicitly specified
z = add(x,y); // type of T determined automatically
```

# C++ Background

### Function objects (Functors)

```
// templated functor to add numbers
template< typename T >
class add
   public:
   T operator()(T a, T b)
      return a + b;
};
int x = 10; int y = 20; int z;
add<int> func; // create an add functor for T=int
z = func(x,y); // invoke functor on x and y
float x = 10; float y = 20; float z;
add<float> func; // create an add functor for T=float
z = func(x,y); // invoke functor on x and y
```

```
// this is a functor
// unlike functions, it can contain state
struct add x {
  add x(int x) : x(x) \{ \}
  int operator()(int y) { return x + y; }
private:
 int x;
};
// Now you can use it like this:
add x add42(42); // create an instance of the functor class
int i = add42(8); // and "call" it
assert(i == 50); // and it added 42 to its argument
std::vector<int> in; // assume this contains a bunch of values)
std::vector<int> out;
// Pass a functor to std::transform, which calls the functor on every
// element in the input sequence, and stores the result to the output
// sequence
// unlike a function pointer this can be resolved and inlined at
// compile time
std::transform(in.begin(), in.end(), out.begin(), add x(1));
assert(out[i] == in[i] + 1); // for all i
```

## C++ Background

### Generic Algorithms

```
// apply function f to sequences x, y and store result in z
template <typename T, typename Function>
void transform(int N, T * x, T * y, T * z, Function f)
   for (int i = 0; i < N; i++)
      z[i] = f(x[i], y[i]);
int N = 100;
int x[N]; int y[N]; int z[N];
                                   // add functor for T=int
add<int> func;
transform(N, x, y, z, func);
                                  // compute z[i] = x[i] + y[i]
transform(N, x, y, z, add<int>()); // equivalent
```

- Thrust provides many standard algorithms
  - Transformations
  - Reductions
  - Prefix Sums
  - Sorting
- Generic definitions
  - General Types
    - Built-in types (int, float, ...)
    - User-defined structures
  - General Operators
    - reduce with plus operator
    - scan with maximum operator

General types and operators

```
#include <thrust/reduce.h>
// declare storage
device vector<int> i vec = ...
                                               Initial value of sum
device vector<float> f vec = ...
// sum of integers (equivalent calls)
reduce(i vec.begin(), i vec.end());
reduce(i vec.begin(), i_vec.end(),
                                      0, plus<int>());
// sum of floats (equivalent calls)
reduce(f vec.begin(), f_vec.end());
reduce(f vec.begin(), f_vec.end(), 0.0f, plus<float>());
// maximum of integers
reduce(i vec.begin(), i vec.end(), 0, maximum<int>());
```

General types and operators

```
struct negate float2
{
     host device
   float2 operator()(float2 a)
       return make float2(-a.x, -a.y);
};
// declare storage
device vector<float2> input = ...
device vector<float2> output = ...
// create functor
negate float2 func;
// negate vectors
transform(input.begin(), input.end(), output.begin(), func);
```

General types and operators

```
// compare x component of two float2 structures
struct compare float2
    host device
  bool operator()(float2 a, float2 b)
       return a.x < b.x;
};
// declare storage
device vector<float2> vec = ...
// create comparison functor
compare float2 comp;
// sort elements by x component
sort(vec.begin(), vec.end(), comp);
```

### Operators with State

```
// compare x component of two float2 structures
struct is greater than
   int threshold;
   is_greater_than(int t) { threshold = t; }
    host device
  bool operator()(int x) { return x > threshold; }
};
device vector<int> vec = ...
// create predicate functor (returns true for x > 10)
is greater than pred(10);
// count number of values > 10
int result = count if(vec.begin(), vec.end(), pred);
```

## Recap

- Algorithms
  - Generic
    - Support general types and operators
  - Statically dispatched based on iterator type
    - Memory space is known at compile time
  - Have default arguments
    - reduce (begin, end)
    - reduce (begin, end, init, binary op)

- Behave like "normal" iterators
  - Algorithms don't know the difference

### Examples

- -constant iterator
- -counting iterator
- transform\_iterator
- permutation\_iterator
- zip\_iterator

- constant iterator
  - Mimics an infinite array filled with a constant value

```
// create iterators
constant_iterator<int> begin(10);
constant_iterator<int> end = begin + 3;

begin[0]  // returns 10
begin[1]  // returns 10
begin[100]  // returns 10

// sum of [begin, end)
reduce(begin, end);  // returns 30 (i.e. 3 * 10)
```







- counting iterator
  - Mimics an infinite array with sequential values

```
// create iterators
counting_iterator<int> begin(10);
counting_iterator<int> end = begin + 3;

begin[0]  // returns 10
begin[1]  // returns 11
begin[100]  // returns 110

// sum of [begin, end)
reduce(begin, end);  // returns 33 (i.e. 10 + 11 + 12)
```





0 1 2 3

- transform iterator
  - Yields a transformed sequence
  - Facilitates kernel fusion (e.g. sum of squares)



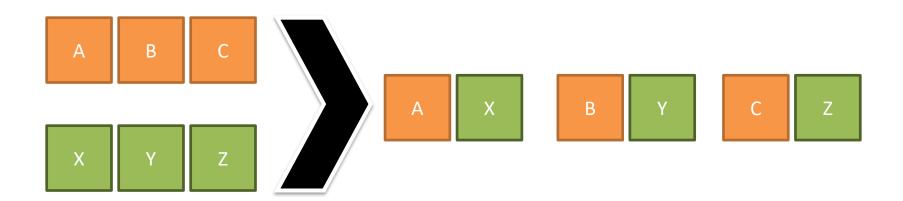
# Fancy Iterators

- transform iterator
  - Conserves memory capacity and bandwidth

```
// initialize vector
device vector<int> vec(3);
vec[0] = 10; vec[1] = 20; vec[2] = 30;
// create iterator (type omitted)
first = make transform iterator(vec.begin(), negate<int>());
last = make transform iterator(vec.end(),
                                            negate<int>());
first[0] // returns -10
first[1] // returns -20
first[2] // returns -30
// sum of [begin, end)
reduce(first, last); // returns -60 (i.e. -10 + -20 + -30)
```

# **Fancy Iterators**

- zip\_iterator
  - Looks like an array of structs (AoS)
  - Stored in structure of arrays (SoA)



### Fancy Iterators

• zip iterator

```
// initialize vectors
device vector<int> A(3);
device vector<char> B(3);
A[0] = 10; A[1] = 20; A[2] = 30;
B[0] = 'x'; B[1] = 'y'; B[2] = 'z';
// create iterator (type omitted)
first = make zip iterator(make tuple(A.begin(), B.begin()));
last = make zip iterator(make tuple(A.end(), B.end()));
first[0] // returns tuple(10, 'x')
first[1] // returns tuple(20, 'y')
first[2] // returns tuple(30, 'z')
// maximum of [begin, end)
maximum< tuple<int,char> > binary op;
reduce(first,last, first[0], binary op); // returns tuple(30, 'z')
// tuple() defines a comparison operator
                                                             75
```

#### **Best Practices**

- Fusion
  - Combine related operations together

- Structure of Arrays
  - Ensure memory coalescing

- Implicit Sequences
  - Eliminate memory accesses

#### **Fusion**

- Combine related operations together
  - Conserves memory bandwidth
- Example: SNRM2
  - Square each element
  - Compute sum of squares and take sqrt()
  - The fused implementation reads the array once while the un-fused implementation performs 2 reads and 1 write per element

#### **Fusion**

Unoptimized implementation

```
// define transformation f(x) \rightarrow x^2
struct square
      host device
    float operator()(float x)
        return x * x;
};
float snrm2 slow(device vector<float>& x)
  // without fusion
  device vector<float> temp(x.size());
  transform(x.begin(), x.end(), temp.begin(), square());
  return sqrt( reduce(temp.begin(), temp.end()) );
```

#### **Fusion**

Optimized implementation (3.8x faster)

```
// define transformation f(x) -> x^2
struct square
     host device
    float operator()(float x)
        return x * x;
float snrm2 fast(device vector<float>& x)
  // with fusion
  return sqrt( transform reduce(x.begin(), x.end(),
                                square(), 0.0f, plus<float>());
```

# Structure of Arrays (SoA)

- Array of Structures (AoS)
  - Often does not obey coalescing rules
    - device vector<float3>
- Structure of Arrays (SoA)
  - Obeys coalescing rules
  - Components stored in separate arrays
    - device vector<float> x, y, z;
- Example: Rotate 3d vectors
  - SoA is 2.8x faster

# Array of Structures (AoS)

```
struct rotate float3
   host device
  float3 operator()(float3 v)
   float x = v.x;
    float y = v.y;
   float z = v.z;
    float rx = 0.36f*x + 0.48f*y + -0.80f*z;
    float ry =-0.80f*x + 0.60f*y + 0.00f*z;
    float rz = 0.48f*x + 0.64f*y + 0.60f*z;
   return make float3(rx, ry, rz);
};
device vector<float3> vec(N);
transform(vec.begin(), vec.end, vec.begin(), rotate float3());
```

# Structure of Arrays (SoA)

```
struct rotate tuple
             device
    host
  tuple<float,float,float> operator()(tuple<float,float,float> v)
    float x = qet<0>(v);
    float y = qet<1>(v);
    float z = get < 2 > (v);
    float rx = 0.36f*x + 0.48f*y + -0.80f*z;
    float ry =-0.80f*x + 0.60f*y + 0.00f*z;
    float rz = 0.48f*x + 0.64f*y + 0.60f*z;
    return make tuple(rx, ry, rz);
};
device vector\langle float \rangle \times (N), y(N), z(N);
transform(make zip iterator(make tuple(x.begin(), y.begin(), z.begin())),
          make zip iterator(make tuple(x.end(), y.end(), z.end())),
          make zip iterator(make tuple(x.begin(), y.begin(), z.begin())),
          rotate tuple());
```

### Implicit Sequences

- Avoid storing sequences explicitly
  - Constant sequences
    - [1, 1, 1, 1, ...]
  - Incrementing sequences
    - [0, 1, 2, 3, ... ]
- Implicit sequences require no storage
  - constant\_iterator
  - counting iterator
- Example
  - Index of the smallest element

#### Implicit Sequences

```
// return the smaller of two tuples
struct smaller tuple
  tuple<float,int> operator()(tuple<float,int> a, tuple<float,int> b)
    if (a < b)
      return a;
    else
      return b;
};
int min index(device vector<float>& vec)
  // create explicit index sequence [0, 1, 2, ...)
  device vector<int> indices(vec.size());
  sequence(indices.begin(), indices.end());
  tuple<float,int> init(vec[0],0);
  tuple<float,int> smallest;
  smallest = reduce(make zip iterator(make tuple(vec.begin(), indices.begin())),
                    make zip iterator(make tuple(vec.end(),
                                                               indices.end())),
                    init,
                    smaller tuple());
  return get<1>(smallest);
```

#### Implicit Sequences

```
// return the smaller of two tuples
struct smaller tuple
  tuple<float,int> operator()(tuple<float,int> a, tuple<float,int> b)
    if (a < b)
      return a;
    else
      return b;
};
int min index(device vector<float>& vec)
  // create implicit index sequence [0, 1, 2, ...)
  counting iterator<int> begin(0);
  counting iterator<int> end(vec.size());
  tuple<float,int> init(vec[0],0);
  tuple<float,int> smallest;
  smallest = reduce(make zip iterator(make tuple(vec.begin(), begin)),
                    make zip iterator(make tuple(vec.end(),
                                                                 end)),
                    init,
                    smaller tuple());
  return get<1>(smallest);
```

# Recap

- Best Practices
  - Fusion
    - 3.8x faster
  - Structure of Arrays
    - 2.8x faster
  - Implicit Sequences
    - 3.4x faster

#### Additional Resources

- Thrust
  - Homepage <a href="http://thrust.github.io/">http://thrust.github.io/</a>
  - More

http://docs.nvidia.com/cuda/thrust/index.html https://developer.nvidia.com/thrust