

# CS 677: Parallel Programming for Many-core Processors

## Lecture 10

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# Project Status Update

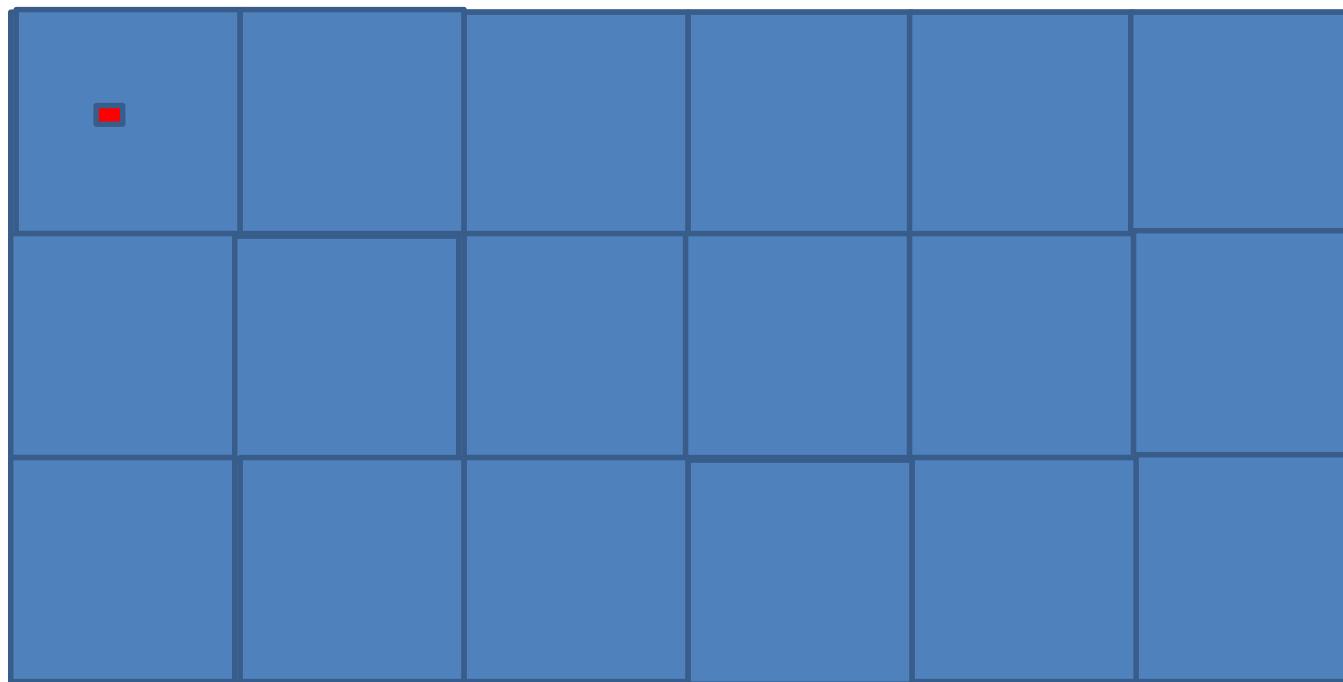
- Due April 12
  1. What is the status of the CPU version? If you are using existing code for this part, cite the source of the code.
  2. What is the status of the GPU version in terms of completeness? Which functionalities have been implemented and what is missing?
  3. What is the status of the GPU version in terms of correctness? Is the, potentially unoptimized, GPU version correct? If not, what is your plan for achieving correctness?
- Be prepared to talk about it in class

# Outline

- Homework 4 discussion
- Thrust

# Tiling P

- Use a thread block to calculate a tile of P
  - Thread Block size determined by the TILE\_SIZE



# Tiling N

- Each N element is used in calculating up to KERNEL\_SIZE \* KERNEL\_SIZE elements of P (all elements in the tile)

3	4	5	6	7
2	3	4	5	6
1	2	3	4	5
2	3	5	6	7
0	1	1	3	1

3	4	5	6	7
2	3	4	5	6
1	2	3	4	5
2	3	5	6	7
0	1	1	3	1

3	4	5	6	7
2	3	4	5	6
1	2	3	4	5
2	3	5	6	7
0	1	1	3	1

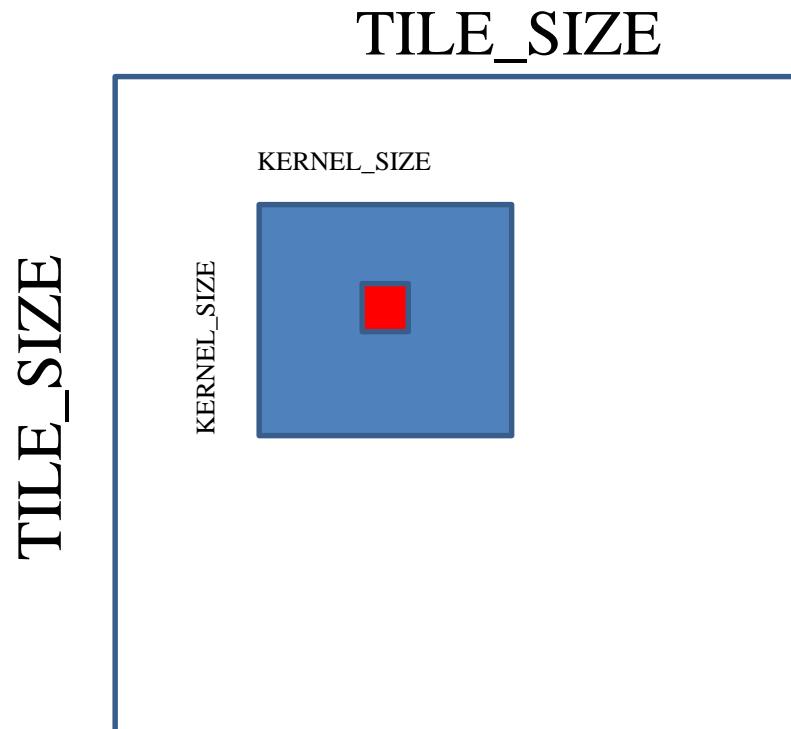
3	4	5	6	7
2	3	4	5	6
1	2	3	4	5
2	3	5	6	7
0	1	1	3	1

3	4	5	6	7
2	3	4	5	6
1	2	3	4	5
2	3	5	6	7
0	1	1	3	1

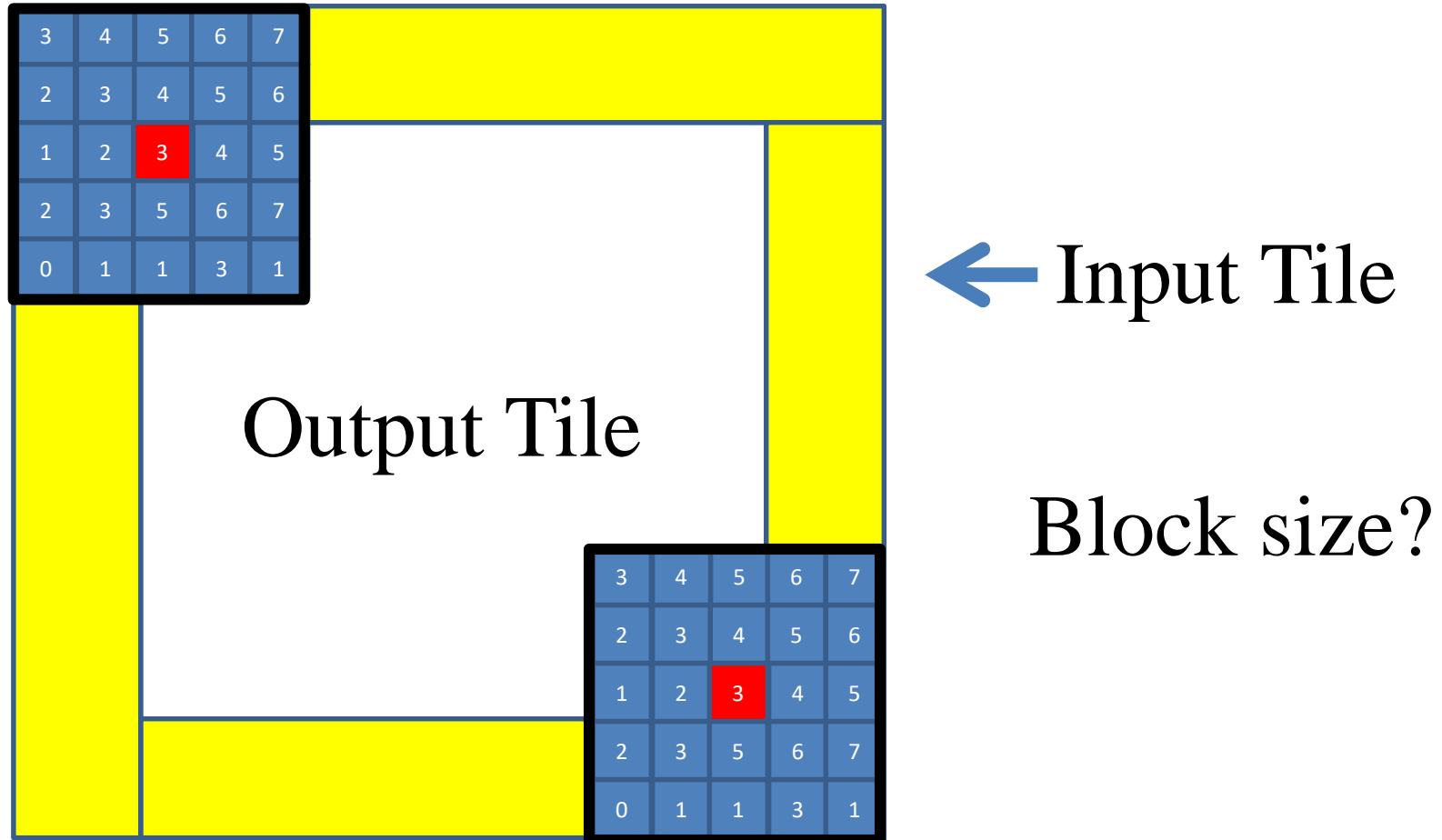
3	4	5	6	7
2	3	4	5	6
1	2	3	4	5
2	3	5	6	7
0	1	1	3	1

# High-Level Tiling Strategy

- Load a tile of  $N$  into shared memory (SM)
  - All threads participate in loading
  - A subset of threads then use each  $N$  element in SM



# Input tiles need to be larger than output tiles



# Dealing with Mismatch

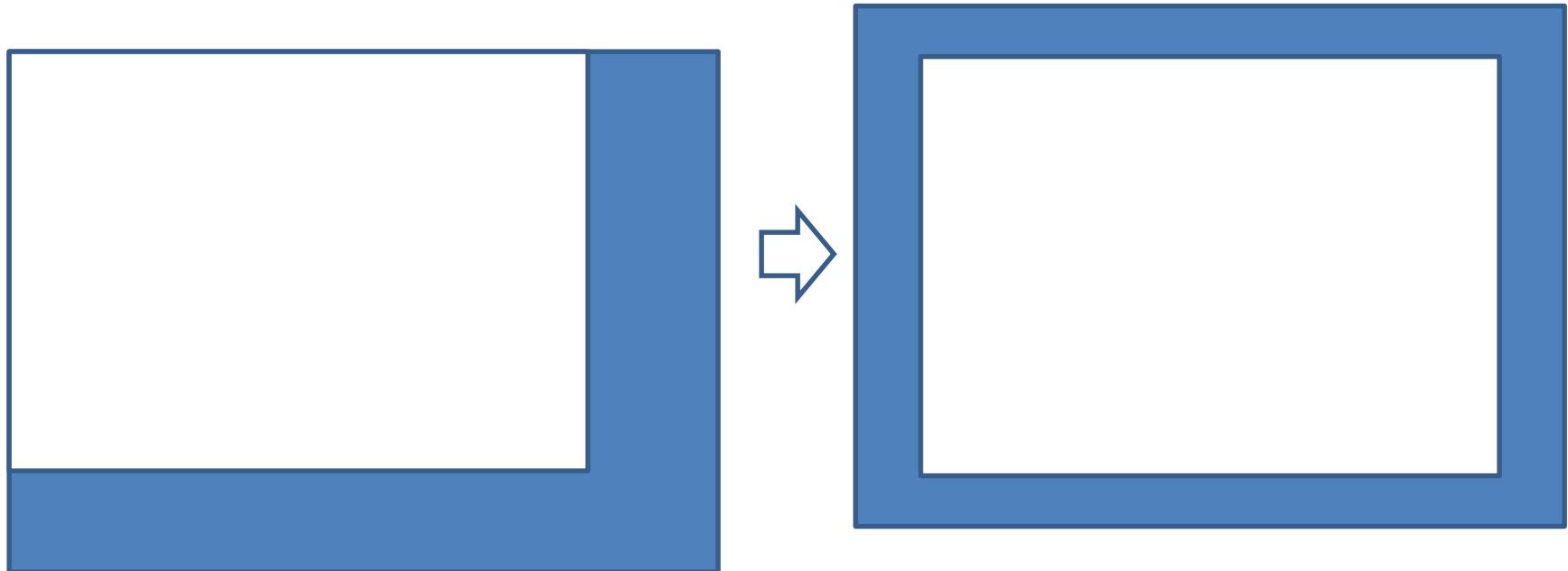
- Use a thread block that matches input tile
  - Each thread loads one element of the input tile
  - Some threads do not participate in calculating output
    - There will be if statements and control divergence

# Setting Block Size

```
#define BLOCK_SIZE (TILE_SIZE + 4)  
  
dim3 dimBlock(BLOCK_SIZE, BLOCK_SIZE);
```

In general, block size should be  
tile size + (kernel size -1)

# Shifting from output coordinates to input coordinates



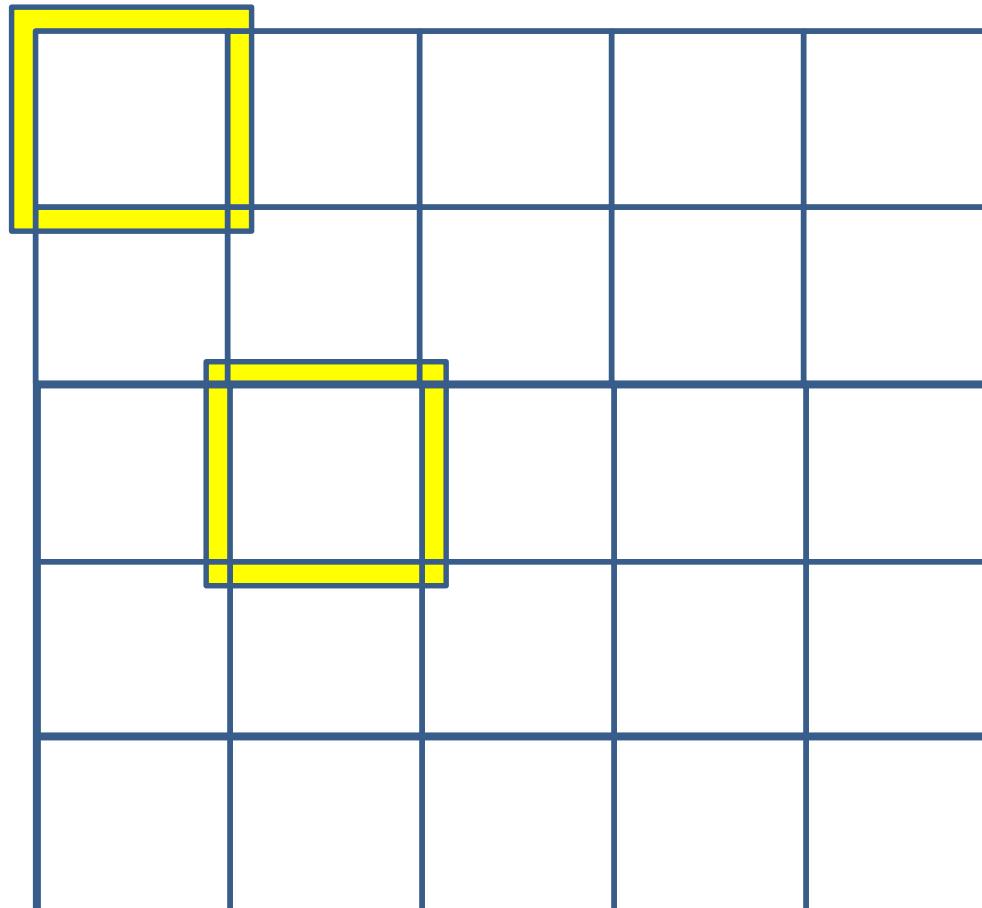
# Shifting from output coordinates to input coordinates

```
int tx = threadIdx.x;  
int ty = threadIdx.y;  
int row_o = blockIdx.y * TILE_SIZE + ty;  
int col_o = blockIdx.x * TILE_SIZE + tx;
```

```
int row_i = row_o - 2;  
int col_i = col_o - 2;
```

This is for 5x5  
mask

# Threads that loads halos outside N should return 0.0



# Taking Care of Boundaries - Ghost Cells

```
float output = 0.0f;

if((row_i >= 0) && (row_i < N.height) &&
   (col_i >= 0) && (col_i < N.width) ) {
    Ns[ty][tx] = N.elements[row_i*N.width + col_i];
}
else{
    Ns[ty][tx] = 0.0f;
}
```

# Some threads do not participate in calculating output

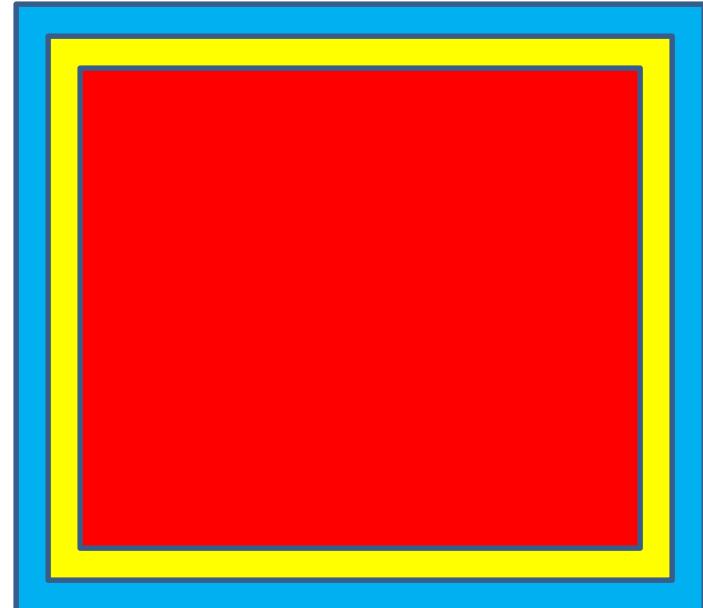
```
if(ty < TILE_SIZE && tx < TILE_SIZE) {  
    for(i = 0; i < 5; i++) {  
        for(j = 0; j < 5; j++) {  
            output += Mc[i][j] * Ns[i+ty][j+tx];  
        }  
    }  
}
```

# Some threads do not write output

```
if(row_o < P.height && col_o < P.width)
    P.elements[row_o * P.width + col_o] =
        output;
```

# Tiling Benefit Analysis

- Start with KERNEL\_SIZE = 5
- Each point in an input tile is used multiple times.
  - Each boundary point (blue) is used 9 times
  - Each second-boundary point (yellow) is used 16 times
  - Each inner boundary point (red) is used 25 times



# Reuse Analysis

- For TILE\_SIZE = 12
  - 44 boundary points
  - 36 second-boundary points
  - 64 inside points
  - Total uses  $44*9 + 36*16 + 64*25 = 396+576+1600 = 2572$
  - Average reuse =  $2572/144 = 17.9$
- As TILE\_SIZE increases, the average reuse approach 25

# In General

- The number of boundary layers is proportional to the KERNEL\_SIZE
- The maximal reuse of each data point is  $(\text{KERNEL\_SIZE})^2$

# Second Approach

- Can block dimensions match output size dimensions?

# Third Approach

- Match BLOCK\_SIZE with TILE\_SIZE instead of TILE\_SIZE+KERNEL\_SIZE-1
- Exploit L2 cache which is shared by all SMs
  - All relevant halo cells have been transferred to shared memory by some block
  - Therefore, they must be in L2 cache
  - Access them directly from “global memory”

# Thrust

Jared Hoberock and Nathan Bell  
Modified by P. Mordohai (March 2013)



# 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;

// 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;

// 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);
    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_vector`
    - `thrust::copy`
  - STL uses `std` namespace
    - `std::vector`
    - `std::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()
```

# 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.

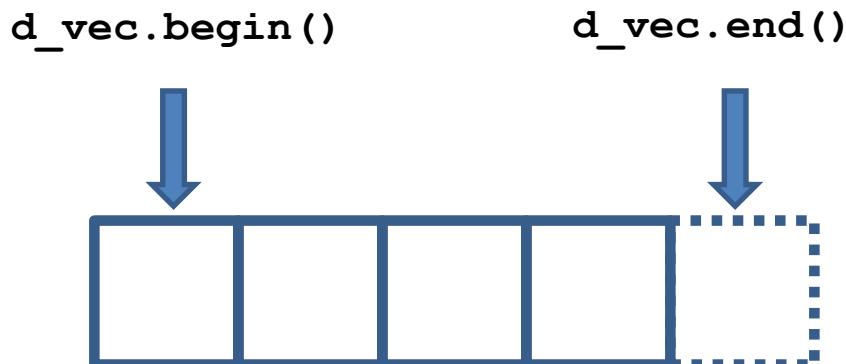
# Iterators

- 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

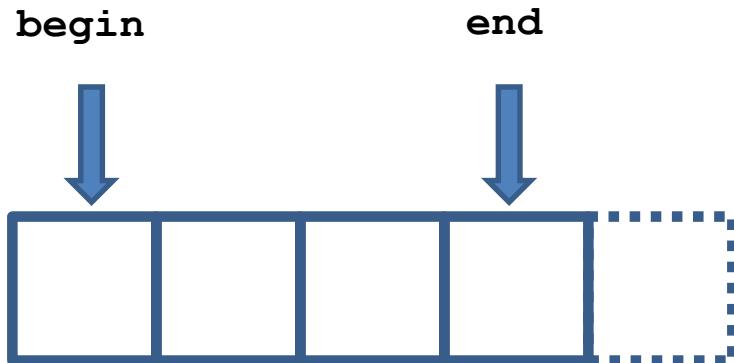
- Iterators act like pointers

```
// 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
```



# Iterators

- Use iterators like pointers

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

thrust::device_vector<int>::iterator begin = d_vec.begin();

*begin = 13;           // same as d_vec[0] = 13;
int temp = *begin;     // same as temp = d_vec[0];

begin++;              // advance iterator one position

*begin = 25;           // same as d_vec[1] = 25;
```

# Iterators

- 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());
```

# Iterators

- 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
```

# Iterators

- 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<int> func;                                // add functor for T=int

transform(N, x, y, z, func);                  // compute z[i] = x[i] + y[i]

transform(N, x, y, z, add<int>()); // equivalent
```

# Algorithms

- 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

# Algorithms

- General types and operators

```
#include <thrust/reduce.h>

// declare storage
device_vector<int> i_vec = ...
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>());
```

An arrow points from the text "Initial value of sum" to the "0" parameter in the second reduce call, indicating its purpose.

# Algorithms

- 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);
```

# Algorithms

- 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);
```

# Algorithms

- 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`)

# Fancy Iterators

- Behave like “normal” iterators
  - Algorithms don't know the difference
- Examples
  - constant\_iterator
  - counting\_iterator
  - transform\_iterator
  - permutation\_iterator
  - zip\_iterator

# Fancy Iterators

- `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)
```



# Fancy Iterators

- counting\_iterator
  - Mimics an infinite array with sequential values

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

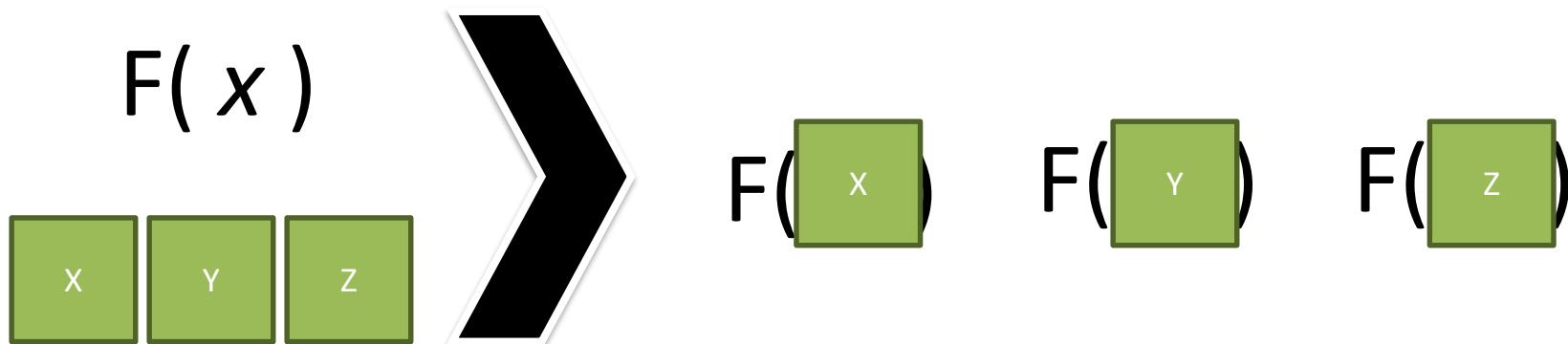
begin[0]    // returns 10
begin[1]    // returns 11
begin[100]  // returns 110

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



# Fancy Iterators

- `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
```

# 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) -> 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
{
    __host__ __device__
    tuple<float, float, float> operator() (tuple<float, float, float> v)
    {
        float x = get<0>(v);
        float y = get<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<float> x(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 <http://thrust.github.io/>
  - More
    - <http://docs.nvidia.com/cuda/thrust/index.html>
    - <https://developer.nvidia.com/thrust>