

Introduction to CUDA

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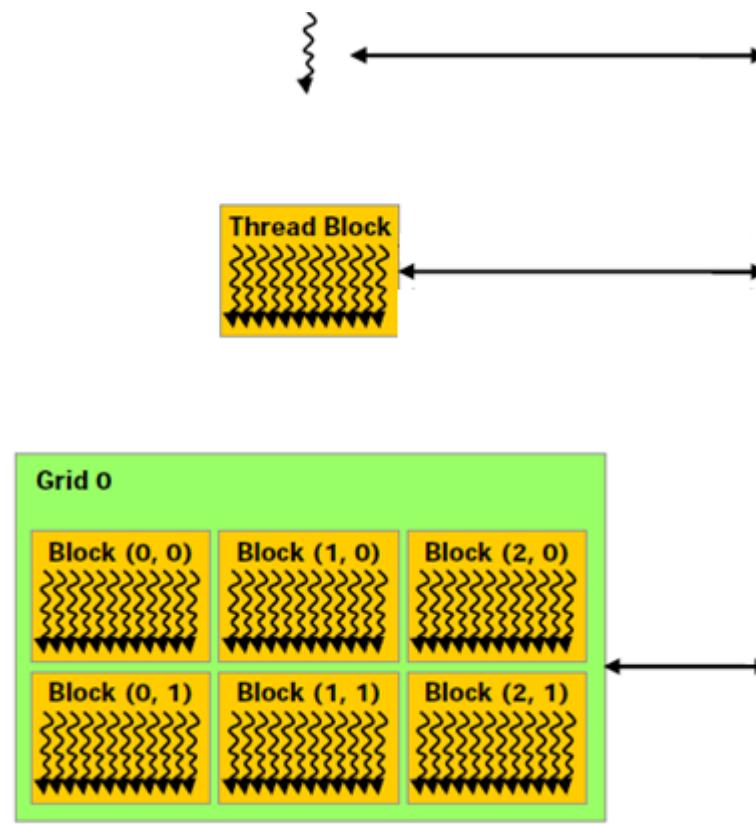
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Threads Execution and Warps

- Blocks and Grids of Blocks



WARP

- SIMD hardware executes all threads of a warp as a bundle.
- Divergence in warp
 - If loop condition is based on thread index
 - `for (i=0, i < threadIdx.x; i++){};`
 - If conditions
 - `If(threadIdx > 2) {}`

Types of Memories

- Registers
 - fadd r1, r2, r3
- Shared memory
 - load r1, r2, offset
 - fadd r1,r2,r3
- Global memory
 - foad r1,r2, offset
 - fadd r1, r2, r3

Memory coalescing

- Global memory access by a half warp coalesced in to a single transaction if:
 - It accesses contiguous region of global memory:
 - 64 bytes - each thread reads 4 bytes.
 - 128 bytes - each thread reads 8 bytes
 - 256 bytes – each thread reads 16 bytes
- This defines how the optimum memory bandwidth can be achieved on GPU

Example-With shared memory

```
#include <stdlib.h>
#include <stdio.h>
#include "cuda.h"
#define BLOCK_DIM 16

__global__ void transpose(float *odata, float *idata, int W, int H)
{
    __shared__ float block[BLOCK_DIM][BLOCK_DIM+1];

    // read the matrix tile into shared memory
    unsigned int xIndex = blockIdx.x * BLOCK_DIM + threadIdx.x;
    unsigned int yIndex = blockIdx.y * BLOCK_DIM + threadIdx.y;

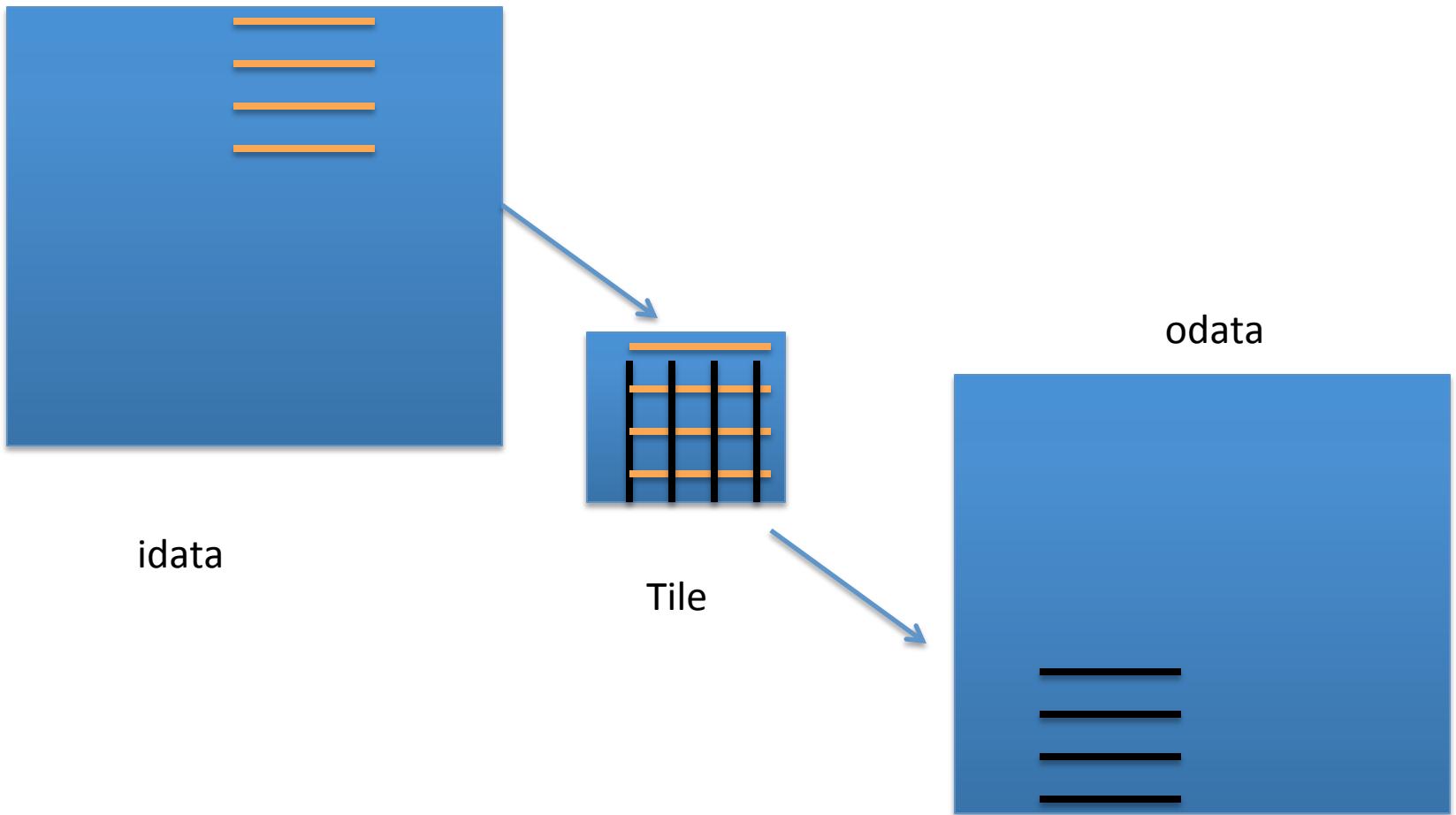
    if((xIndex < W) && (yIndex < H)){
        unsigned int index_in = yIndex * W + xIndex;
        block[threadIdx.y][threadIdx.x] = idata[index_in];}

    __syncthreads();

    // write the transposed matrix tile to global memory
    xIndex = blockIdx.y * BLOCK_DIM + threadIdx.x;
    yIndex = blockIdx.x * BLOCK_DIM + threadIdx.y;

    if((xIndex < H) && (yIndex < W)){
        unsigned int index_out = yIndex * H + xIndex;
        odata[index_out] = block[threadIdx.x][threadIdx.y];}
}
```

Transpose using Shared Memory



Sum reduction example-1

	Thread -0	Thread -1	Thread -2	Thread -3	Thread -4	Thread -5	Thread -6	Thread -7
Step-1	0	1	2	3	4	5	6	7
Step-2	0+1		2+3		4+5		6+7	
Step-3	0+2				4+7			
Step-4	0+4							

Sum reduction example-1

```
__shared__ float psum[];  
  
unsigned int t = threadIdx.x;  
i = blockIdx.y*blockDim.y+threadIdx.y;  
  
psum[t] = x[i]  
__syncthreads();  
  
for (int stride =1; stride < blockDim.x; stride *=2){  
  
__syncthreads();  
  
if (t % (2*stride) == 0)pum[t] += psum[t+stride];  
}
```

Sum reduction example-2

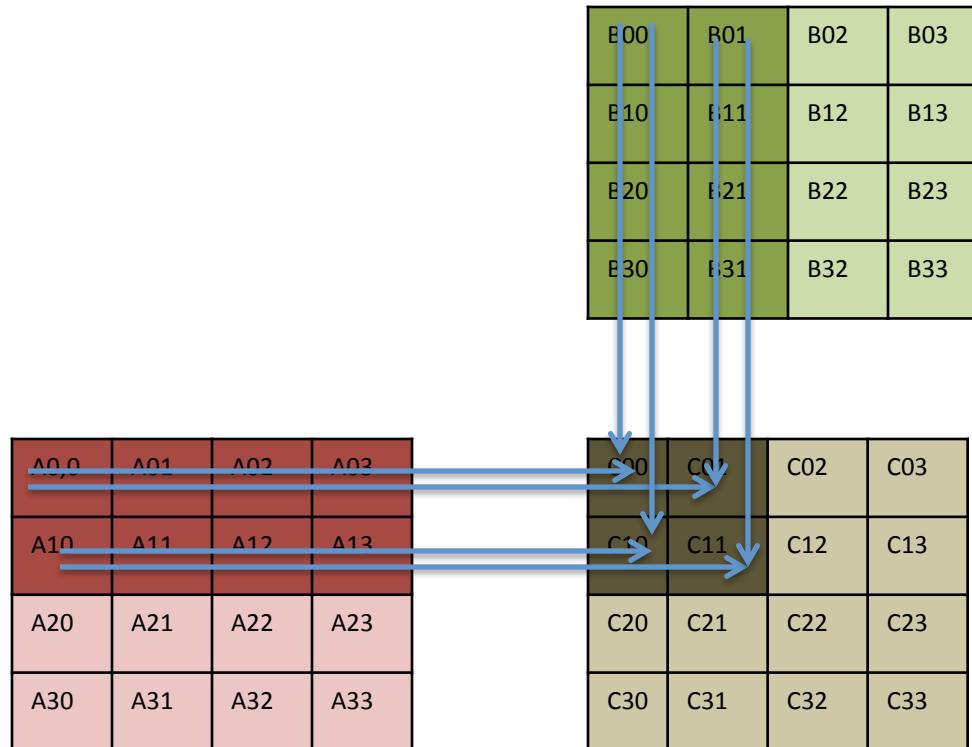
Thread-0	Thread-1	Thread-2	Thread-3	Thread-4	Thread-5	Thread-6	Thread-7
0	1	2	3	4	5	6	7
0+8	1+9	2+10	3+11	4+12	5+13	6+14	7+15
0+4	1+5	2+6	3+7				
0+2	1+3						
0+2							

The diagram illustrates a sum reduction operation across 8 threads. The initial values are 0 through 7. Thread 0 adds 8 to its value. Thread 1 adds 9 to its value. Thread 2 adds 10 to its value. Thread 3 adds 11 to its value. Thread 4 adds 12 to its value. Thread 5 adds 13 to its value. Thread 6 adds 14 to its value. Thread 7 adds 15 to its value. Arrows indicate the flow of data from each thread's local value to the final sum.

Sum reduction example-2

```
__shared__ float psum[];  
  
unsigned int t = threadIdx.x;  
i = blockIdx.y*blockDim.y+threadIdx.y;  
  
psum[t] = x[i]  
__syncthreads();  
  
for (int stride =blockDim.x; stride > 1; stride /=2){  
  
__syncthreads();  
  
if ( t < stride ) pum[t] += psum[t+stride];  
}  
}
```

Example – Matrix multiplication



Code-host

```
__host__ void matrixmul_h(float A[N][N], float B[N][N], float C[N][N])
{
    for( int row =0; row < N; ++row){
        for( int col =0; col < N; ++col){
            float element = 0;
            for (int k = 0; k < N; ++k) element += A[row][k] * B[k][col];
            C[row][col] = element;
        }
    }
    return;
}
```

Code-GPU

```
__global__ void matrixmul_d(float* A, float* B, float* C)
{
    unsigned int row  = blockIdx.y*blockDim.y+threadIdx.y;
    unsigned int col  = blockIdx.x*blockDim.x+threadIdx.x;

    if( (row < N) && (col < N) ) {
        float element = 0;
        for (int k = 0; k < N; ++k) element += A[row*N+k] * B[k*N+col];
        C[row*N+col] = element;
    }
    return;
}
```

Access Order

C00=A00, B00+A01,B10+ A02,B20+A03,B 03	C01=A00, B01+A01,B11+ A02,B21+A03,B 31		
C10=A10,B01+ A11,B10+A12,B 20+A13,B03	C11=A10, B01+A11,B11+ A12,B21+A13,B 31		

Tiled Matrix multiplication

$$C_{00-1} = A_{00} * B_{00} + A_{01} * B_{10}$$

$$C_{01-1} = A_{00} * B_{01} + A_{01} * B_{11}$$

$$C_{10-1} = A_{10} * B_{00} + A_{11} * B_{10}$$

$$C_{11-1} = A_{10} * B_{01} + A_{11} * B_{11}$$

$$C_{00} = C_{00-1} + C_{00-2}$$

$$C_{01} = C_{01-1} + C_{01-2}$$

$$C_{10} = C_{10-1} + C_{10-2}$$

$$C_{11} = C_{11-1} + C_{11-2}$$

A _{0,0}	A _{0,1}	A _{0,2}	A _{0,3}
A _{1,0}	A _{1,1}	A _{1,2}	A _{1,3}
A _{2,0}	A _{2,1}	A _{2,2}	A _{2,3}
A _{3,0}	A _{3,1}	A _{3,2}	A _{3,3}

B _{0,0}	B _{0,1}	B _{0,2}	B _{0,3}
B _{1,0}	B _{1,1}	B _{1,2}	B _{1,3}
B _{2,0}	B _{2,1}	B _{2,2}	B _{2,3}
B _{3,0}	B _{3,1}	B _{3,2}	B _{3,3}

C _{0,0}	C _{0,1}	C _{0,2}	C _{0,3}
C _{1,0}	C _{1,1}	C _{1,2}	C _{1,3}
C _{2,0}	C _{2,1}	C _{2,2}	C _{2,3}
C _{3,0}	C _{3,1}	C _{3,2}	C _{3,3}

Code- Tiled

```
__global__ void matrixmul_d(float* A, float* B, float* C)
{
    __shared__ float A_ds[TILE_WIDTH][TILE_WIDTH];
    __shared__ float B_ds[TILE_WIDTH][TILE_WIDTH];
    unsigned int bx= blockIdx.x;
    unsigned int tx= threadIdx.x;
    unsigned int by= blockIdx.y;
    unsigned int ty= threadIdx.y;
    unsigned int row = by*TILE_WIDTH+ty;
    unsigned int col = bx*TILE_WIDTH+tx;

    float element = 0;
    for (int m = 0; m < N/TILE_WIDTH; ++m) {
        A_ds[tx][ty] = A[N*(m*row+TILE_WIDTH)+tx];
        B_ds[tx][ty] = B[N*col+m*TILE_WIDTH+ty];
        __syncthreads();
        for (int k = 0; k < TILE_WIDTH; ++k) element += A_ds[tx][k]*B_ds[k][ty];
        __syncthreads();
    }
    C[row*N+col] = element;
    return;
}
```

Communication between host and device

- The bandwidth between host and device is much less than the memory bandwidth on the device
 - The communication between host and device is costly
 - Less data transfer between host and device is preferable for better performance
- Streams could help in computing while communicating

Stream

- Applications manage concurrency through streams.
- A sequence of commands that execute in order.
- Different streams may execute their commands out of order.
- Defined by creating a stream object.
 - Objects can be used as parameters to sequence of kernels and memory copies.

Stream example

```
#include<cuda.h>
#include<stdio.h>
# define N 1024
# define BLOCK_SIZE 64
__global__ void VecAdd(float* A, float* B, float* C){
    int j = blockIdx.x*blockDim.x+threadIdx.x;
    if(j<N)C[j]=A[j]+B[j];
    return; }

main(){
float *a, *b, *ab;
int vecsize = sizeof(float)*N;
    cudaMallocHost((void **) &a , 2*vecsize);
    cudaMallocHost((void **) &b , 2*vecsize);
    cudaMallocHost((void **) &ab , 2*vecsize);
float *a_d, *b_d, *ab_d;
    cudaMalloc((void **) &a_d , 2*vecsize);
    cudaMalloc((void **) &b_d , 2*vecsize);
    cudaMalloc((void **) &ab_d , 2*vecsize);
int i, j;
    for (j=0; j< 2*N; j++){
        a[j] = rand()%100;
        b[j] = rand()%100;
        ab[j]=0.0; }

    cudaStream_t stream[2];
    for (i=0; i<2; ++i) cudaStreamCreate(&(stream[i]));

    for (i=0; i<2; ++i) cudaMemcpyAsync(a_d+i*N, a+i*N, vecsize,
        cudaMemcpyHostToDevice, stream[i]);

    for (i=0; i<2; ++i) cudaMemcpyAsync( b_d+i*N, b+i*N,
        vecsize ,cudaMemcpyHostToDevice,stream[i]);

    dim3 dimBlock(BLOCK_SIZE);
    dim3 dimGrid (N/dimBlock.x);
    if( N % BLOCK_SIZE != 0 ) dimGrid.x+=1;
    for (i=0; i<2; ++i) VecAdd<<<dimGrid,dimBlock,0,stream[i]>>> (a_d
        +i*N,b_d+i*N,ab_d+i*N);

    for (i=0; i<2; ++i)cudaMemcpyAsync(ab + i * N, ab_d + i * N, vecsize,
        cudaMemcpyDeviceToHost, stream[i]);
    cudaThreadSynchronize();
    for (i=0; i<2; i++) cudaStreamDestroy(stream[i]);

    cudaFree(a_d); cudaFree(b_d); cudaFree(ab_d);
    cudaFreeHost(a); cudaFreeHost(b); cudaFreeHost(ab);
```

Stream..

- `cudaStreamCreate(cudaStream_t &stream);`
 - Creates Stream
- `cudaMallocHost((void**)&hostPtr, size);`
 - Page locked memory allocation on host
- `cudaMemcpyAsync(to,from,size,direction,stream);`
 - Asynchronous memory copy
- `Kernel<<<gridDim, blockDim, 0, stream>>>(...);`
 - Asynchronous kernel launch
- `cudaStreamDestroy(stream);`
 - Delete the stream object
- `cudaStreamQuery(stream);`
 - Query stream
- `cudaStreamSynchronize(stream);`
 - Synchronize a single stream
- `cudaThreadSynchronize();`
 - Wait till all stream finishes

Event management

- Events could created in CUDA code
 - For example
 - To measure the time
 - To block CPU until a CUDA call finish
 - Functions:
 - `cudaEvent_t start, stop;`
 - `cudaEventCreate(&start);`
 - `cudaEventCreate(&stop);`
 - `cudaEventRecord(start, 0);`
 - `kernel<<<gridDim, blockDim>>>(...);`
 - `cudaEventRecord(stop, 0);`
 - `cudaEventSynchronize(stop);`
 - `float etime;`
 - `cudaEventElapsedTime(&etime, start, stop);`
 - `cudaEventDestroy(start);`
 - `cudaEventDestroy(stop);`