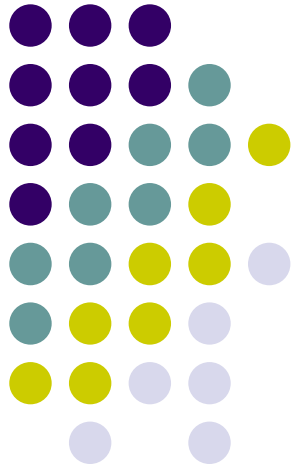
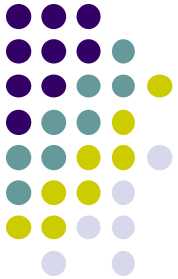
Practical Parallel Computing (実践的並列コンピューティング) 2021 No. 9

Part2: GPU (3)
May 13, 2021



Toshio Endo
School of Computing & GSIC
endo@is.titech.ac.jp

Overview of This Course



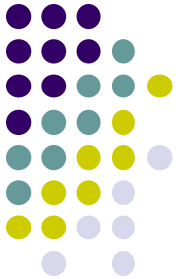
- Part 0: Introduction
  - 2 classes
- Part 1: OpenMP for shared memory programming
  - 4 classes
- Part 2: GPU programming
  - 4 classes ← We are here (3/4)
  - OpenACC (1.5 classes) and CUDA (2.5 classes)
- Part 3: MPI for distributed memory programming
  - 3 classes

Comparing OpenMP/OpenACC/CUDA

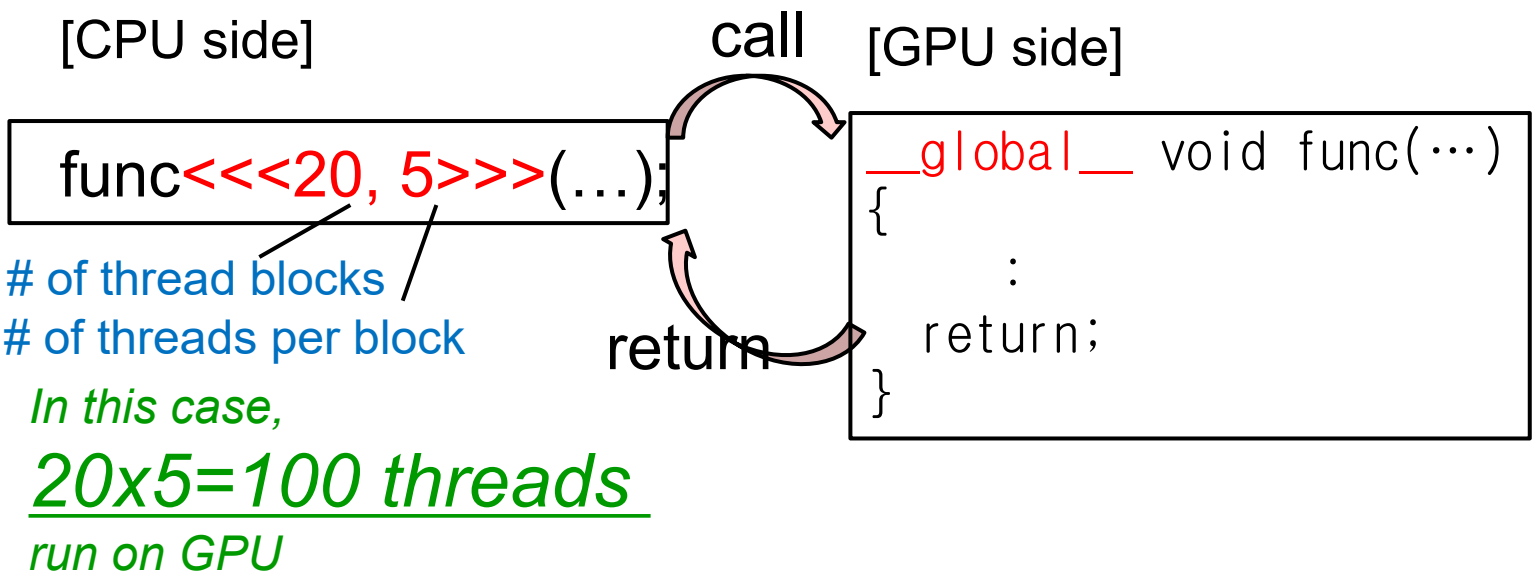
	OpenMP	OpenACC	CUDA
Processors	CPU	CPU+GPU	
File extension	.c, .cc		.cu
To start parallel (GPU) region	#pragma omp parallel	#pragma acc kernels	func<<<…, …>>>()
To specify # of threads	export OMP_NUM _THREADS=…	(num_gangs, vector_length etc)	
Desirable # of threads	# of CPU cores or less	# of GPU cores or “more”	
To get thread ID	omp_thread_num()	-	blockIdx, threadIdx
Parallel for loop	#pragma omp for	#pragma acc loop	-
Task parallel	#pragma omp task	-	-
To allocate device memory	-	#pragma acc data	cudaMalloc()
To copy to/from device memory	-	#pragma acc data #pragma acc update	cudaMemcpy()
Functions on GPU	-	#pragma acc routine	__global__,__device__

※ “# of XXX” = “The number of XXX”

Calling A GPU Kernel Function from CPU

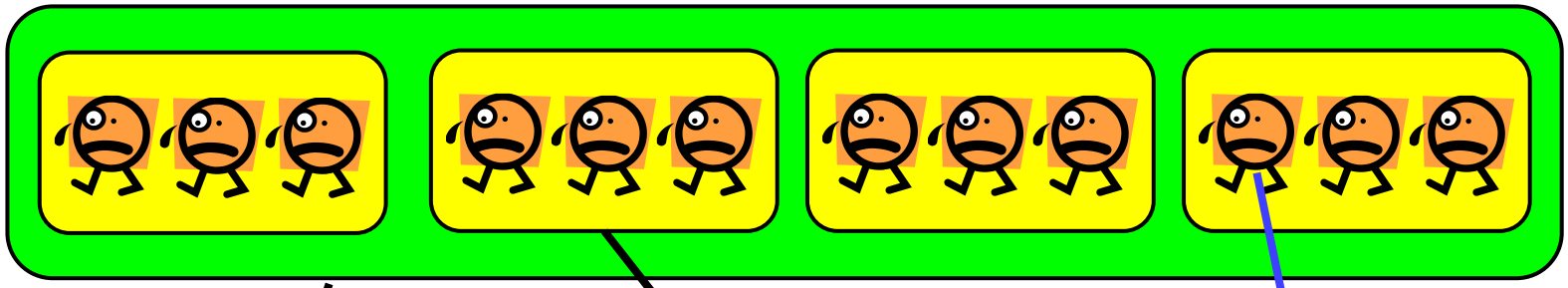


- A region executed by GPU must be a distinct function
  - called a GPU kernel function



Threads in CUDA

Specify 2 numbers (at least) for number of threads, when calling a GPU kernel function



A grid

A thread block

A thread

cf) func <<< 4, 3 >>> (); → 12 threads

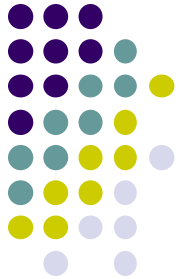
Number of thread blocks = gridDim

Number of threads per block = blockDim

The reason is related to GPU hardware

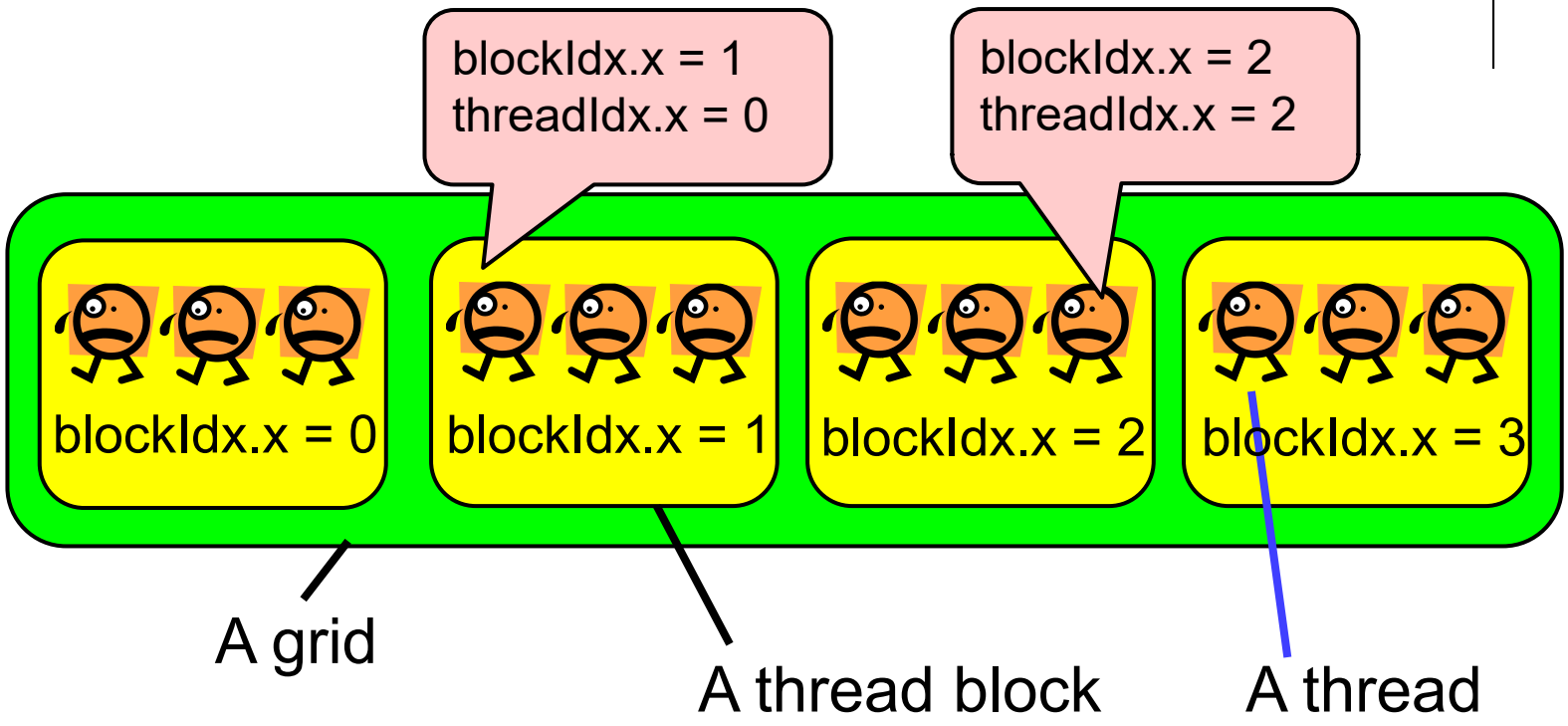
Thread block ⇔ SMX, Thread ⇔ CUDA core

To See Who am I



- By reading the following special variables, each thread can see its thread ID in GPU kernel function
- My ID
  - blockIdx.x: Index of the block the thread belong to (≧0)
  - threadIdx.x: Index of the thread (inside the block) (≧0)
- Number of thread/blocks
  - gridDim.x: How many blocks are running
  - blockDim.x: How many threads (per block) are running

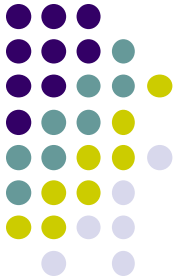
Thread Block ID, Thread ID



For every thread, gridDim.x = 4, blockDim.x = 3

Note: In order to see the entire sequential ID, we should compute
blockIdx.x * blockDim.x + threadIdx.x

The Case of add-cuda Sample



- /gs/hs1/tga-ppcomp/21/add-cuda
- We want to do

```
for (i = 0; i < 100; i++) { DA[i] += DB[i]; }
```

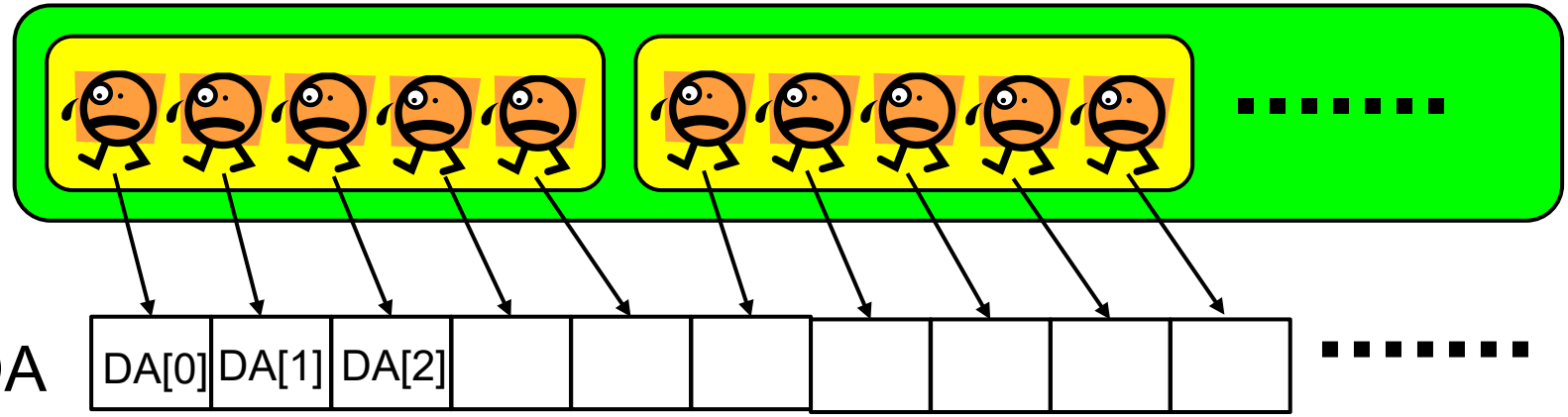
[CPU side]

```
add<<<20, 5>>>(…);
```

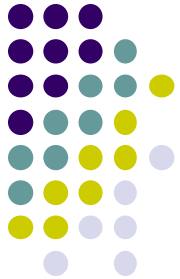
20x5=100 threads will execute add function

[GPU side]

```
__global__ void add(int *DA, int *DB)
{
    int i = blockIdx.x * blockDim.x + threadIdx.x;
    DA[i] += DB[i];
    return;
}
```



How Number of Threads is Designed? (1)

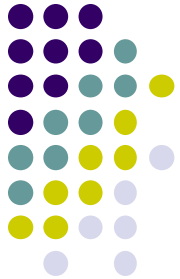


On CUDA, a different strategy is required from on OpenMP

- On OpenMP, number of threads (OMP_NUM_THREADS) should be ≦ CPU cores (or hyper threads)
  - The number is basically determined by hardware
  - ≦14 on q_node node, ≦56 on f_node

- On CUDA, it is better to use number of thread ≧ GPU cores
  - ≧ 3584 on TSUBAME3's P100 GPU
  - You can use >1,000,000 threads!

How Number of Threads is Designed? (2)



We have to deicide 2 numbers

<<<block number, block size>>>

A better way would be

(1)We decide total number of threads P

(2)We tune each block size BS

- Good candidates are 16, 32, 64, … 1024

(3)Then block number is P/BS

- We consider indivisible cases later

“mm” sample: Matrix Multiply (related to [G2])

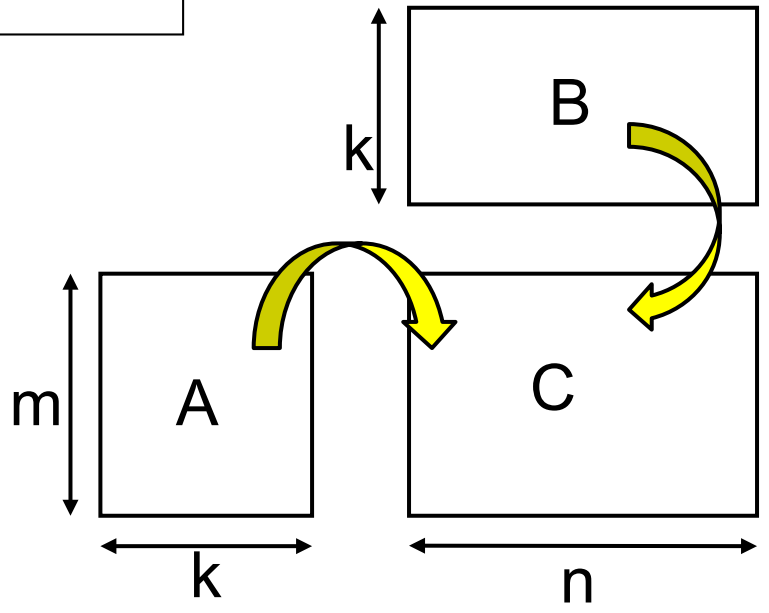
CUDA versions are at
- /gs/hs1/tga-ppcomp/21/mm-v1-cuda/
- /gs/hs1/tga-ppcomp/21/mm-cuda/

A: a (m × k) matrix, B: a (k × n) matrix

C: a (m × n) matrix

C ← A × B

- Supports variable matrix size
- Execution:./mm [m] [n] [k]



On CUDA, We need to design
(1) How we parallelize computation
(2) How we put data on host memory & device memory

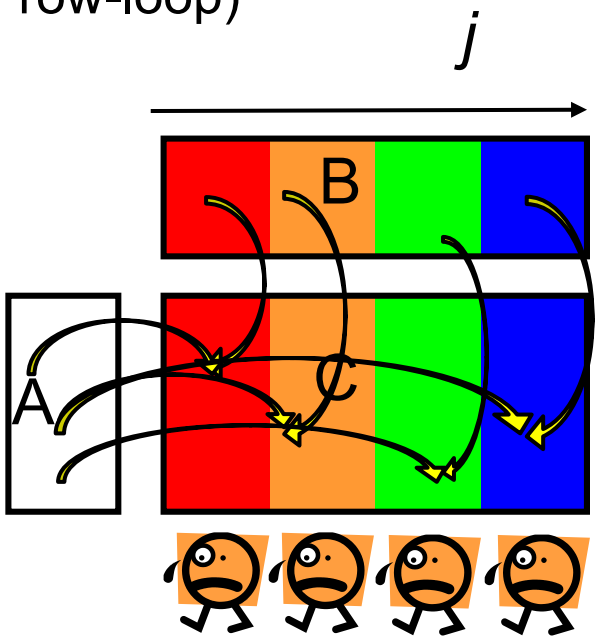
How We Parallelize Computation

In mm, we can compute different C elements in parallel

•On the other hand, it is harder to parallelize dot-product loop

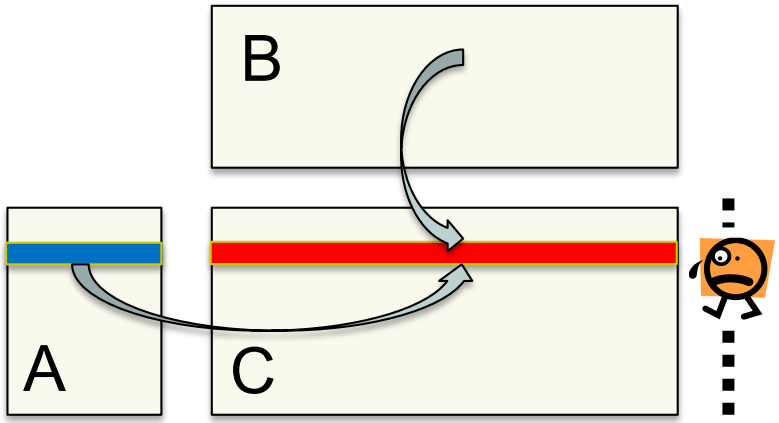
OpenMP

- Parallelize column-loop (or row-loop)



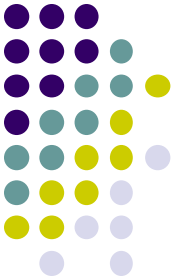
CUDA (mm-v1-cuda)

- We can create many threads
- 1 thread computes 1 row
  - We use m threads



※ This is not the unique way

Parallelism in mm-v1-cuda



- It is ok to make >1000, >10000 threads on CUDA
- We use *m* threads for *m* rows computation

```
add<<<m/BS, BS>>>(.....);
```

gridDim

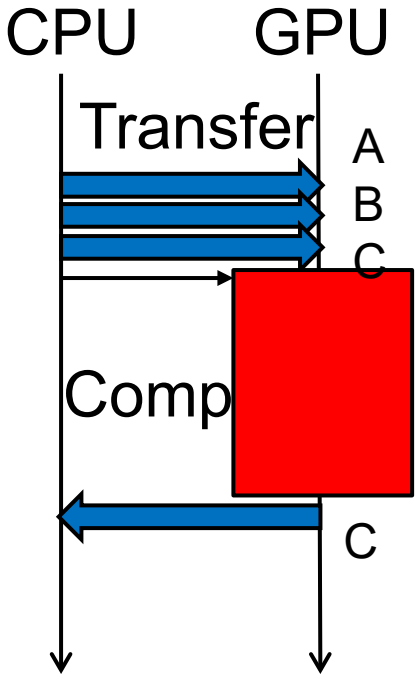
blockDim (BS=16 in this sample)

1 element for 1 row → No need of "i" loop in this sample

Note1: <<<m, 1>>> also works, but speed is not good
<<<1, m>>> causes an error if m>1024 (CUDA's rule)

Note2: To support the case m is indivisible by BS, we should use <<<(m+BS-1)/BS, BS>>>
→But # of threads may be larger m. "Extra" threads (id≧m) should not work. See mm-v1-cuda/mm.c

Data Transfer in mm-v1-cuda



(1) A, B, C are copied from CPU to GPU

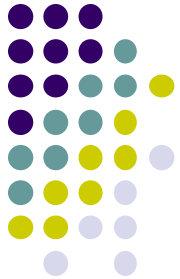
- cudaMemcpy(DA, A, …) …

(2) Computation is done on GPU

(3) C is copied from GPU to CPU

- cudaMemcpy(C, DC, …)

Notes in Time Measurement



- clock(), gettimeofday() must be called from CPU
- For accurate measurement, we should call cudaDeviceSynchronize() before measurement
  - Actually GPU kernel function call and cudaMemcpy(HostToDevice) are non-blocking

Speed of mm-v1-cuda

- Measured with a P100 GPU on TSUBAME3
  - ./mm 1000 1000 1000 and so on
- The program outputs 2 speeds
  - Speed with data transfer costs → shown on the table
  - Speed without data transfer costs

m=n=k	mm-acc (Gflops)	mm-v1-cuda (Gflops)
1000	143	14
2000	173	27
4000	164	50
6000	138	70
8000	137	85

Why slow?

Discussion on Speed (related to [G2])

mm-v1-cuda is slower than mm-acc!

- In mm-acc, i-loop and j-loop has “loop independent”
  → $m \times n$ elements are computed in parallel

- In mm-v1-cuda, we use m threads in total
  - → We should use more threads on a GPU!
  - At least, $\geqq$ 3,584 = number of CUDA cores
  - We see 8000 threads are still insufficient (slower than mm-acc)

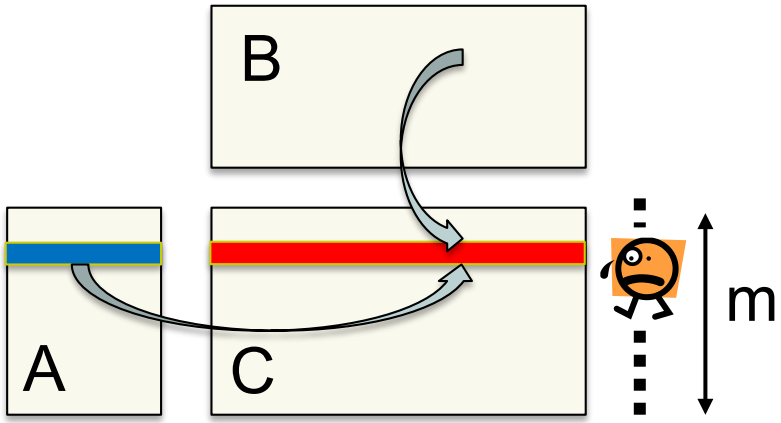
How to Use More Threads in mm

In mm, computation of each C element is independent with each other

mm-v1-cuda

- 1 thread computes 1 row

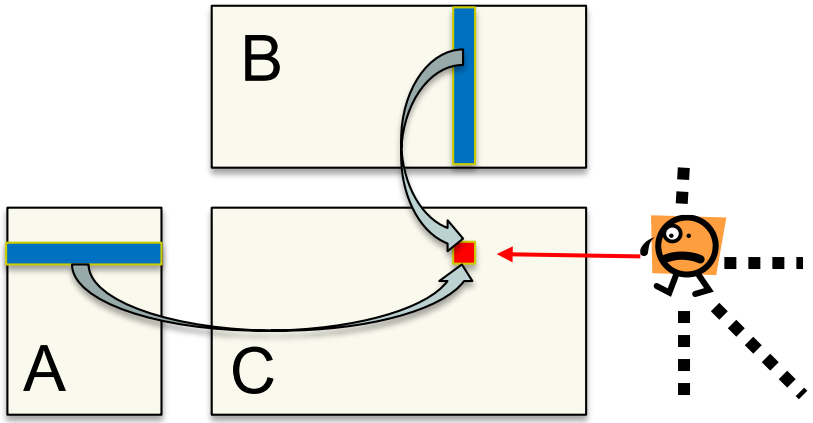
➔ We use m threads



mm-cuda

- 1 thread computes 1 element

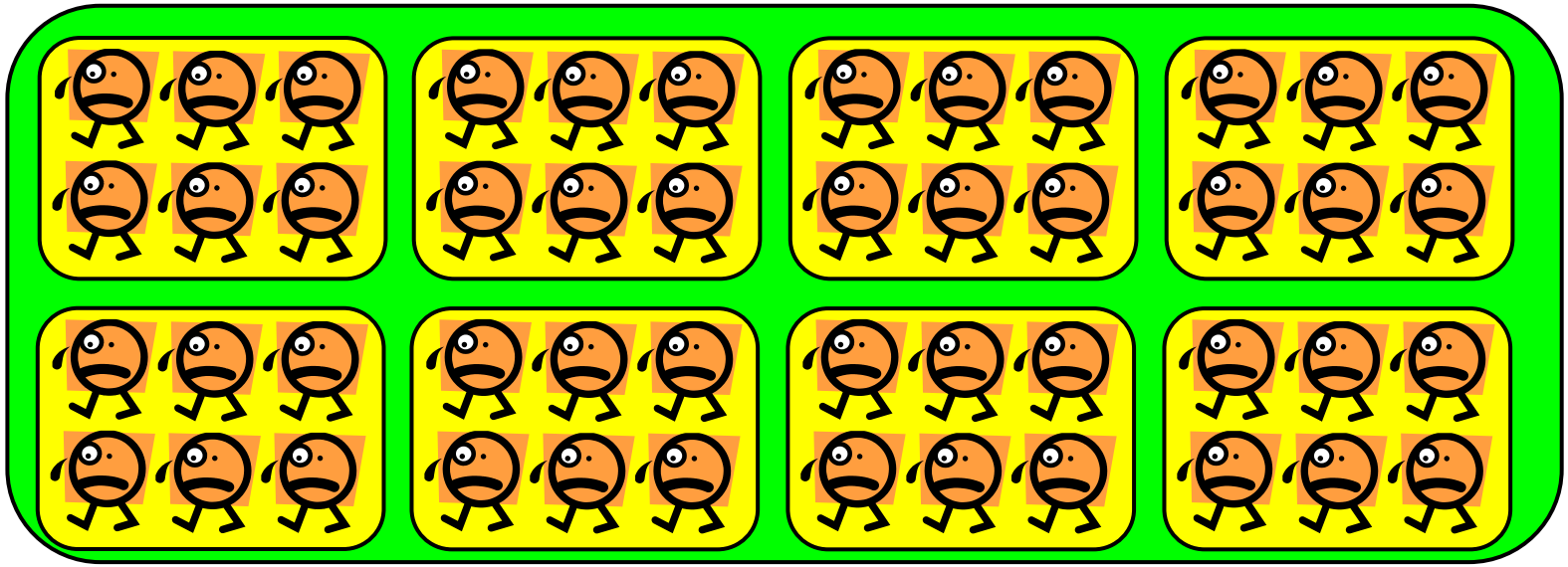
➔ We use m × n threads !!



Creating Threads with 2D/3D IDs

- Now we want to make m*n (may be >1,000,000) threads
  - <<<(m*n)/BS, BS>>> is ok, but coding is bothersome
- On CUDA, gridDim and blockDim may have "dim3" type, 3D vector structure with x, y, z fields

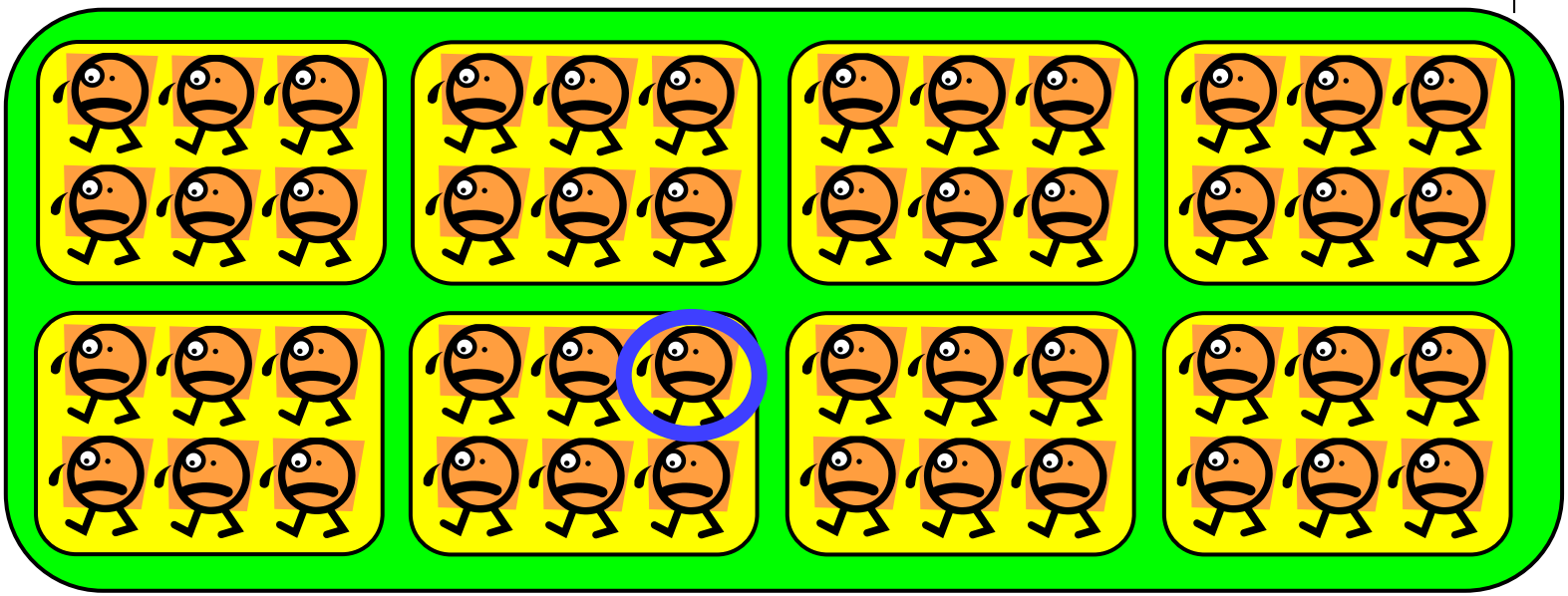
cf) func <<< dim3(4,2,1), dim3(3,2,1) >>> (); → 48 threads



※ This example is the case of 2D (Z dimensions are 1)

Thread IDs in multi-dimensional cases

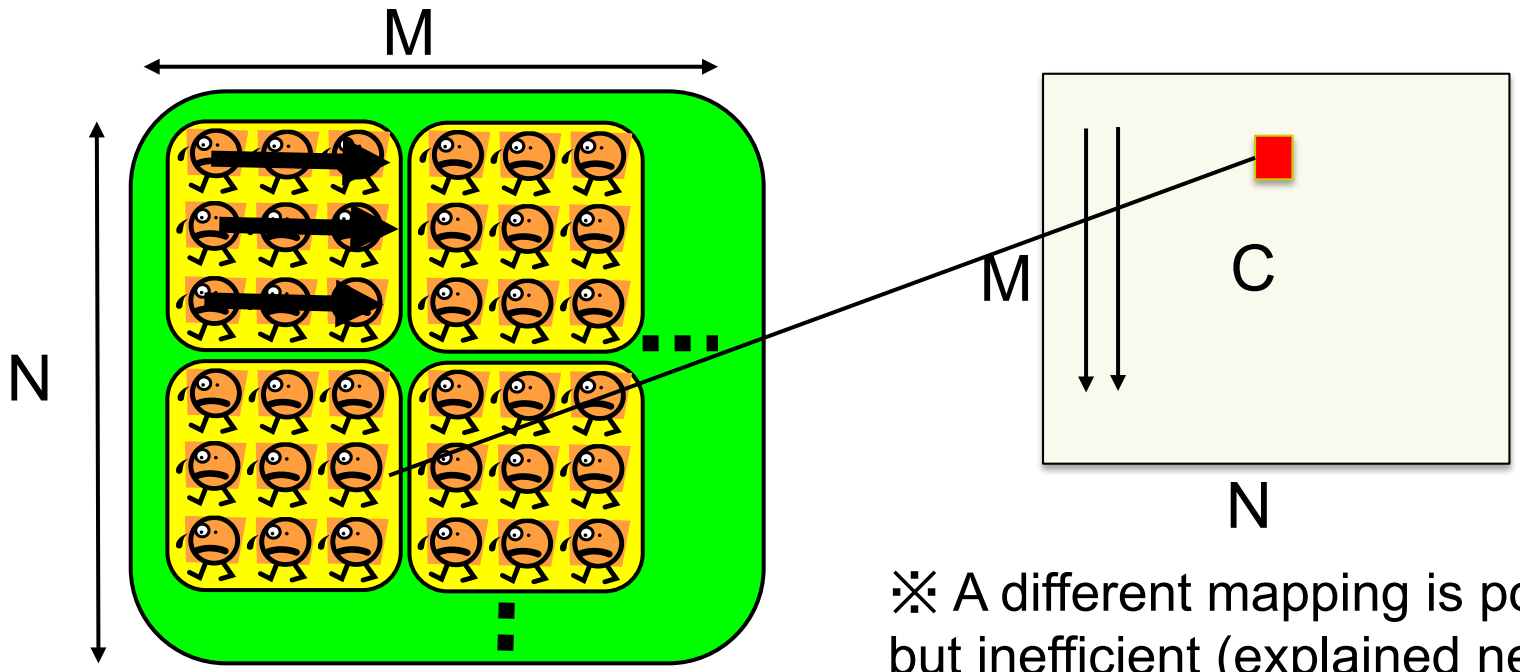
In the case of func <<< dim3(4,2,1), dim3(3,2,1) >>> ();



- For every thread,
  gridDim.x=4, gridDim.y=2, gridDim.z=1
  blockDim.x=3, blockDim.y=2, blockDim.z=1
- For the thread with blue mark,
  blockIdx.x=1, blockIdx.y=1, blockIdx.z=0
  threadIdx.x=2, threadIdx.y=0, threadIdx.z=0

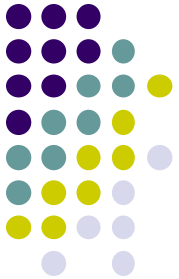
Threads in mm-cuda Sample

- The total number of threads are m*n
- How do we determine gridDim, blockDim?
  - <<<m, n>>> does not work for constraints explained later ☹
- Here, we use fixed blockDim (x=16, y=16 → 256 threads per block)
  - Then gridDim is computed from M, N
- x is mapped to row index, y is mapped to column index (※)



※ A different mapping is possible, but inefficient (explained next time)

Code in mm-cuda



gridDim blockDim

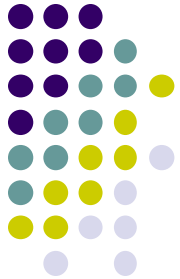
```
matmul_kernel<<<dim3(m / BS, n / BS, 1), dim3(BS, BS, 1)>>>
        (DA, DB, DC, m, n, k);
```

BS=16 in this sample
Actually, we use rounding up

In matmul_kernel function,

```
        :
    j = blockIdx.y * blockDim.y + threadIdx.y;
    i = blockIdx.x * blockDim.x + threadIdx.x;
        :   This thread computes Cij ← Only k loop
```

About Programming Efforts



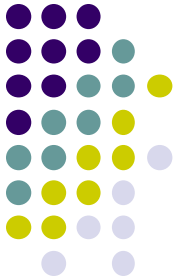
- So far, mm-cuda with m X n threads has been explained
  - How fast is it? Please measure it (Related to [G2])

- On the other hand, codes are more complex than mm-acc

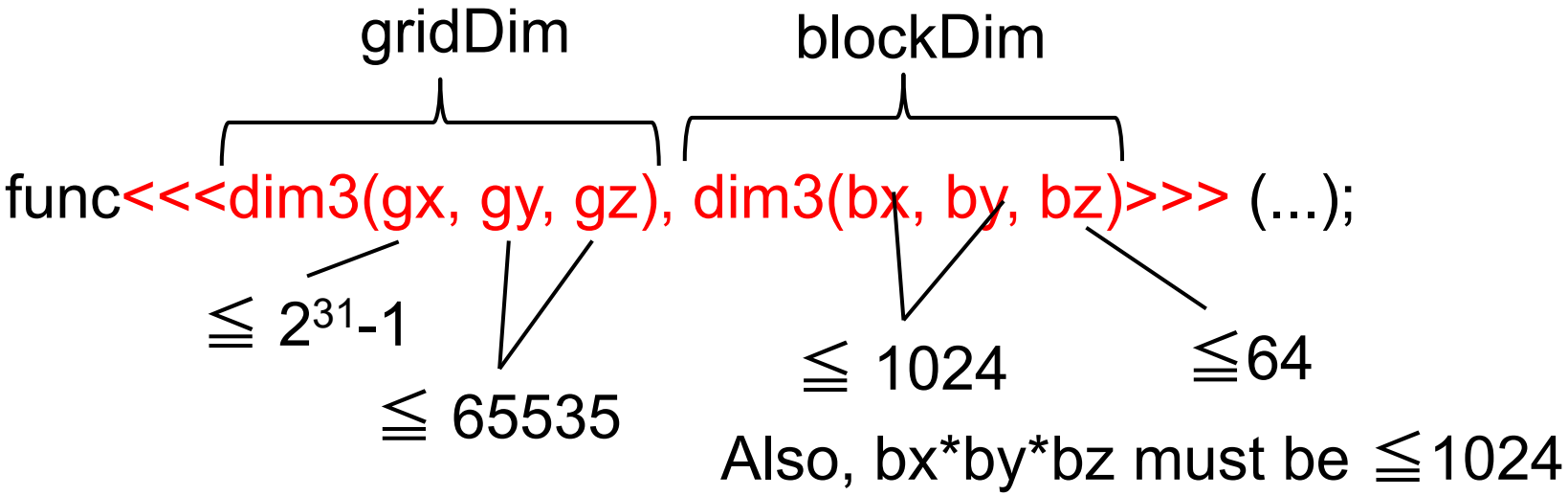
  Programmer have to
  - Call cudaMalloc, cudaMemcpy
  - Determine the number of blocks, threads
  - Determine the structure of for-loop
    - We have to change code largely when parallelization method is changed
    - In OpenACC, adding “#pragma acc loop independent” works

    :

CUDA Rules on Number of Threads



func<<<A, B>>> (...); *(A, B are integers)* is same as
func<<<dim3(A,1,1), dim3(B,1,1)>>> (...);

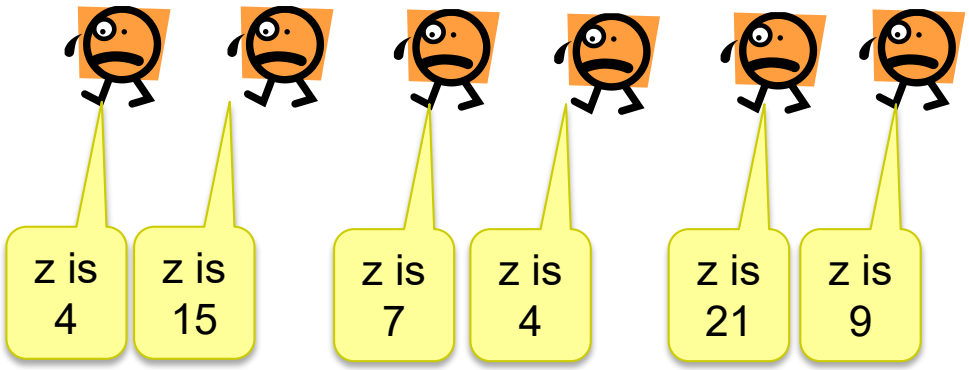


BlockDim has severe limitation ☹

Cf) <<<m, n>>> causes an error if n>1024 ☹

Rules for Memory/Variables

- Variables declared in GPU kernel functions are "thread private"

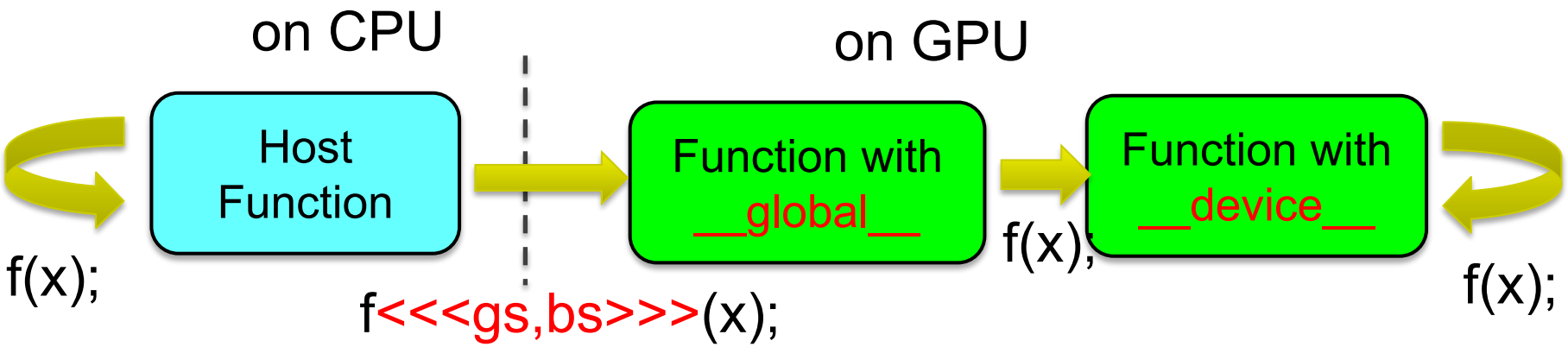


- Device memory is shared by all CUDA threads
  - Be careful to avoid race condition problem (multiple threads write same address)
  - Reading same address is ok
- Do not forget host memory and device memory are separated

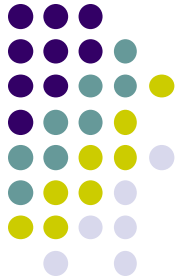
Two Types of GPU Kernel Functions

1) Functions with __global__ keyword
   - "Gateway" from CPU
   - Return value type must be "void"
2) Function with __device__ keyword
   - Callable only from GPU
   - Can have return values
   - Recursive call is OK

→ In OpenACC, #pragma acc routine

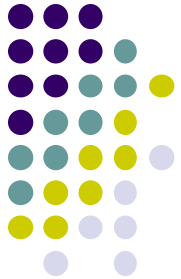


What Can be Done in GPU Functions?



- Basic computations (+, -, *, /, %, &&, ||...) are OK
- if, for, while, return are OK
- Device memory access is OK
- Host memory access is NG
- Calling host functions is NG
- Calling most of functions in libc or other libraries for CPUs are NG
  - Several mathematical functions, sin(), sqrt()… are OK
  - printf() is OK
  - Calling malloc()/free() on GPU is OK, if the size must be small
    - Usually, use cudaMalloc() on CPU

Assignments in GPU Part (Abstract)



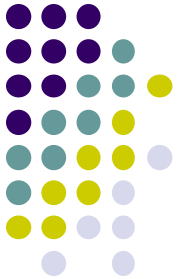
Choose one of [G1]—[G3], and submit a report

Due date: May 27 (Thursday)

[G1] Parallelize “diffusion” sample program by OpenACC or CUDA

[G2] Evaluate speed of “mm-acc” or “mm-cuda” in detail

[G3] (Freestyle) Parallelize *any* program by OpenACC or CUDA.



Next Class:

- GPU Programming (4)
  - diffusion using CUDA
  - Discussion on speed and GPU hardware

- Also please note due date of OpenMP assignment is May 13 (Today!)