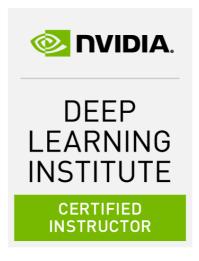
#### **CUDA Programming - 2D Thread Organization**

by **Dr. Nileshchandra Pikle**Assistant Professor
&

"A certified CUDA instructor by NVIDIA"



### Thread Organization Extended to 2D

- Threads can be organized in 2D or 3D
- **dim3** is an integer vector type that can be used in CUDA code.
- Its most common application is to pass the grid and block dimensions in a kernel invocation.
- Eg.

dim3 grid(x, y, z)

Grid is a vector of **3 dimension** and of type **dim3** 

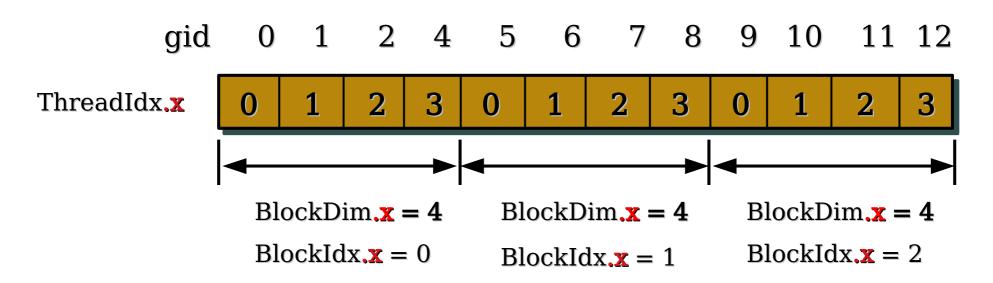
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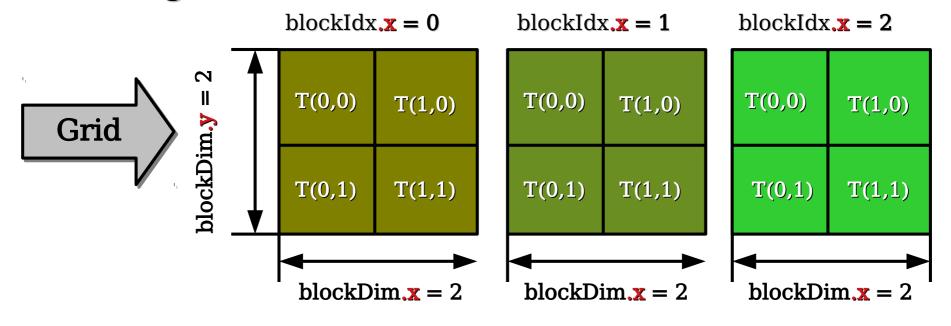
```
dim3 grid(512); // 512 x 1 x 1
dim3 block(1024, 1024); // 1024 x 1024 x 1
fooKernel<<< grid, block >>>();
```

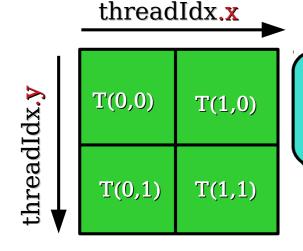
1D grid and 1D block

int gid = blockIdx.x \*blockDim.x + threadIdx.x;

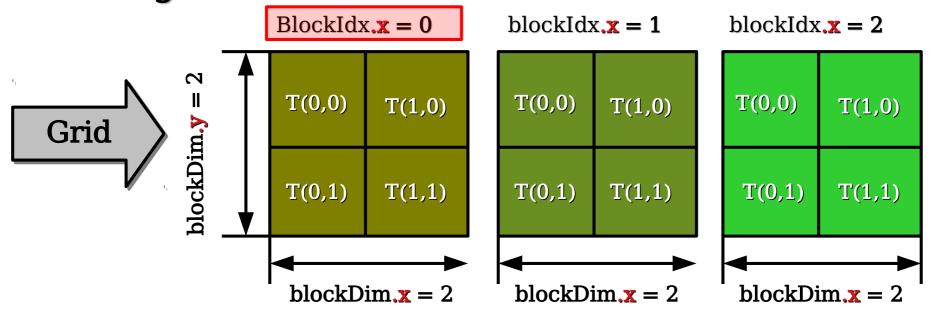


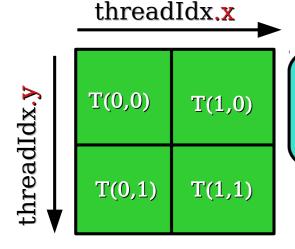
1D grid and 2D block



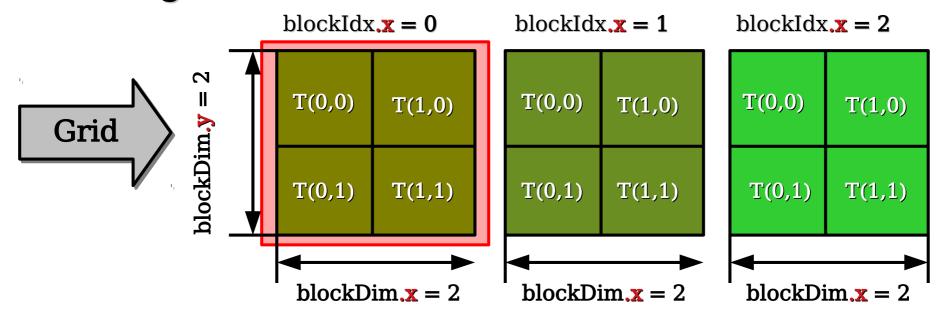


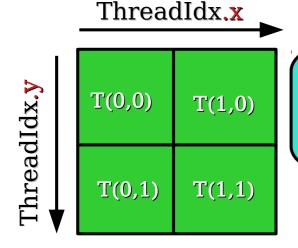
1D grid and 2D block





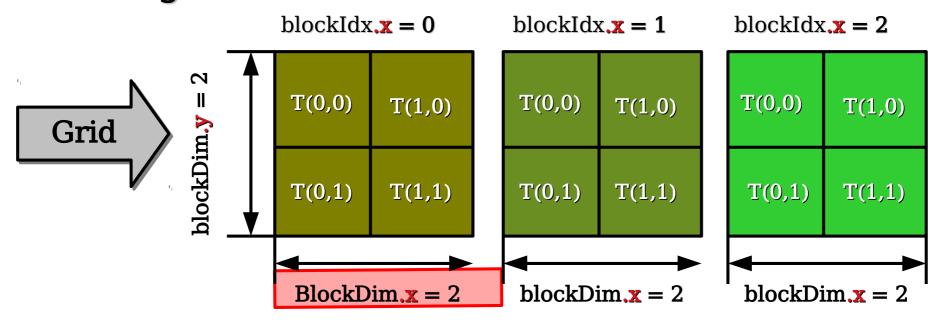
1D grid and 2D block

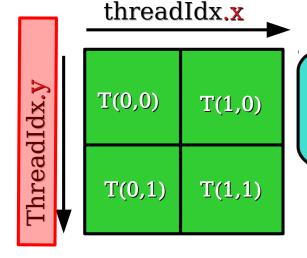




int gid = blockIdx.x \* blockDim.x \*blockDim.y
threadIdx.y \* blockDim.x + threadIdx.x;

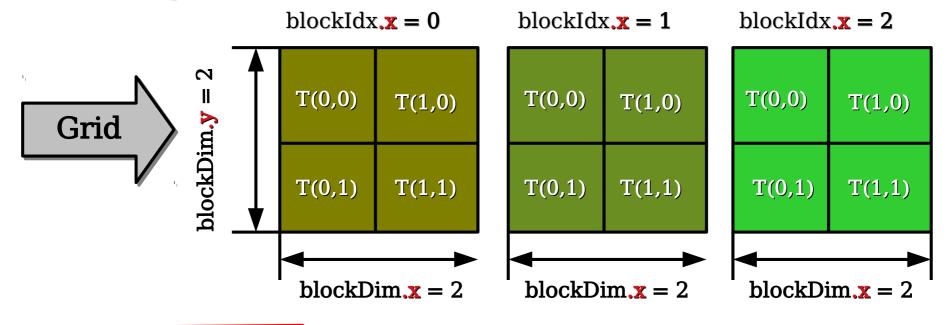
1D grid and 2D block

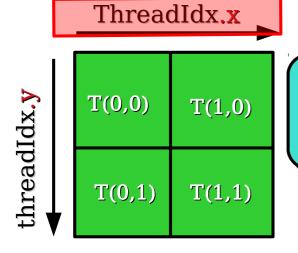




int gid = blockIdx.x \* blockDim.x \*blockDim.y
 threadIdx.y \* blockDim.x + threadIdx.x;

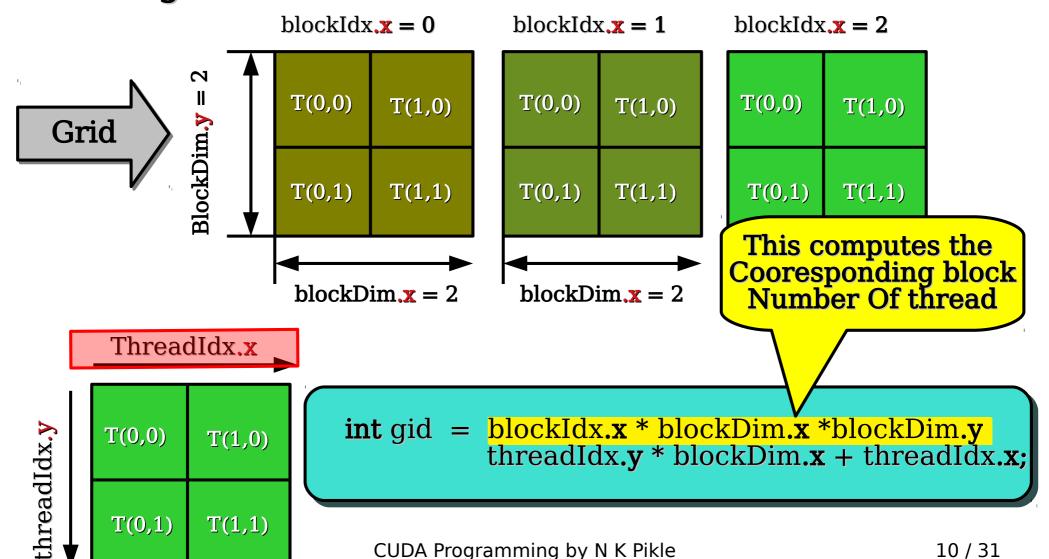
1D grid and 2D block



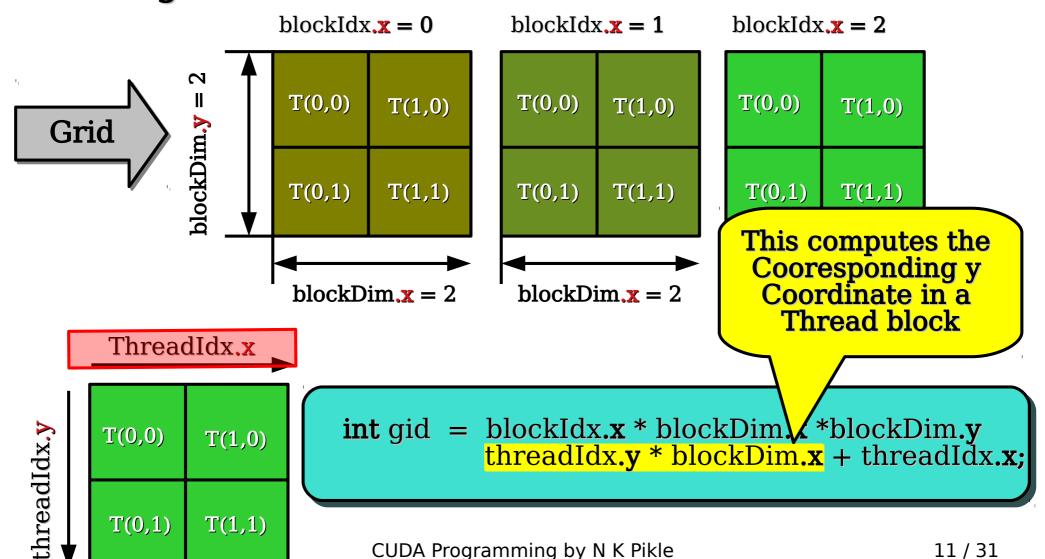


int gid = blockIdx.x \* blockDim.x \*blockDim.y
 threadIdx.y \* blockDim.x + threadIdx.x;

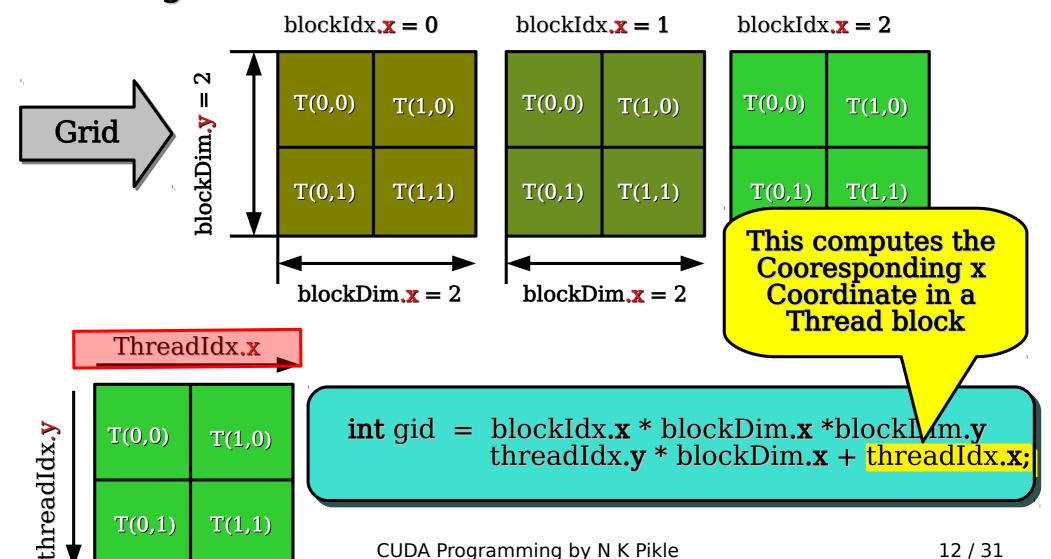
1D grid and 2D block



1D grid and 2D block



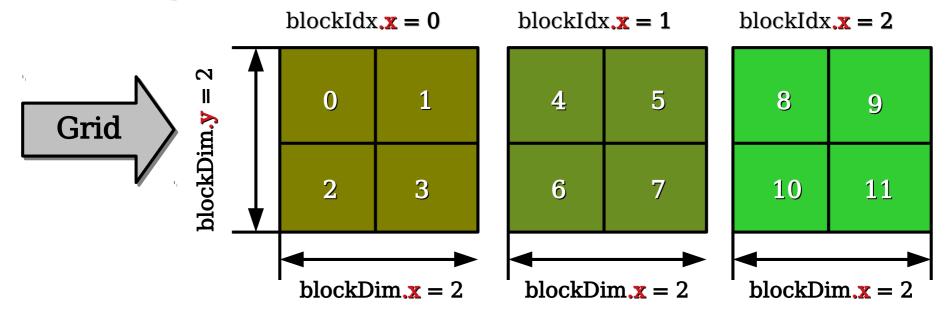
1D grid and 2D block

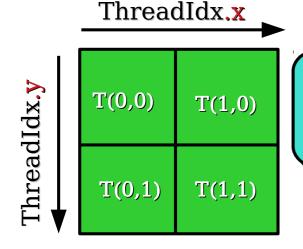


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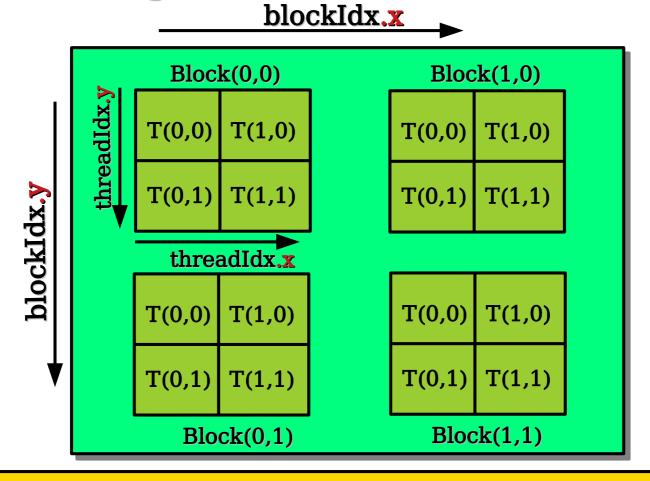
12 / 31

1D grid and 2D block



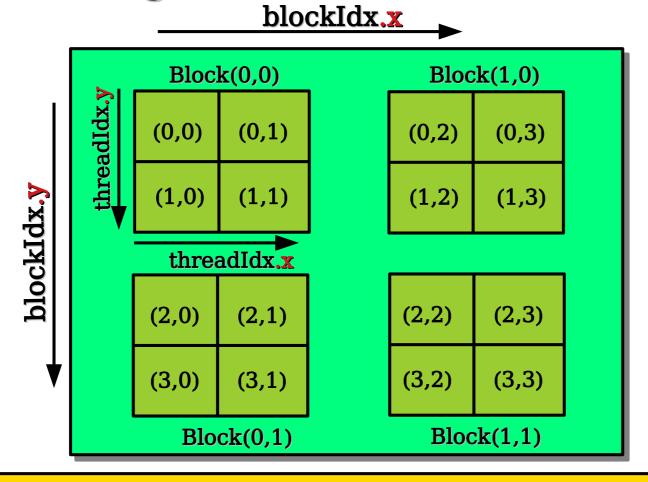


## Thread Organization 2D block & 2D grid



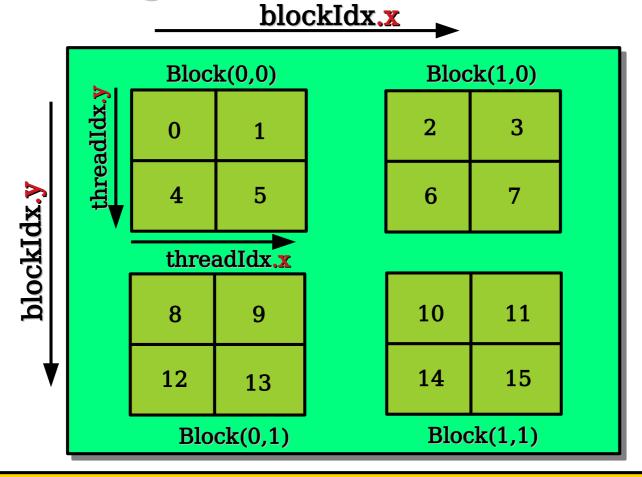
```
int row = threadIdx.y + blockIdx.y * blockDim.y
int col = threadIdx.x + blockIdx.x * blockDim.x
```

## Thread Organization 2D block & 2D grid



```
int row = threadIdx.y + blockIdx.y * blockDim.y
int col = threadIdx.x + blockIdx.x * blockDim.x
```

## Thread Organization 2D block & 2D grid



int gid = row \* blockDim.x \* gridDim.x + col

#### Convolution mask

1	1	1
1	1	1
1	1	1

#### **Input Image**

1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1

To take care of boundary conditions either apply zeroPadding or avoid the out of range computations

0	0	0	0	0	0	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	4	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	6	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	6	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	1	6	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} \max \{s\}[t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	1	1	4	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	1	1	1	0
0	6	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	1	1	1	0
0	1	9	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	1	1	1	0
0	1	1	9	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	1	1	1	0
0	1	1	1	9	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

0	0	0	0	0	0	0
0	1	1	1	1	1	0
0	1	1	1	1	6	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

# Convolution mask

1	1	1
1	1	1
1	1	1

$$O[i][j] = \sum_{s=-a}^{+a} \sum_{t=-a}^{+a} mask[s][t] * A[i+s][j+t]$$

#### Final Output Image

4	6	6	6	4
6	9	9	9	6
6	9	9	9	6
6	9	9	9	6
4	6	6	6	4

Refer program 2D\_Convolution.cu

#### Summary

- 1. CUDA programming model Heterogeneous Computing
- 2. CUDA Thread organization: Threads grouped together to form blocks and blocks to grid.-Threads can be organized as 1D, 2D or 3D.
- 3. Case studies: To understand basic CUDA programming model, thread and data mapping in 1D and 2D