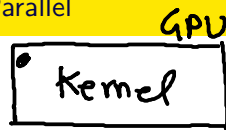


That was Blocks in Parallel. What about Threads in Parallel



Function Call Change

Cuda

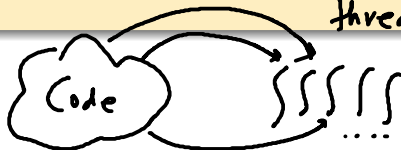
```
// add<<<1, 1>>>(da, db, dc); // single thread GPU
// add<<<N, 1>>>(da, db, dc); // N blocks on GPU
add<<<1, N>>>(da, db, dc); // N threads on GPU
```

Changes in Kernel Code

```
__global__ void add(int *a, int *b, int *c)
{
    c[threadIdx.x] = a[threadIdx.x] + b[threadIdx.x];
}
```

Single instruction
multiple
threads

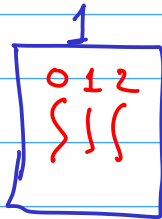
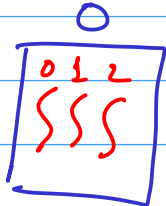
- Rest of Host code would be the same



$3 \times 2 = 6$ threads.

fn Name <<< 2, 3 >>> (—) ;

blocks threads.



1 | 1 | 1 | 1 | 1 | 1 | 1 | 1

+

1 | 1 | 1 | 1 | 1 | 1 | 1 | 1

=

2 | 2 | 2 | 2 | 2 | 2 | 2 | 2

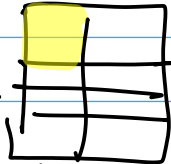
A[8]

B[8]

C[8]

GPU RAM

8 cores



$\frac{1}{8} = 12.5\%$

add <<< 8 , 1 >>> (A, B, C);



blockId
threadId

$c[blockIdx.x] = a[blockIdx.x] + b[blockIdx.x]$

1 | 1 | 1 | 1 | 1 | 1 | 1 | 1

+

1 | 1 | 1 | 1 | 1 | 1 | 1 | 1

=

2 | 2 | 2 | 2 | 2 | 2 | 2 | 2

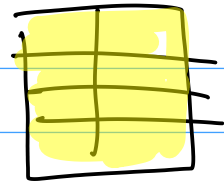
A[8]

B[8]

C[8]

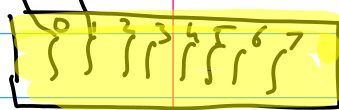
GPU RAM

8 rows



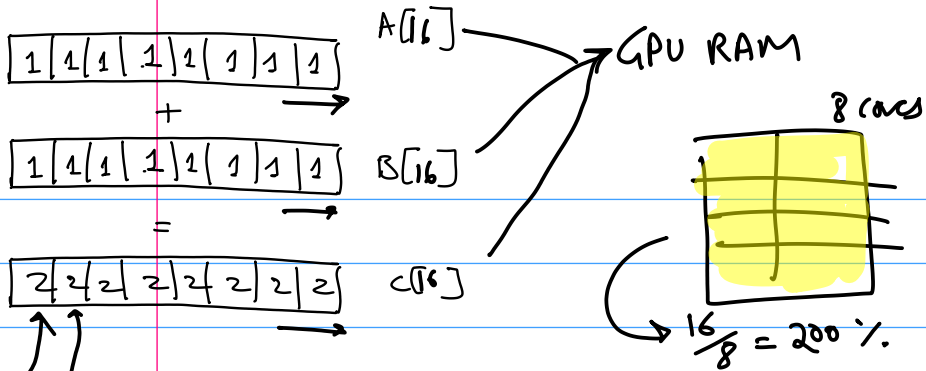
$\frac{8}{8} = 100\%$

add <<< 1, 8 >>> (A, B, C);

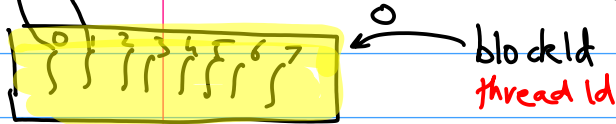


block id
thread id

$$C[\text{threadId} \times n] = a[\text{threadId} \times n] + b[\text{threadId} \times n]$$



add $\lll 1, 16 \ggg (A, B, C);$



$$C[\text{threadId} \times n] = a[\text{threadId} \times n] + b[\text{threadId} \times n]$$

0 15

1	1	1	1	1	1	1	1
---	---	---	---	---	---	---	---

+

1	1	1	1	1	1	1	1
---	---	---	---	---	---	---	---

=

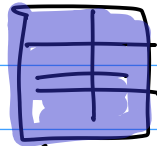
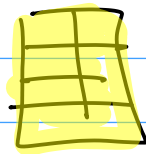
2	2	2	2	2	2	2	2
---	---	---	---	---	---	---	---

A[16]

B[16]

C[16]

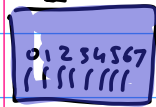
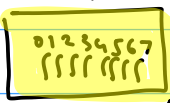
GPU RAM



$8/8 = 100\%$

$8/8 = 100\%$

add <<< 2 , 8 >>> (A, B, C);



blockId

threadId

0...1

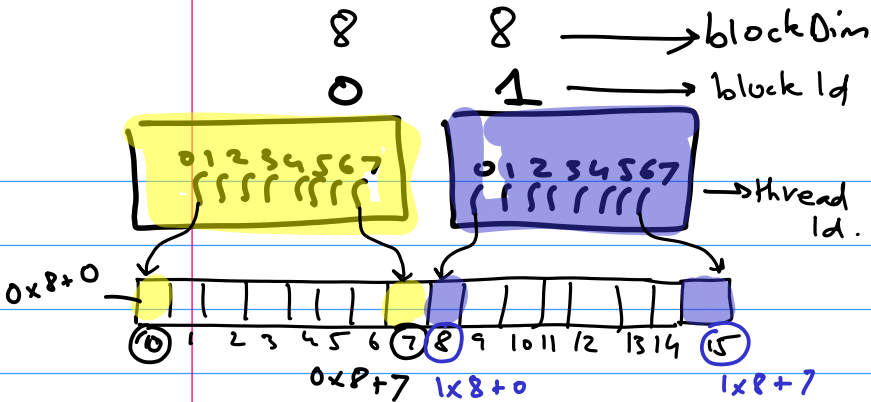
0...1

0...1

C[index 0...7]

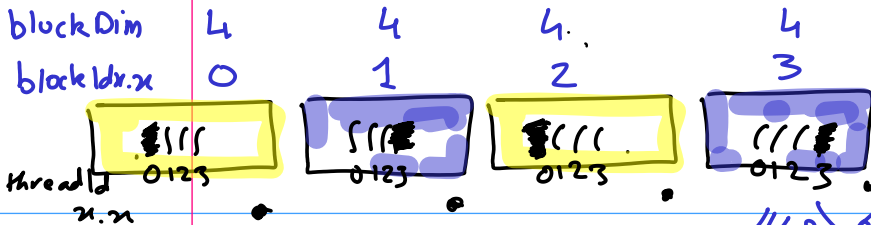
= a[index 0...7]

+ b[index 0...7]



blockId \implies Index
 threadId \implies

$$\text{index} = \text{blockId} \times \text{blockDim} + \text{threadId}$$



$$\text{index} = \text{blockIdx.x} * \text{blockDim.x} + \text{threadIdx.x}$$

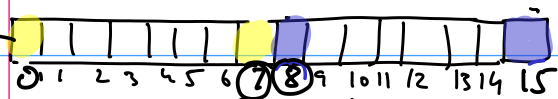
~~« 2, 8 »~~

« 4, 4 »

X, Y

0x4+0

~~0x8+0~~



~~0x8+7~~

~~1x8+0~~

~~1x8+7~~

1x4+3

2x4+0

3x4+3

$\text{for}(j=0; j < 4; j++)$
 $\text{for}(i=0; i < 4; i++)$

j

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

i

index = $j * 4 + i$

block thread

$$k \times 4 \times 2$$

$$+ j \times 4$$

$$+ i$$

— slice 0 —

Slice 1

4	5	6	7
0	1	2	3

12	13	14	15
8	9	10	11

storage

1D 0-16

Threads

1D

16x1
1x16

Threads

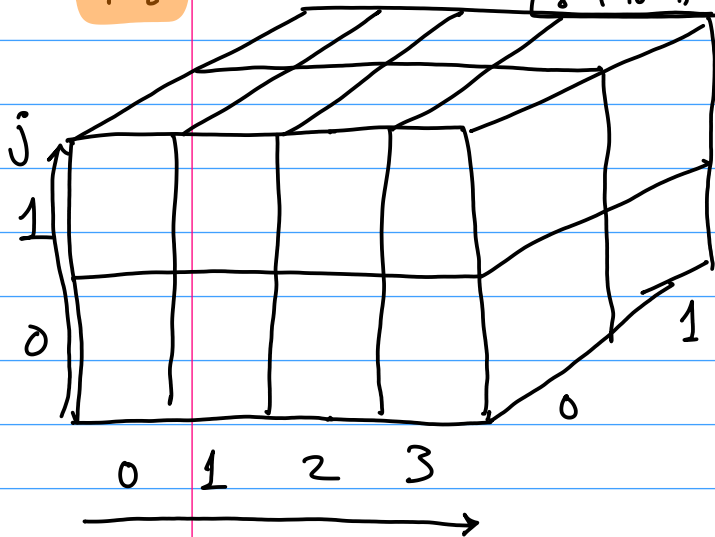
2D

4x4

threads

3D

2x4x2



$k \times 4 \times 2$

$+ j \times 4$

$+ i$

→ Cuda

Index

$\text{blockIdx.z} * \text{blockDim.x} * \text{blockDim.y}$

$+ \text{blockIdx.y} * \text{blockDim.x}$

$+ \text{threadIdx.x}$

3D

2D

1D

Data

1D

Thread

→ 1D

→ 2D

→ 3D

What if

2D

~~~~~

↓  
↓

array of size 8

8x1

=

12%

1x8

=

50%

~~4x2~~

=

100% ~~100%~~

2x2x2

=

200%

Thread Dimensions

load balance

# Blocks or Threads. Whatever. Code is running in Parallel anyways

- Let's look at sample specs for NVIDIA Kepler K40
  - Cores = 2,880
  - Multiprocessors = 15
  - Cores / multi-processor = 192
- If we make parallel blocks, 1 block will be assigned to 1 multi-processor. This means all multi-processors will be busy, but within the multi-processor, 1 core has work to do, the remaining others are free.
- If we make parallel threads and 1 block, only 1 multi-processor will be busy, the remaining will be free.
- In either case, the GPU is **under-utilized**. For maximum utilization, use both (blocks + threads)
- **Note: Above is programmer's interpretation. The actual execution model loosely follows this interpretation but has other restrictions also.**

## Case: Higher Dimensions

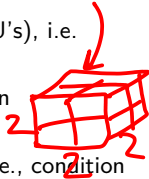
$$1D = \langle\langle\langle N, 1 \rangle\rangle\rangle$$

$$\langle\langle\langle 1, N \rangle\rangle\rangle$$

3D

- So far, we have passed single values to kernel function call.
- For higher dimensions, use:
  - $\text{dimGrid}(g_x, g_y, g_z)$  for dimensions of the grid
  - $\text{dimBlock}(t_x, t_y, t_z)$  for dimensions of the thread block
- Total threads in thread-block cannot exceed 512 (or 1024 for some GPU's), i.e. condition  $t_x \times t_y \times t_z \leq 512$  must be satisfied.
- Grid dimensions cannot exceed 32,768 in either dimension, i.e., condition  $\max(g_x, g_y) \leq 32,768$
- $\text{dimGrid}$  component must be evenly divisible by  $\text{dimBlock}$  component, i.e., condition  $\text{mod}(g_n, t_n) = 0$  must be satisfied.

$\text{dimGrid}(1,1,1)$   
 $\text{dimBlock}(2,2,2)$



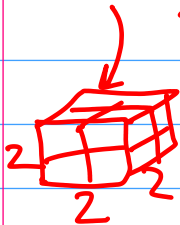
CL\_DEVICE\_MAX\_WORK\_ITEM\_DIMENSIONS:3

CL\_DEVICE\_MAX\_WORK\_ITEM\_SIZES: 1024 / 1024 / 64 (512/512/32)

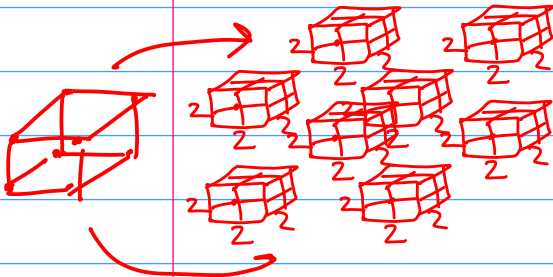
CL\_DEVICE\_MAX\_WORK\_GROUP\_SIZE: 1024 (512)

$$2D = \langle\langle\langle M, N \rangle\rangle\rangle$$

$\dimGrid(2,2,2)$   
 $\dimBlock(2,2,2)$



~~3D threads~~



60



## Case: Higher Dimensions (cont.)

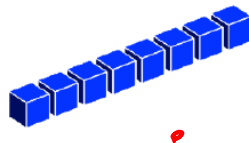
```
dim3 dimGrid(3,1,1);
dim3 dimBlock(8,1,1);

kernel<fn <<< dimGrid, dimBlock >>> (arg1, arg2, ...);
```

3 8

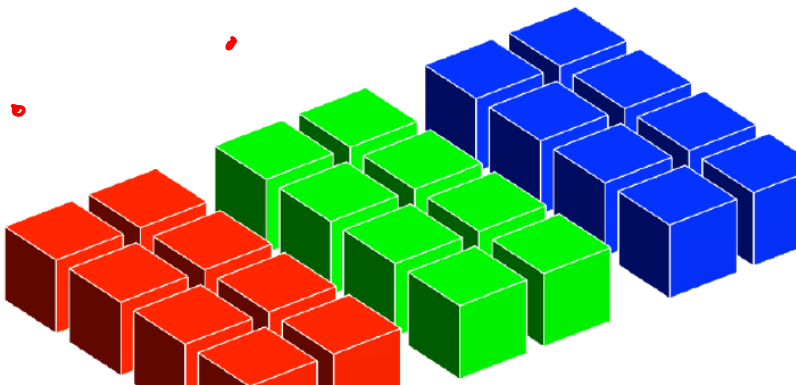
2D

dim2 dimGrid(3,1)  
dim2 dimBlock(8,1)



## Case: Higher Dimensions (cont.)

```
dim3 dimGrid(3,1,1);  
dim3 dimBlock(2,4,1);  
  
kerneln <<< dimGrid, dimBlock >>> (arg1, arg2, ...);
```



## Case: Higher Dimensions (cont.)

$$I = I_x * \text{gridDim}.x * \text{blockDim}.x + I_y$$

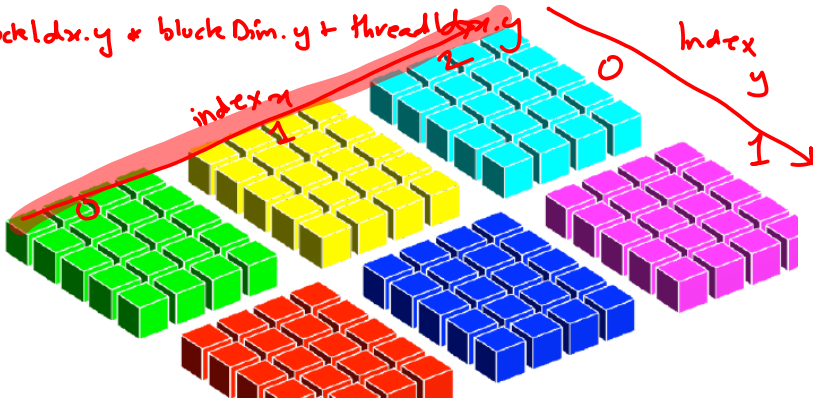
```
dim3 dimGrid(3,2,1);
dim3 dimBlock(4,5,1);

kerneln <<< dimGrid, dimBlock >>> (arg1, arg2, ...);
```

 $+ I_y$ 

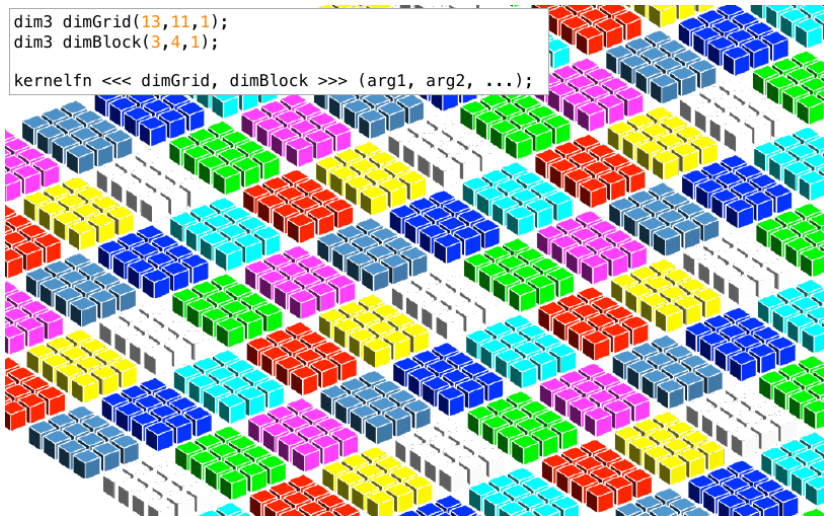
$$I_x = \text{blockIdx}.x * \text{blockDim}.x + \text{threadIdx}.x$$

$$I_y = \text{blockIdx}.y * \text{blockDim}.y + \text{threadIdx}.y$$



## Case: Higher Dimensions (cont.)

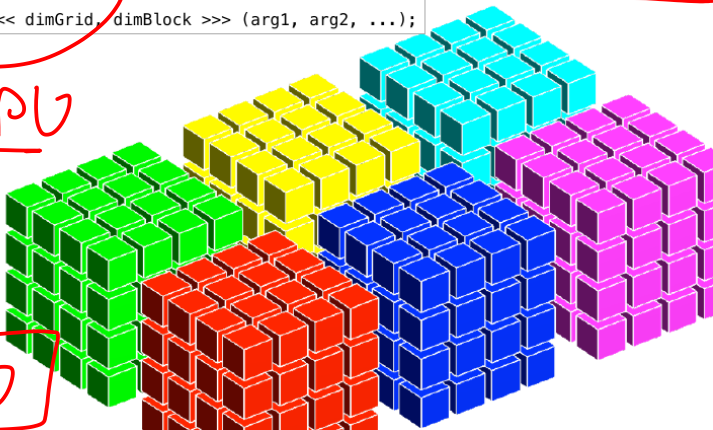
```
dim3 dimGrid(13,11,1);  
dim3 dimBlock(3,4,1);  
  
kernelnf <<< dimGrid, dimBlock >>> (arg1, arg2, ...);
```



## Case: Higher Dimensions (cont.)

```
dim3 dimGrid(3,2,1);  
dim3 dimBlock(4,4,4);  
kerneln <<< dimGrid, dimBlock >>> (arg1, arg2, ...);
```

threads

1 GPU

M GPU

② Host Code → Kernel

① Cuda Drivers.

↳ NVIDIA

↳ Proprietary.

↙  
Open Source

~~Black list.~~ Nouveau

(White list)  
NVIDIA

# Combining both Concepts of Blocks and Threads

## Formula to identify a point

```
int index = blockIdx.x * blockDim.x + threadIdx.x;
```

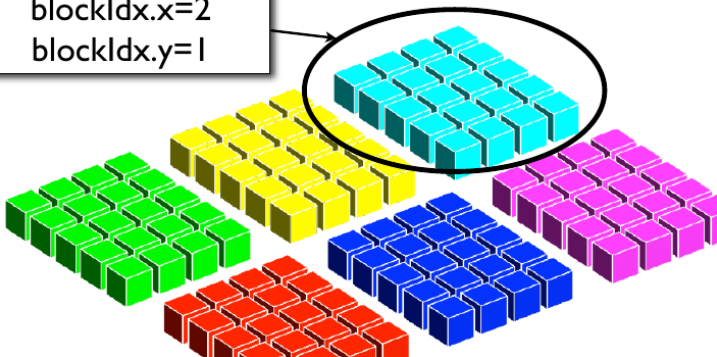
- `blockIdx.x` = Index of thread block in grid
- `blockDim.x` = Number of threads in 1D thread-block
- `threadIdx.x` = Index of thread in 1D thread block
- Each thread execute kernel code is automatically provided the following variables
  - `threadIdx.x`, `threadIdx.y`, `threadIdx.z`
  - `blockDim.x`, `blockDim.y`, `blockDim.z`
  - `blockIdx.x`, `blockIdx.y`

\* `gridDim.x` , `gridDim.y` , `gridDim.z`

# Combining both Concepts of Blocks and Threads (cont.)

```
dim3 dimGrid(3,2,1);  
dim3 dimBlock(4,5,1);  
  
kernelFn <<< dimGrid, dimBlock >>> (arg1, arg2, ...);
```

blockIdx.x=2  
blockIdx.y=1

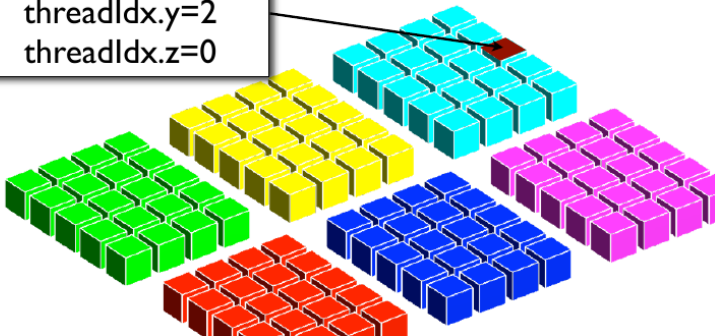




# Combining both Concepts of Blocks and Threads (cont.)

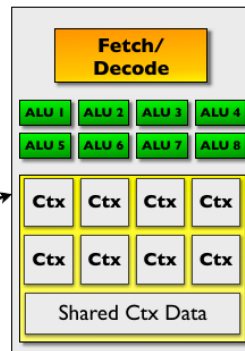
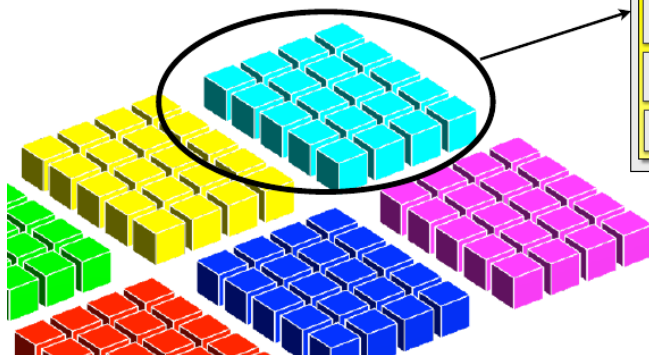
```
dim3 dimGrid(3,2,1);  
dim3 dimBlock(4,5,1);  
  
kernelFn <<< dimGrid, dimBlock >>> (arg1, arg2, ...);
```

threadIdx.x=3  
threadIdx.y=2  
threadIdx.z=0



# Higher Dimensions: Execution Model Revisited

All the threads in an individual thread-block are handled by the same streaming multiprocessor.



## Higher Dimensions: Execution Model Revisited (cont.)

In this example batches of 8 threads will be processed concurrently

