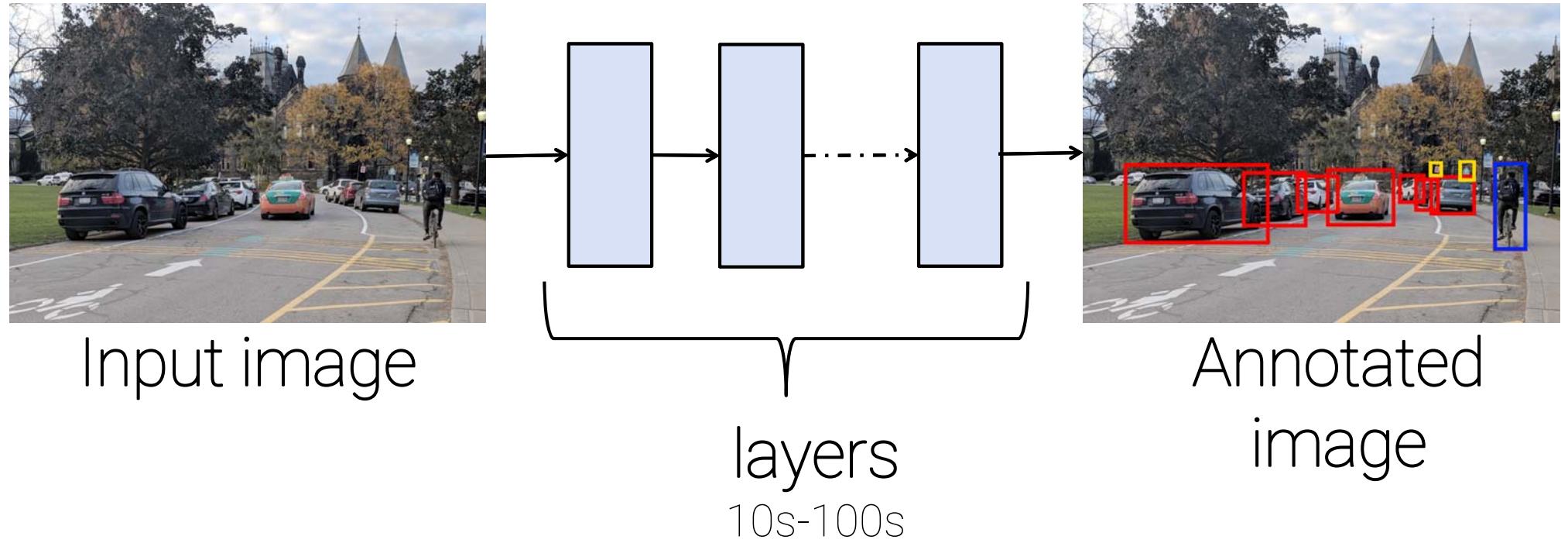
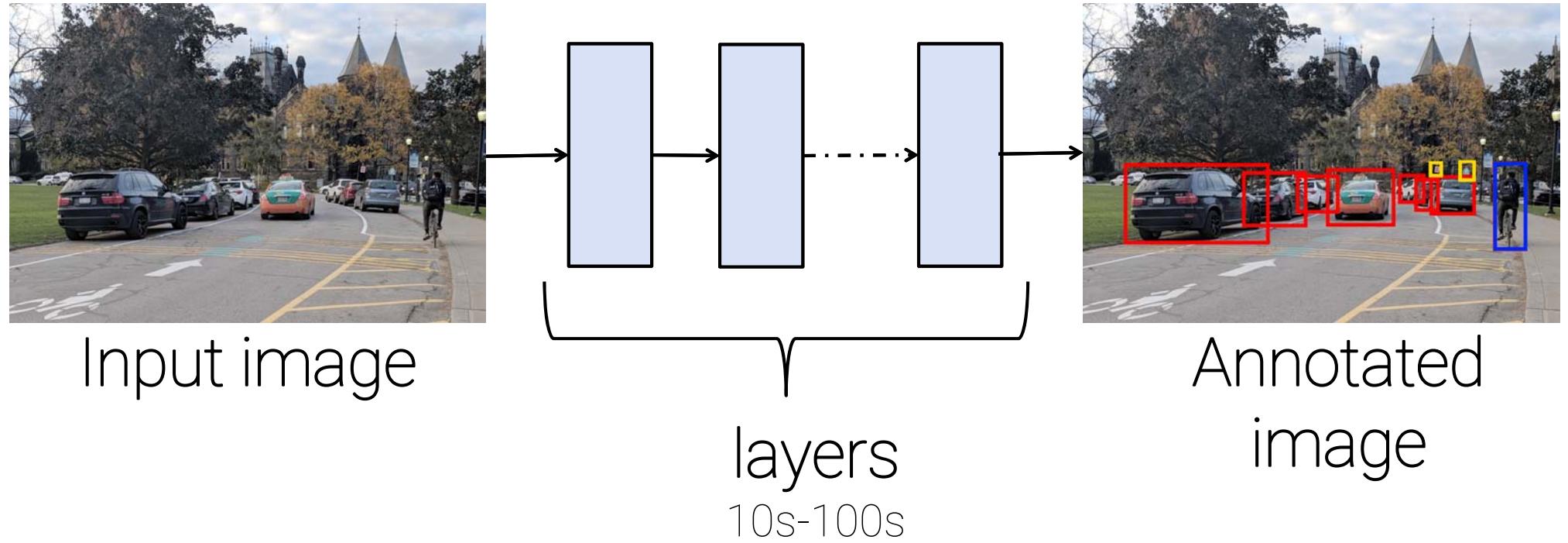




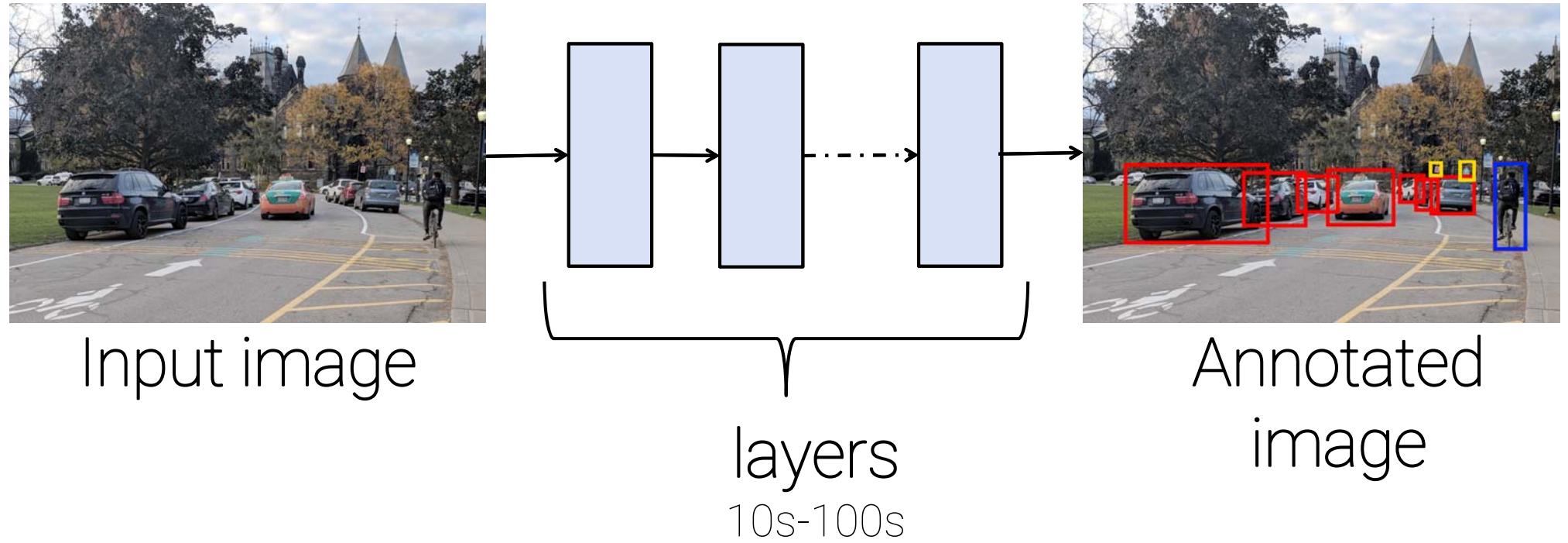
# Bit-Tactical: Sparse/Dense Deep Learning Inference Accelerator(\*)

Alberto Delmas, Patrick Judd, Mostafa Mahmoud,  
Milos Nikolic, Zisis Poulos, Sayeh Sharify,  
Kevin Siu, Dylan Stuart  
**Andreas Moshovos**





Faster  
+  
More Energy Efficient



Fewer Operations  
Fewer Data Transfers

Fewer Operations  
Fewer Data Transfers

# Fewer Operations Fewer Data Transfers

$\text{Out} += \mathbf{A}_0 \times \mathbf{W}_0$

$\text{Out} += \mathbf{A}_1 \times \mathbf{W}_1$

$\text{Out} += \mathbf{A}_2 \times \mathbf{W}_2$

$\text{Out} += \mathbf{A}_3 \times \mathbf{W}_3$

$\text{Out} += \mathbf{A}_4 \times \mathbf{W}_4$

⋮

# Eliminate Ineffectual Operations

**Out += A<sub>0</sub> × W<sub>0</sub>**

**Out += A<sub>1</sub> × W<sub>1</sub>**

**Out += A<sub>2</sub> × W<sub>2</sub>**

**Out += A<sub>3</sub> × W<sub>3</sub>**

**Out += A<sub>4</sub> × W<sub>4</sub>**

⋮

# Eliminate Ineffectual Operations

$\text{Out} += A_0 \times W_0$

$\text{Out} += O \times W_1$

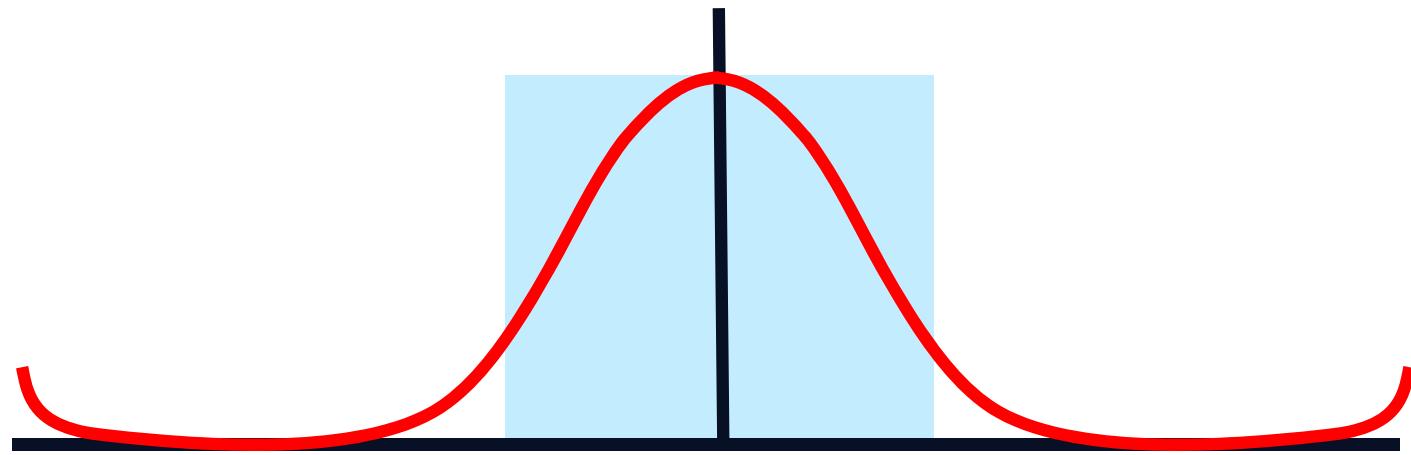
$\text{Out} += A_2 \times W_2$

$\text{Out} += A_3 \times O$

$\text{Out} += A_4 \times W_4$

⋮

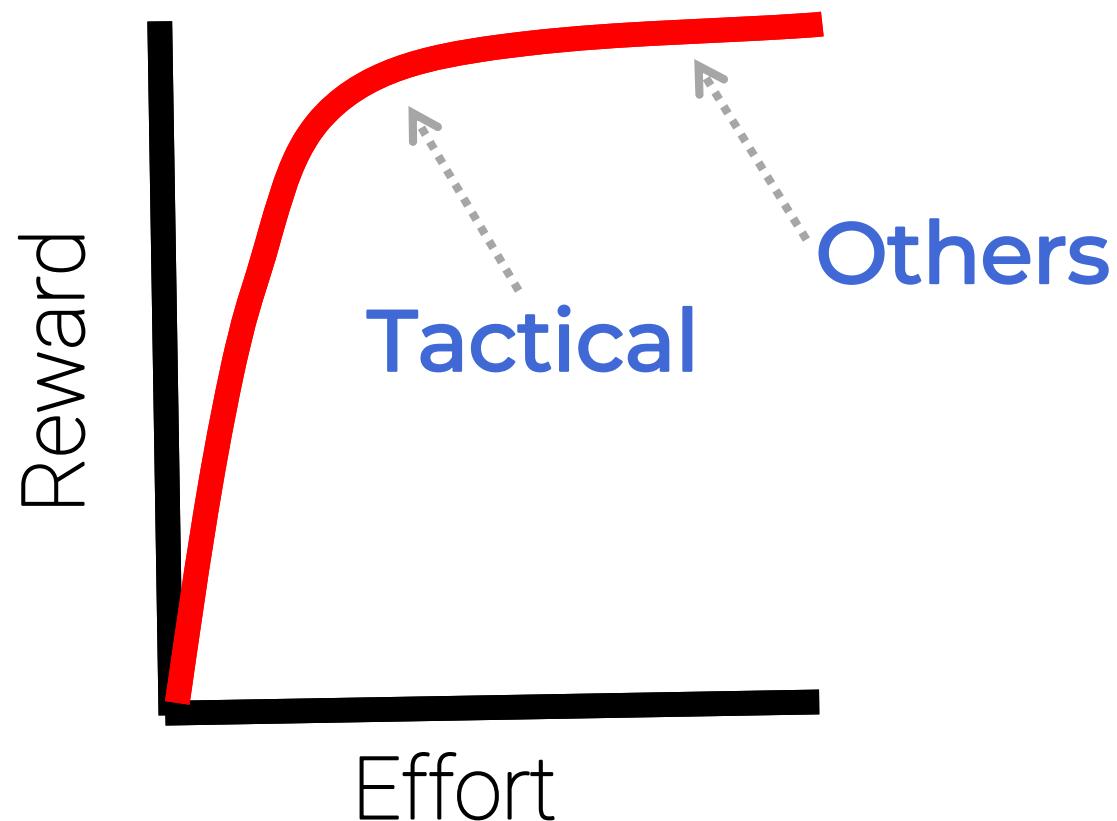
# Takeaway #1



$0.5x - 0.2x$  Off-Chip Transfers  
On-Chip Storage

# Takeaway #2

**Out += A x 0**



# Takeaway #2

Out += A × 0

Aim to get **most** not **all**

Simple Design  
Software Scheduler

# Takeaway #3

Out += 0 × W

Don't!

Only ~50% A==0

# Takeaway #3

$\text{Out} += 0100001 \times W$

Go for Bit Sparsity

> 90%

# Takeaway #4

**vs. SCNN**

**Speed**

**5.3x**

**Energy**

**0.85x**

Do as you are told?

**Out += A<sub>0</sub> × W<sub>0</sub>**

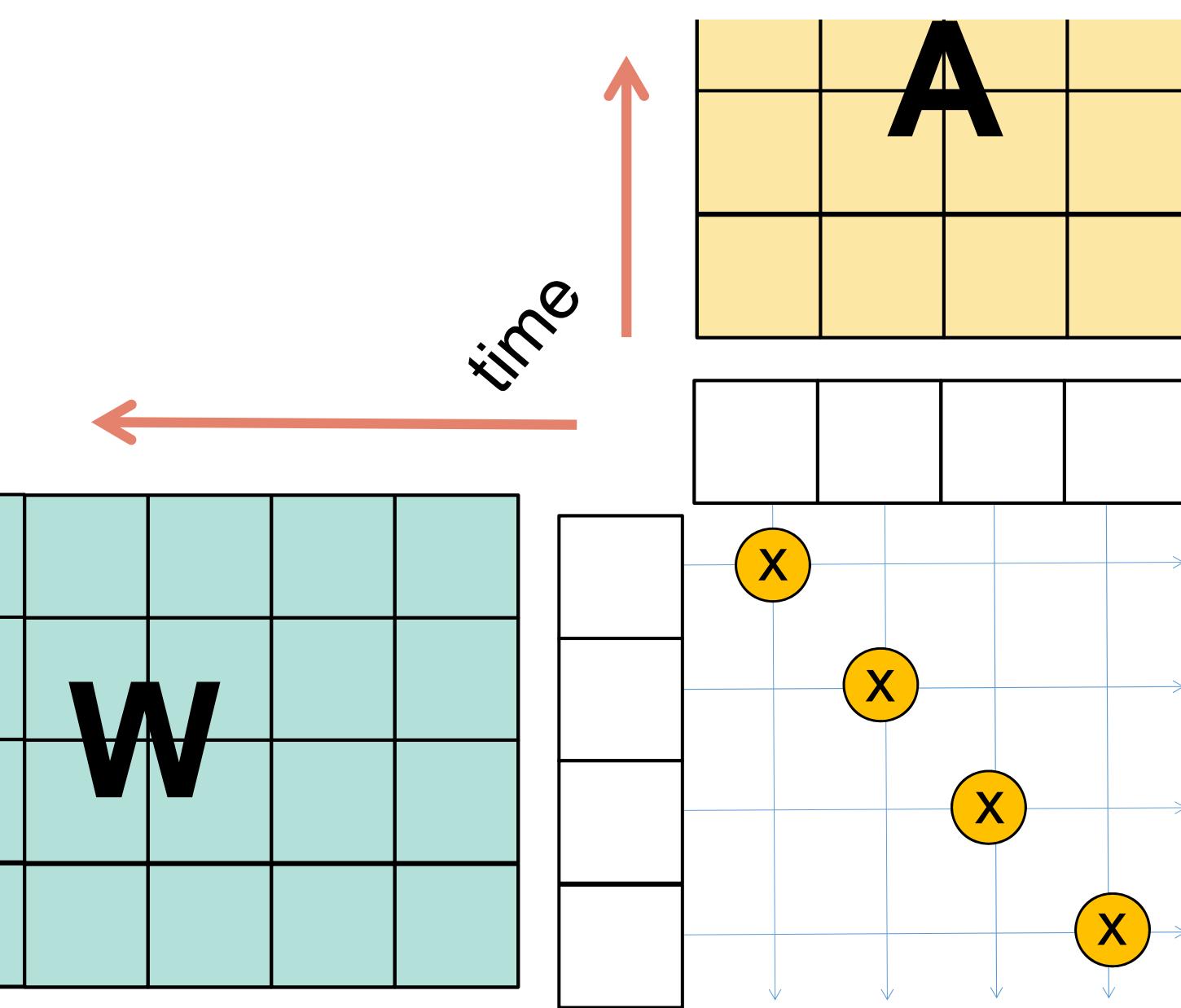
**Out += A<sub>1</sub> × W<sub>1</sub>**

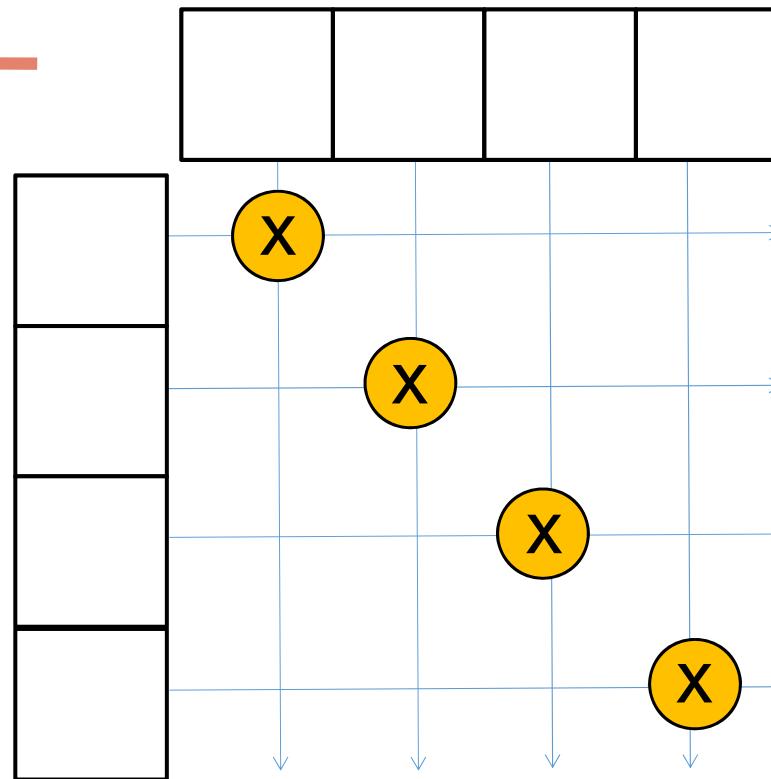
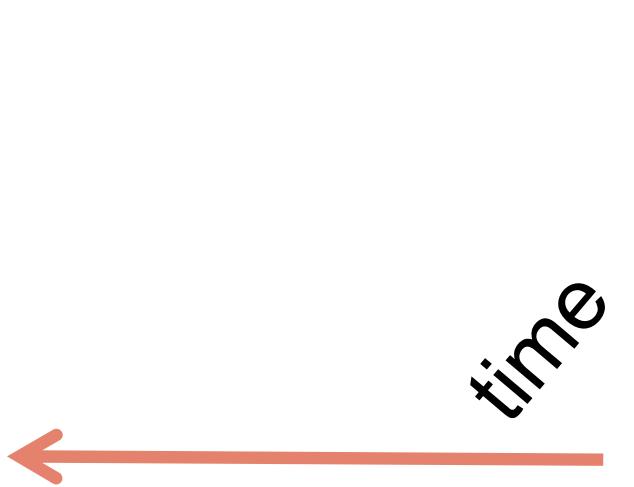
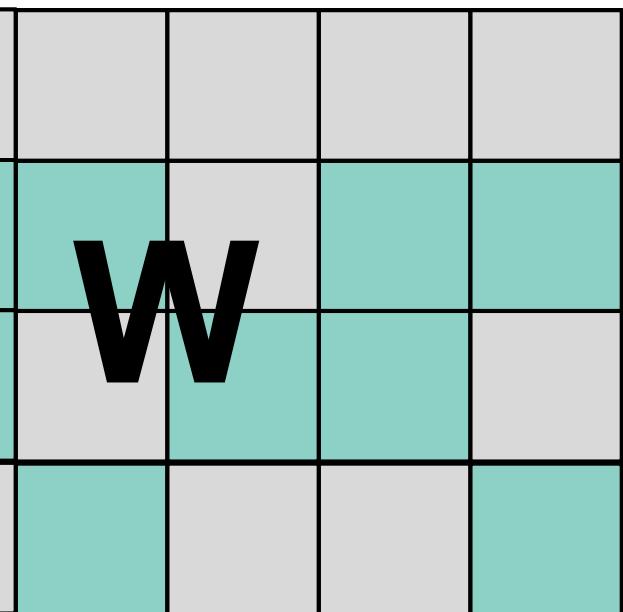
**Out += A<sub>2</sub> × W<sub>2</sub>**

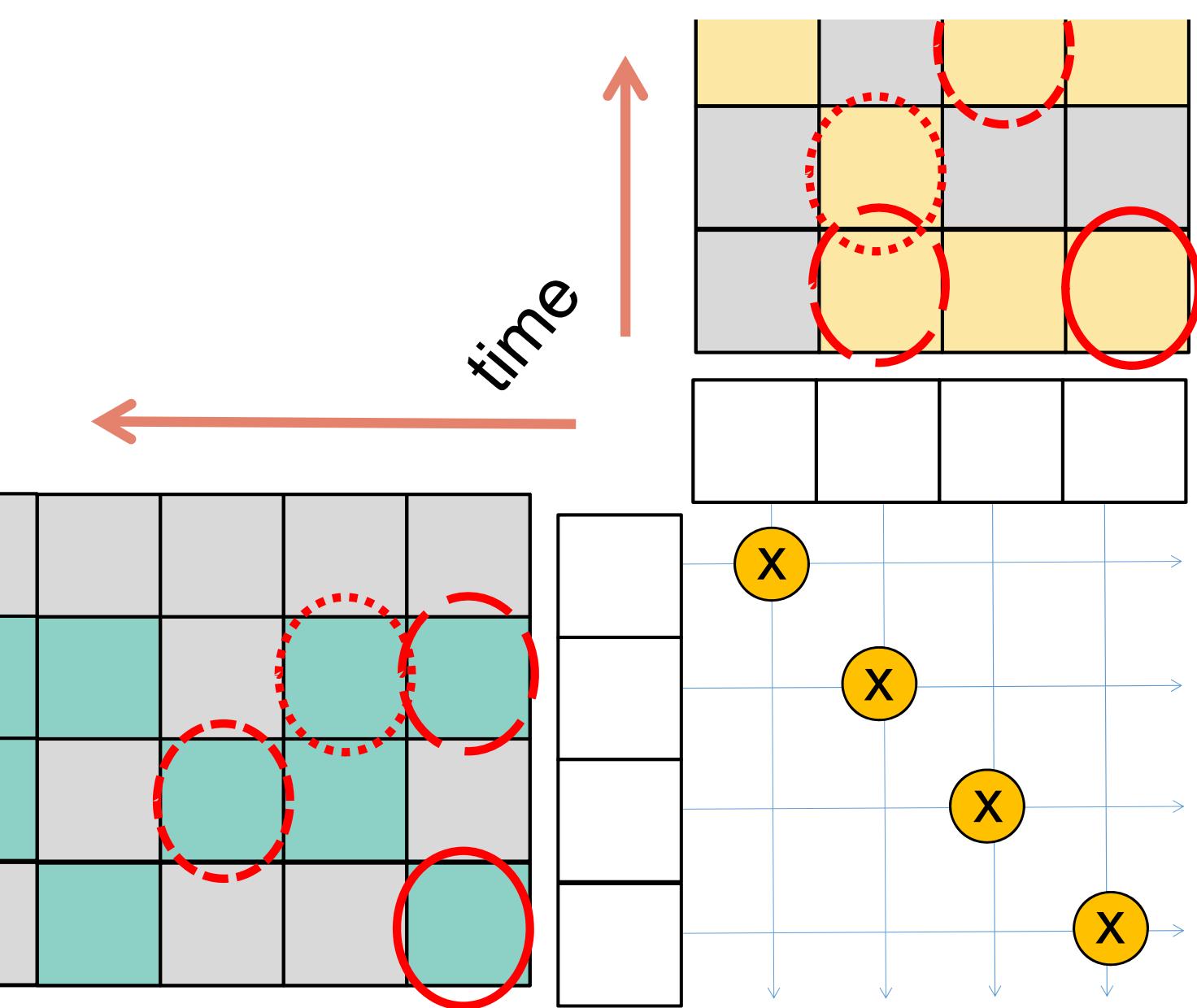
**Out += A<sub>3</sub> × W<sub>3</sub>**

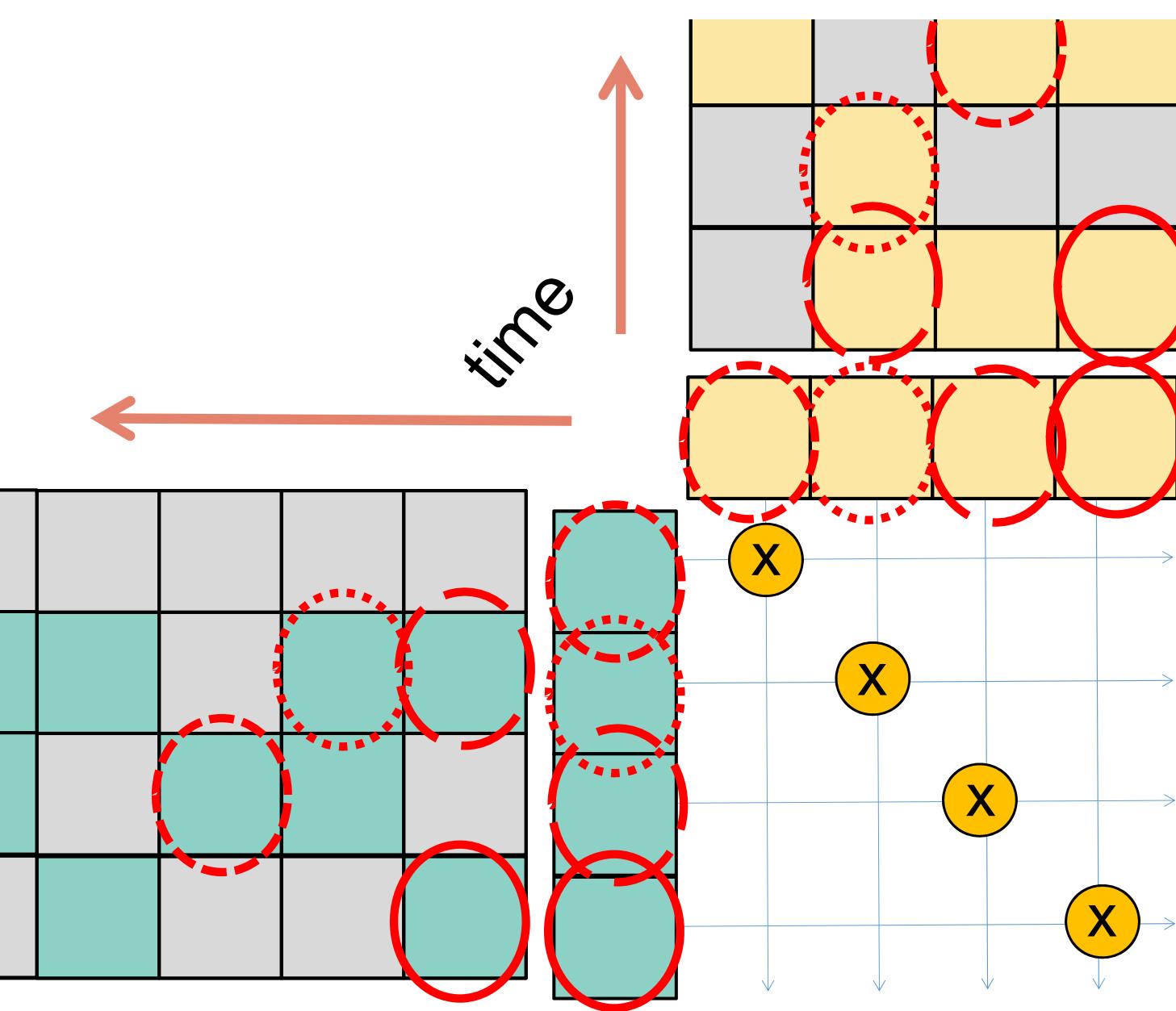
**Out += A<sub>4</sub> × W<sub>4</sub>**

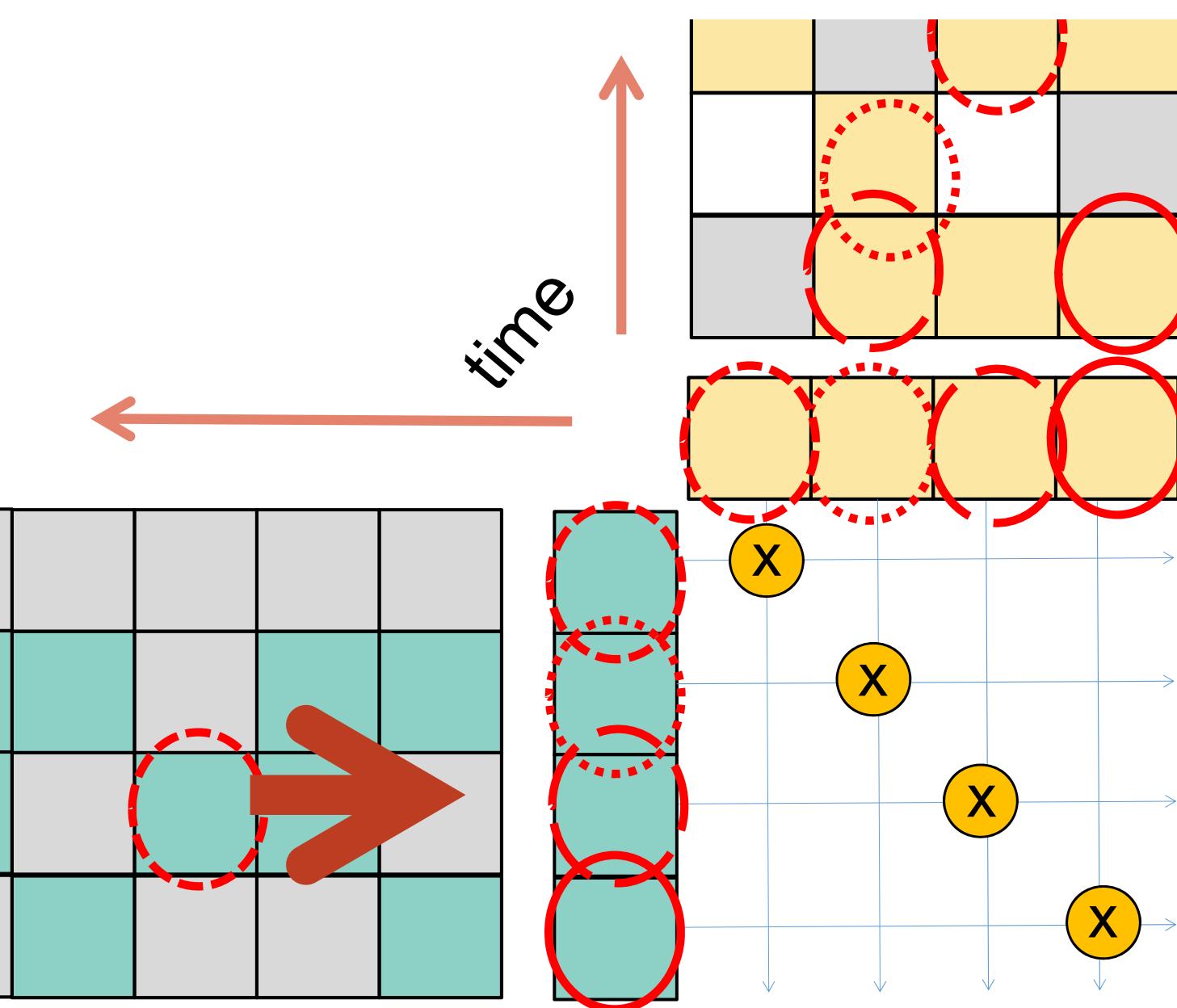




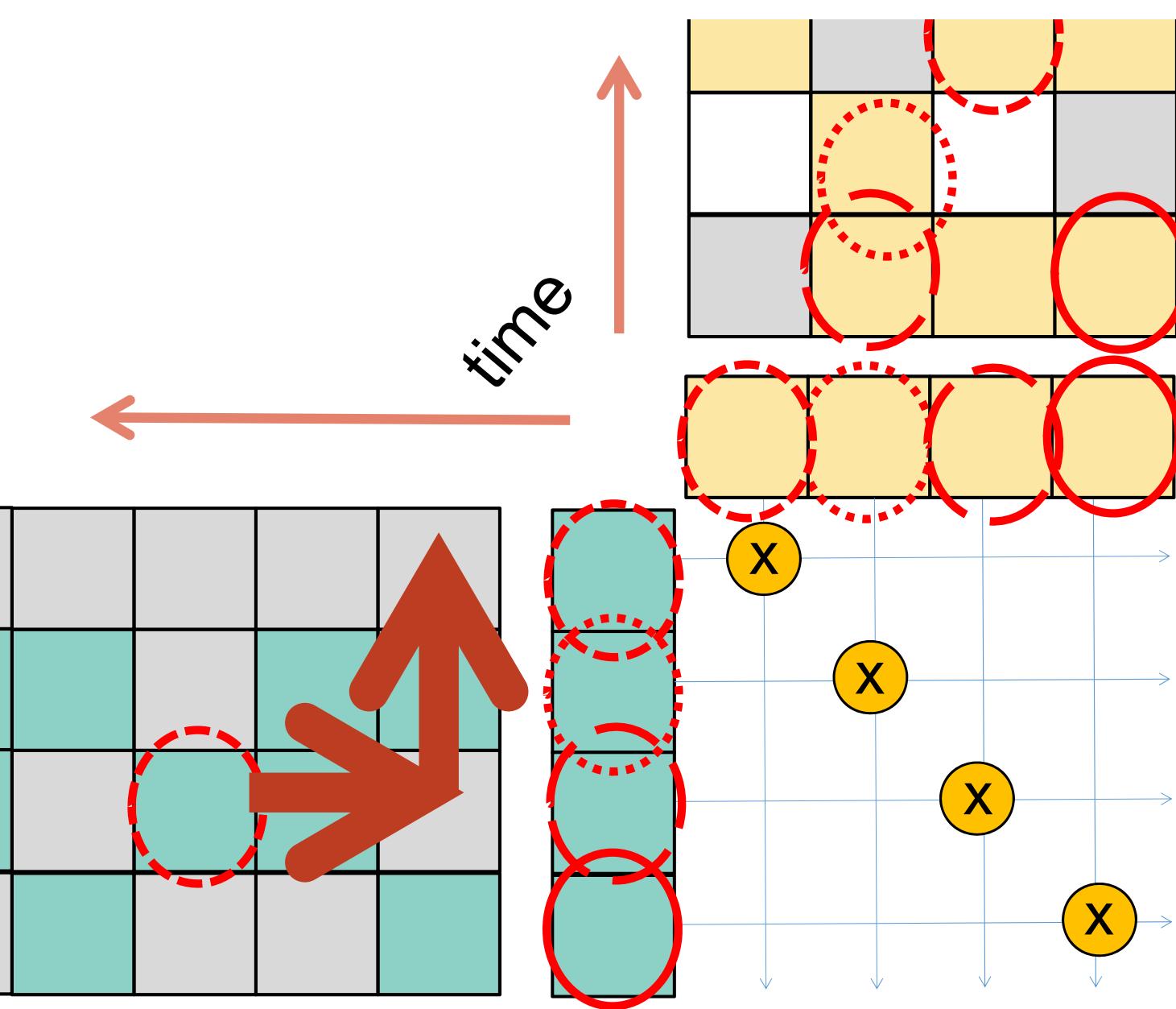




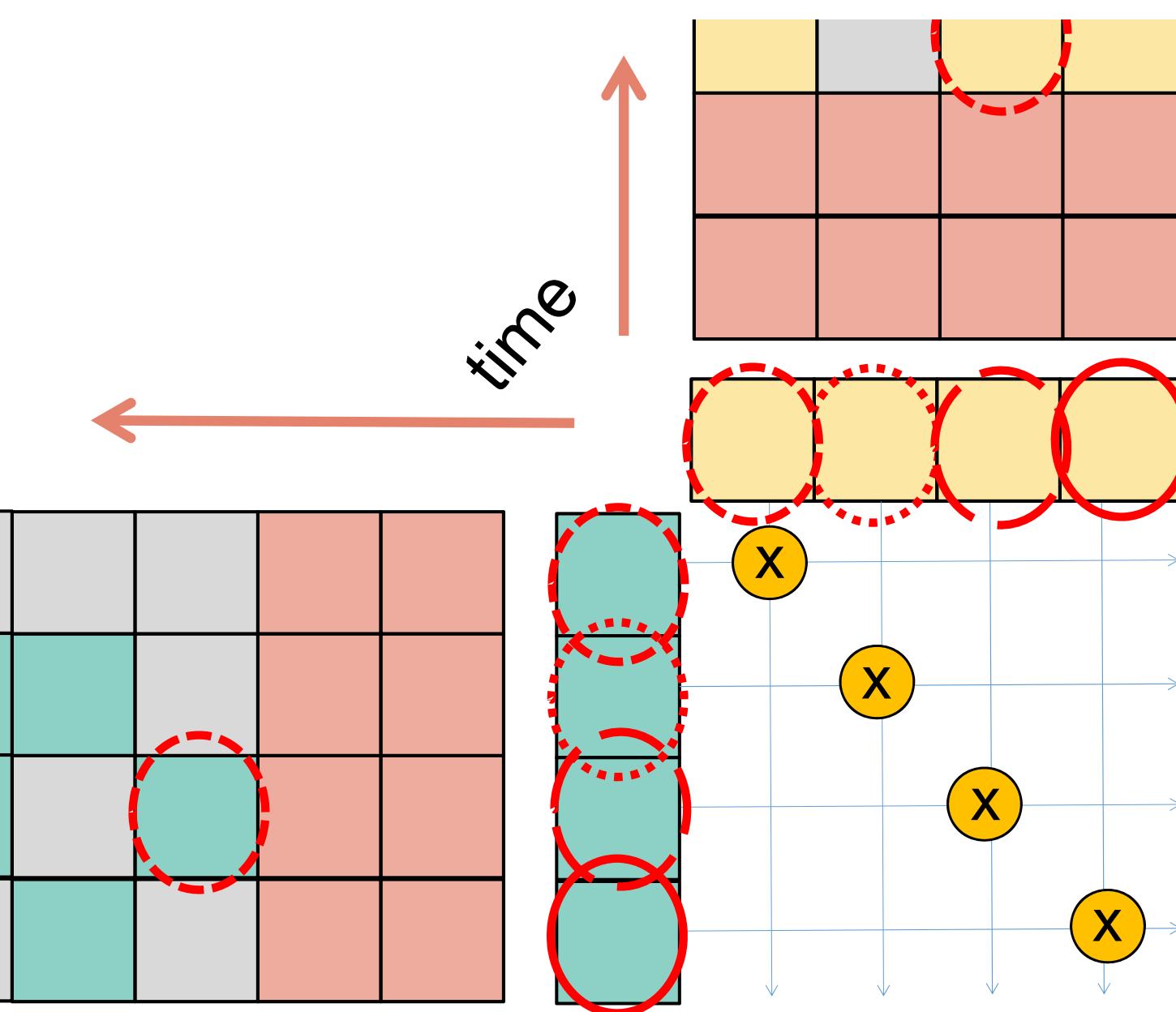




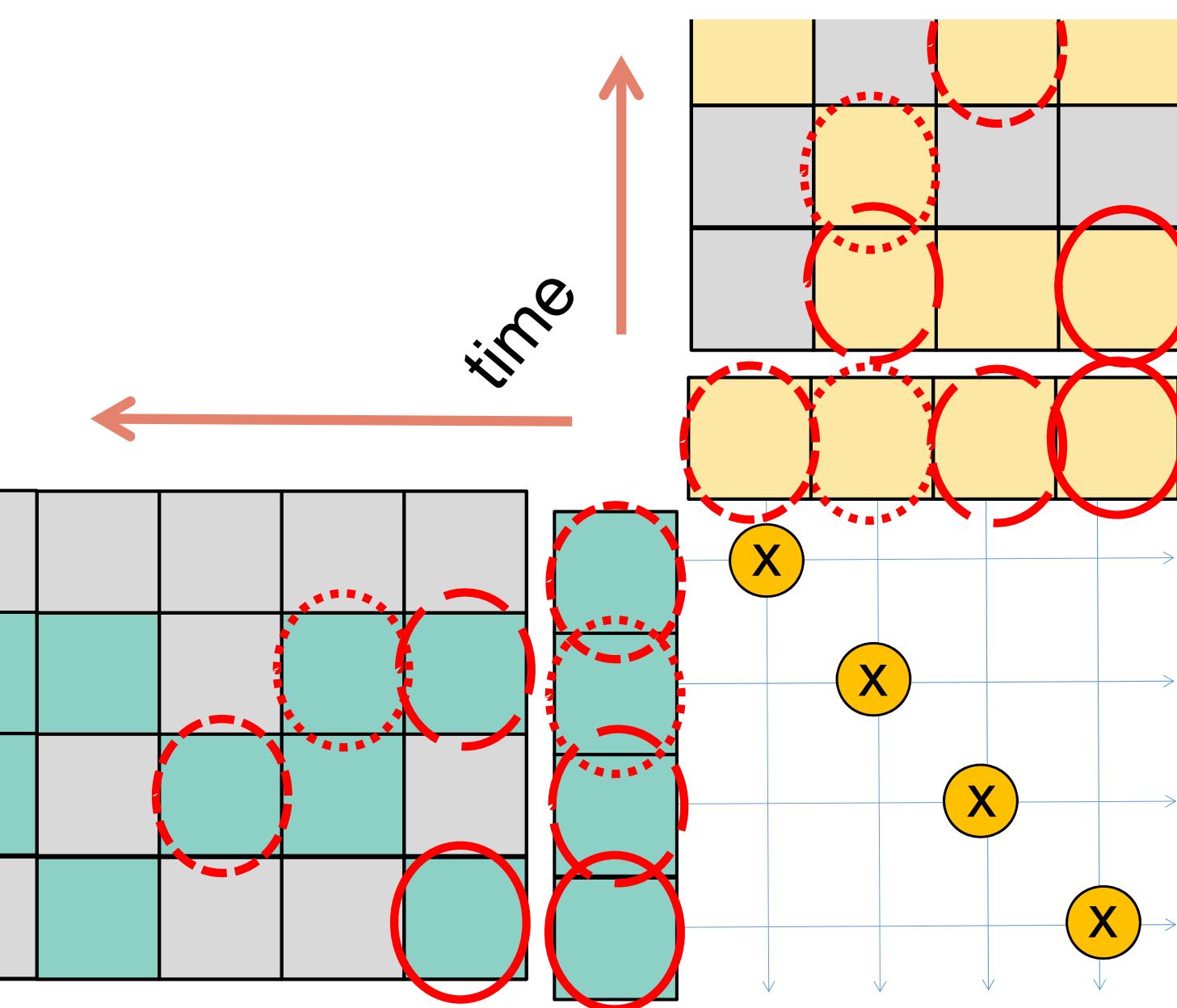
2 steps in time



Another 2 steps in space

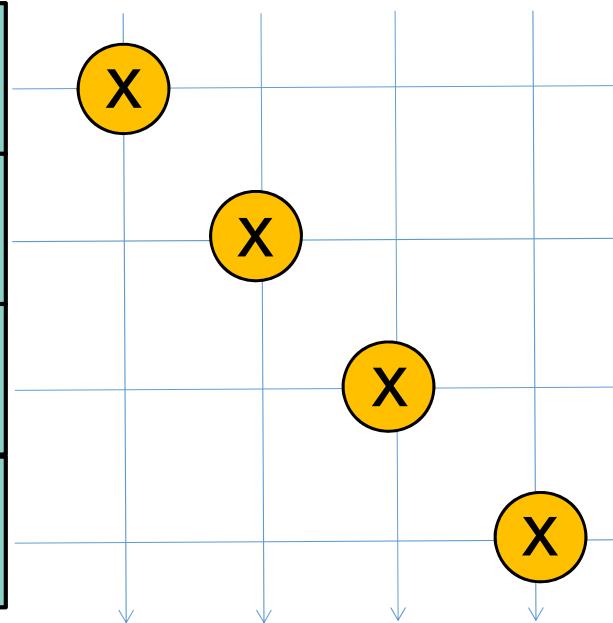
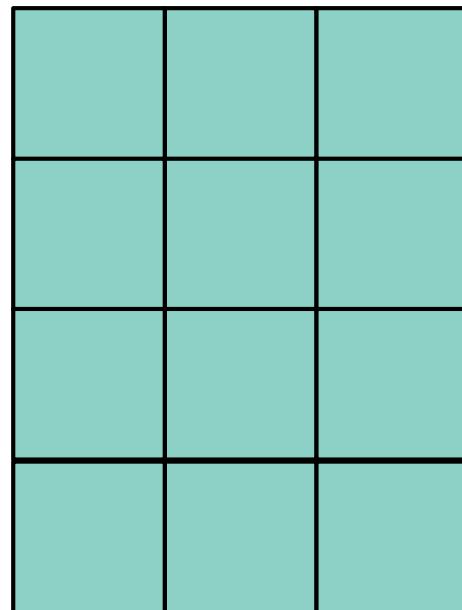
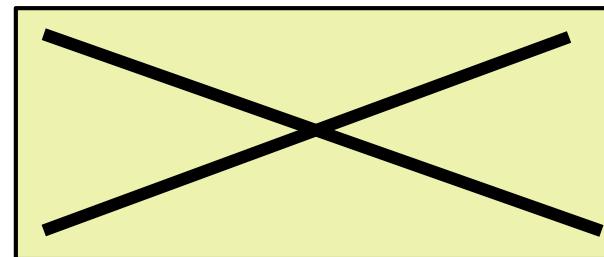
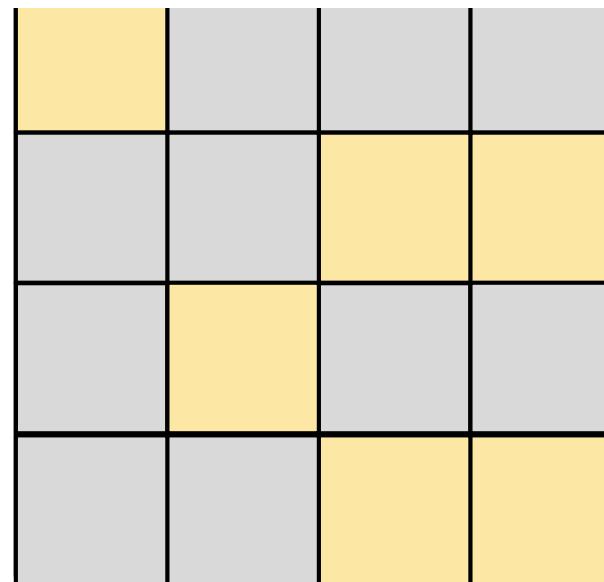


**“unrestricted” motion**



## A: Dynamic Sparsity Pattern

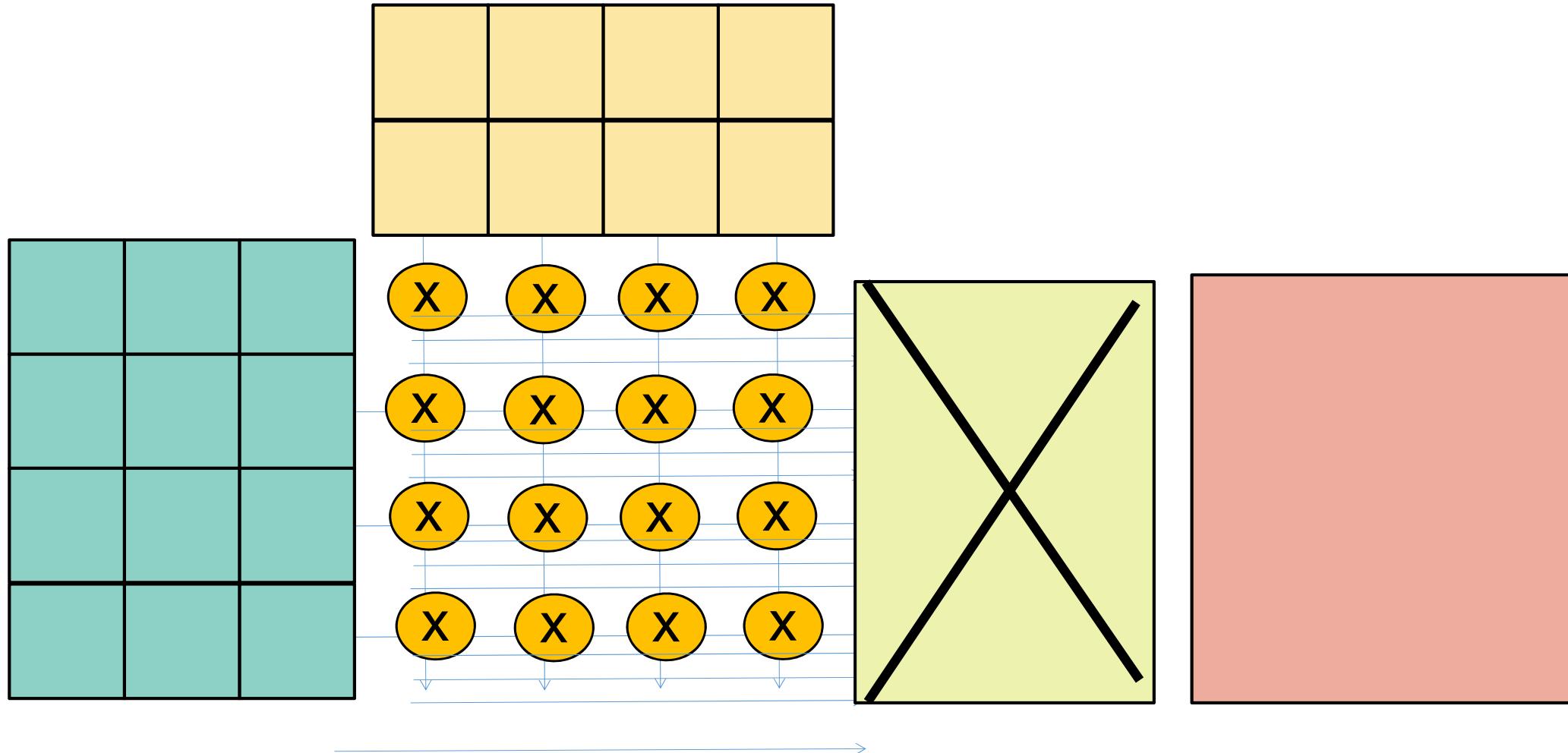
# Cambricon-X



time  
“any”

space  
“any”

# Nvidia's SCNN



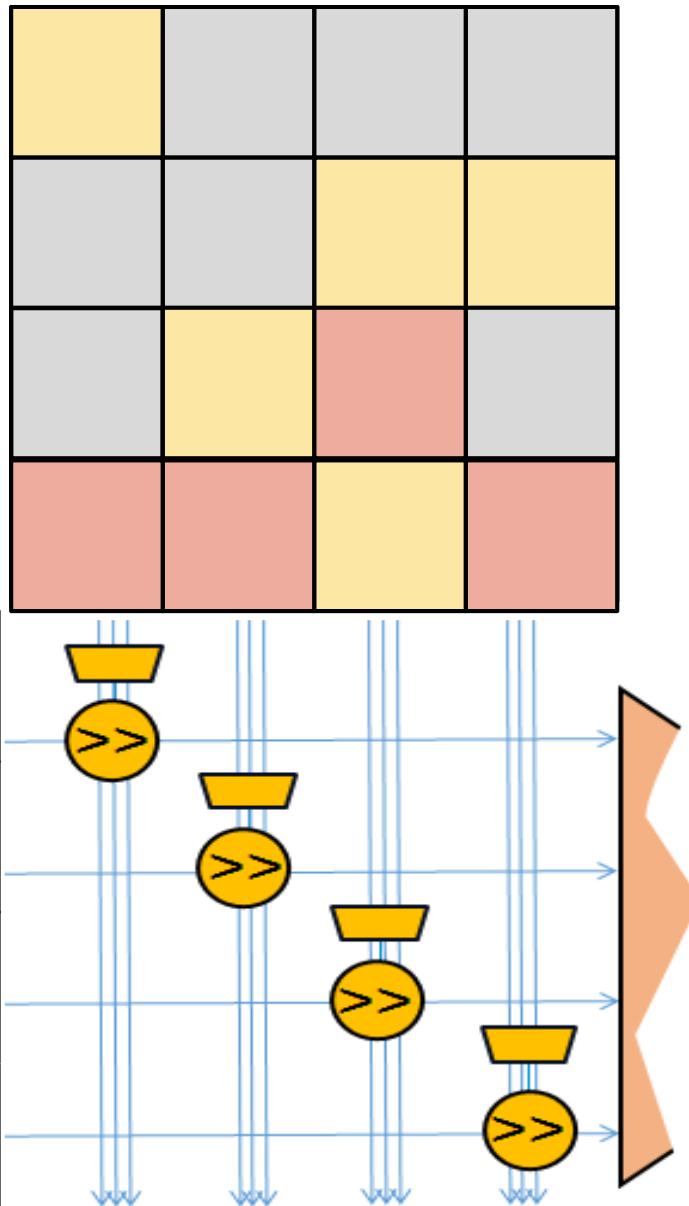
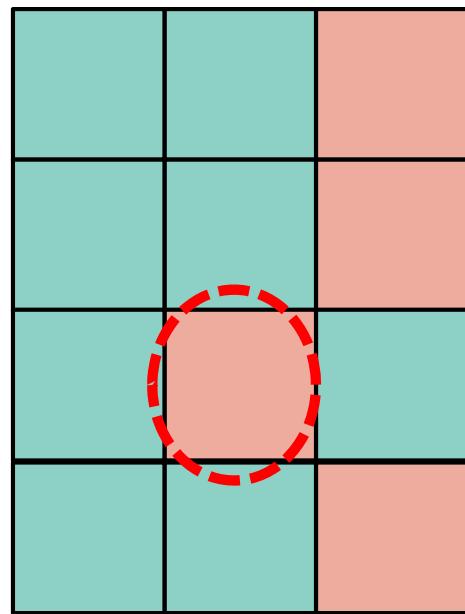
time

“any”

space

“any”

# Bit-Tactical



Weights

time

1-2 steps

space

2-5 places

Do as you are told?

Out += A<sub>0</sub> × W<sub>0</sub>

Out += A<sub>2</sub> × W<sub>2</sub>

Out += 0 × W<sub>3</sub>

Out += A<sub>4</sub> × W<sub>4</sub>



Do as you are told?

$\text{Out} += A_0 \times W_0$

$\text{Out} += A_2 \times W_2$

$\text{Out} += \text{O} \times W_3$

$\text{Out} += A_4 \times W_4$

⋮

Do as you are told?

Out += A<sub>0</sub> × W<sub>0</sub>

Out += A<sub>2</sub> × W<sub>2</sub>

Out += A<sub>3</sub> × W<sub>3</sub>

Out += A<sub>4</sub> × W<sub>4</sub>



Do as you are told?

Out += 000100100 x w<sub>0</sub>

Out += 000110100 x w<sub>2</sub>

Out += 000010000 x w<sub>3</sub>

Out += 010000010 x w<sub>4</sub>



Do as you are told?

Out += 000100100 x w<sub>0</sub>

Out += 000110100 x w<sub>2</sub>

Out += 000010000 x w<sub>3</sub>

Out += 010000010 x w<sub>4</sub>

⋮

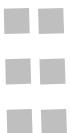
## Do as you are told?

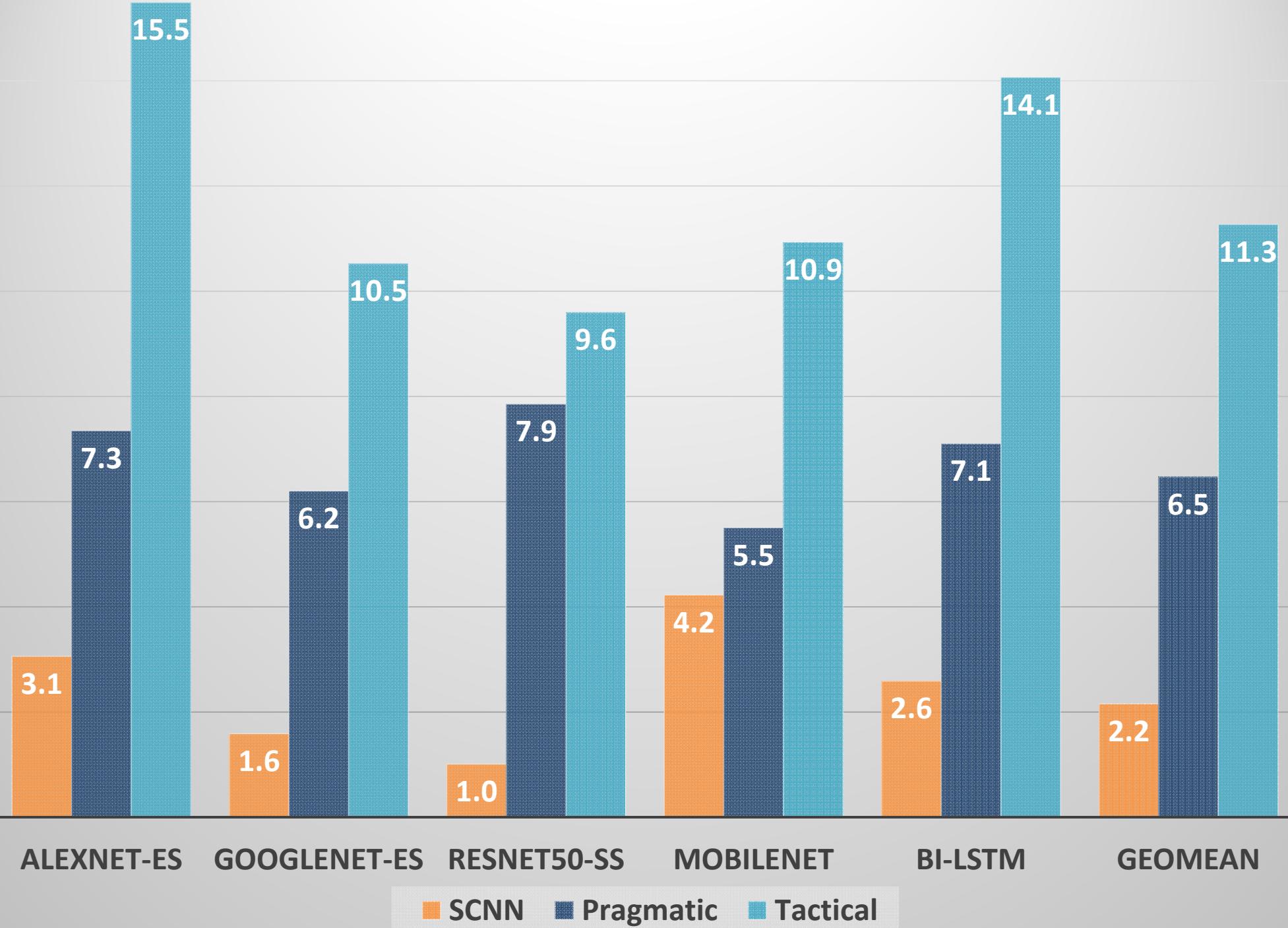
Out += 1 1 x W<sub>0</sub>

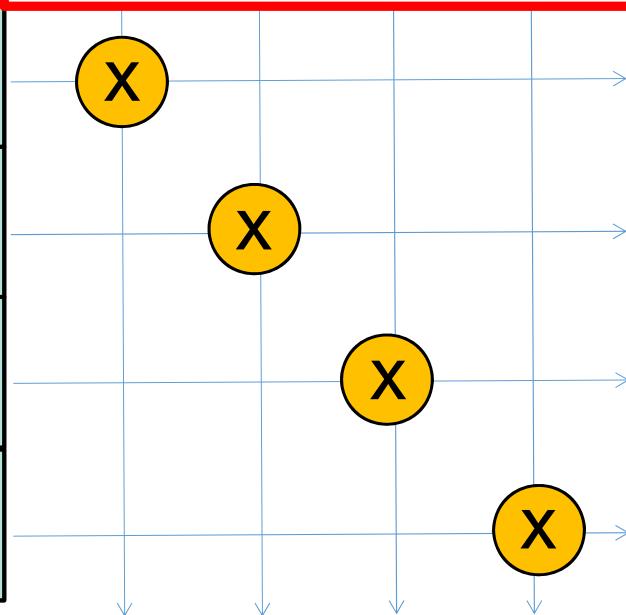
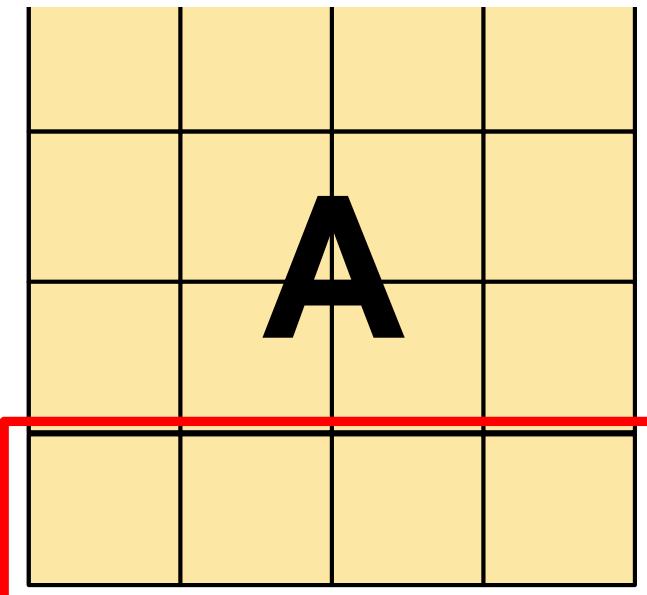
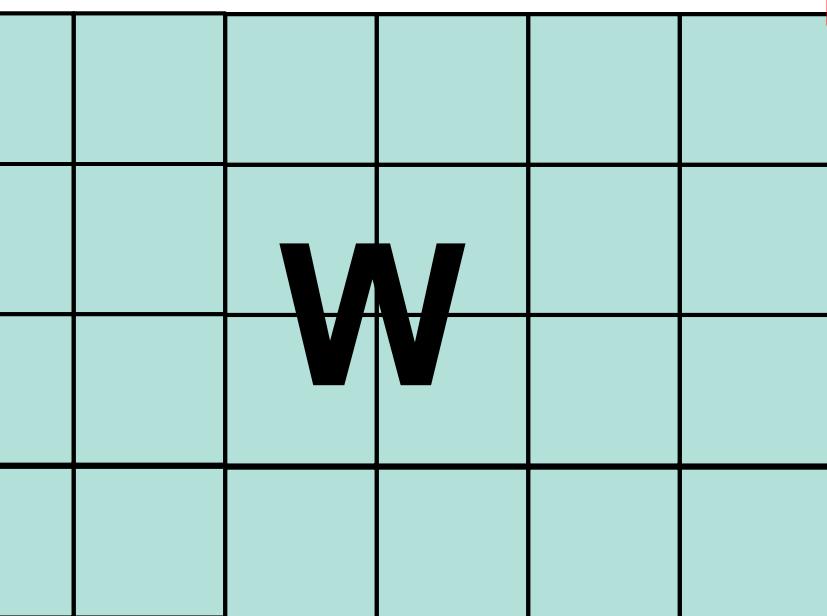
Out += 11 1 x W<sub>2</sub>

Out += 1 x W<sub>3</sub>

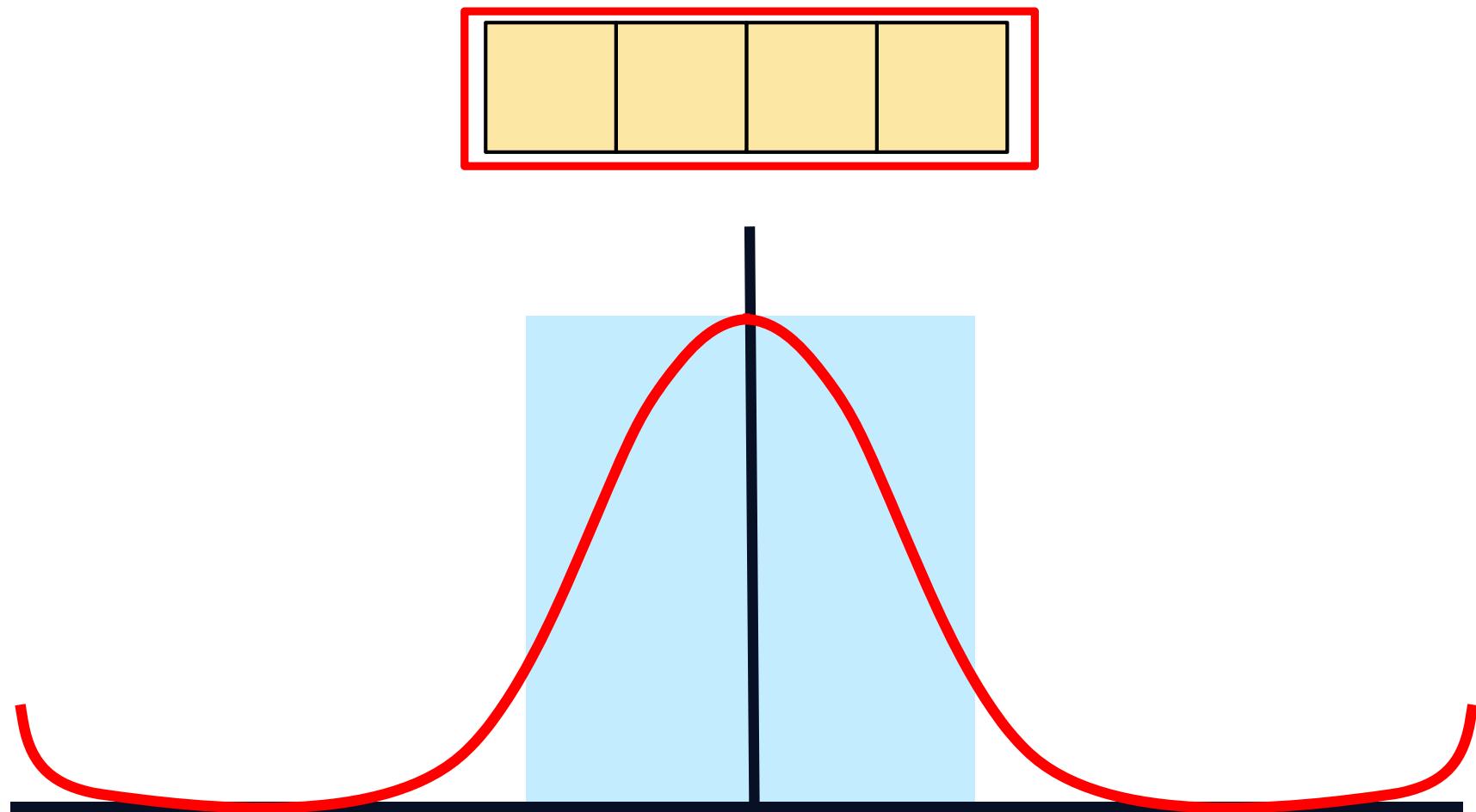
Out += 1 1 x W<sub>4</sub>





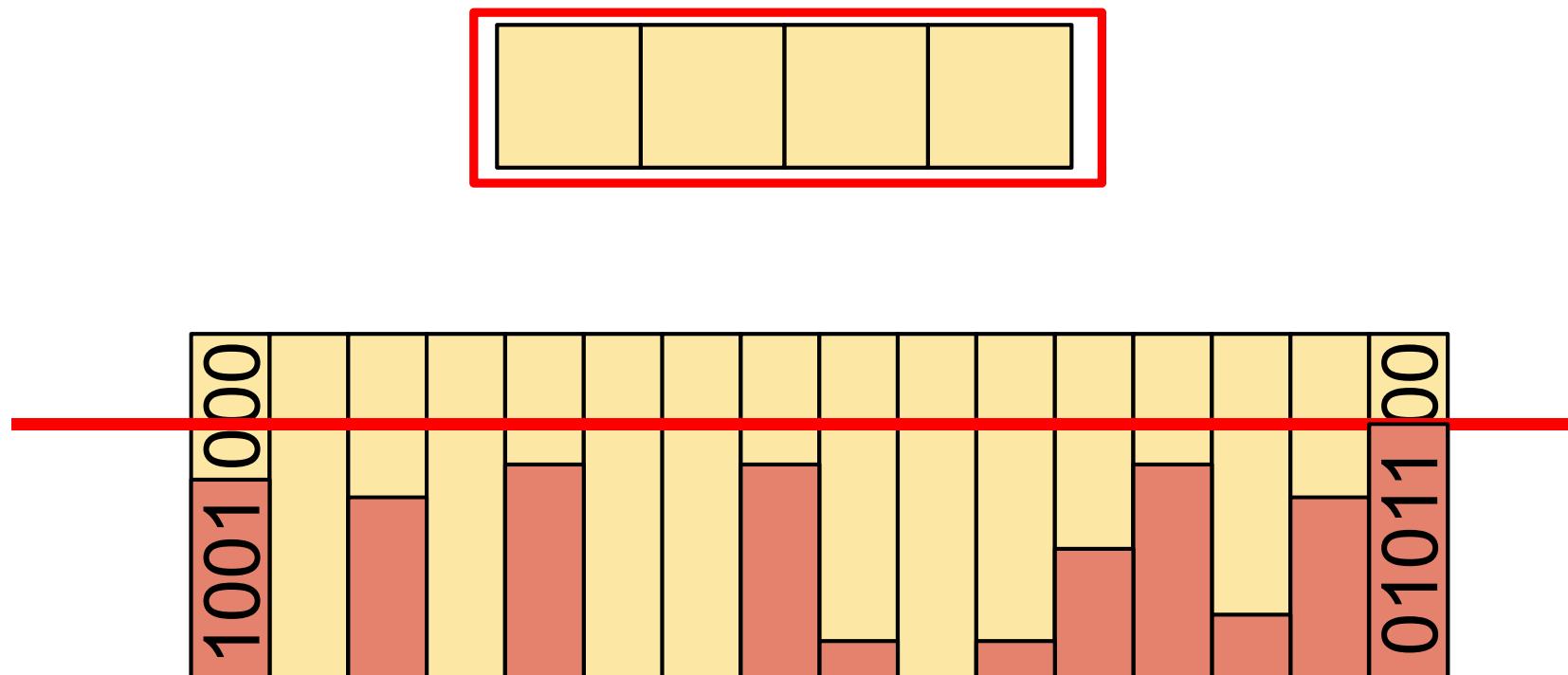


# Making Typical Values Matter



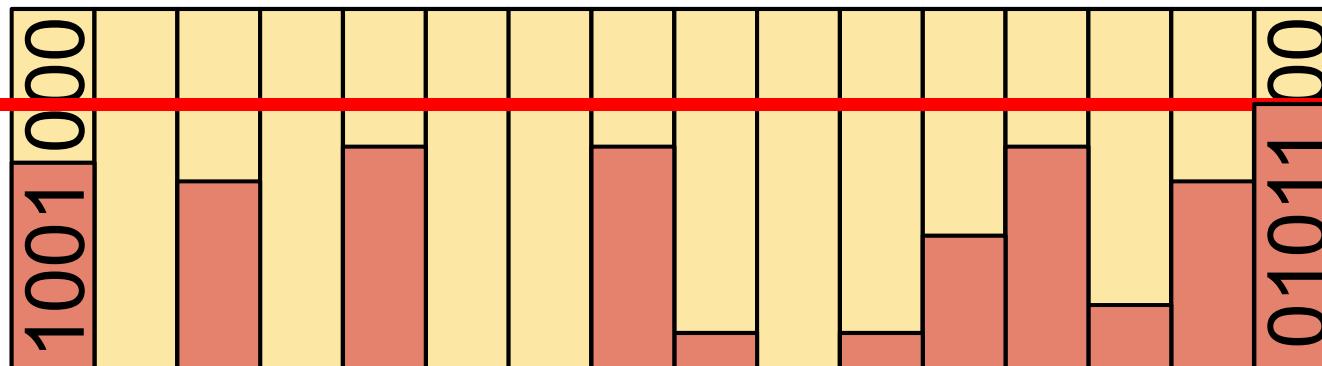
DPRed: Making Typical Activation and Weight Values Matter In Deep Learning Computing, Delmas et al.,  
<https://arxiv.org/abs/1804.06732>

# Fine-Grain Precision Adaptation

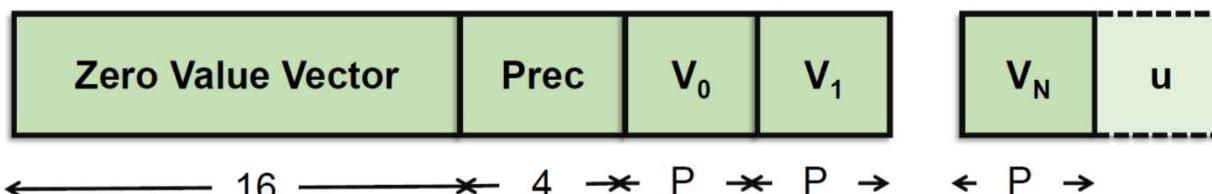
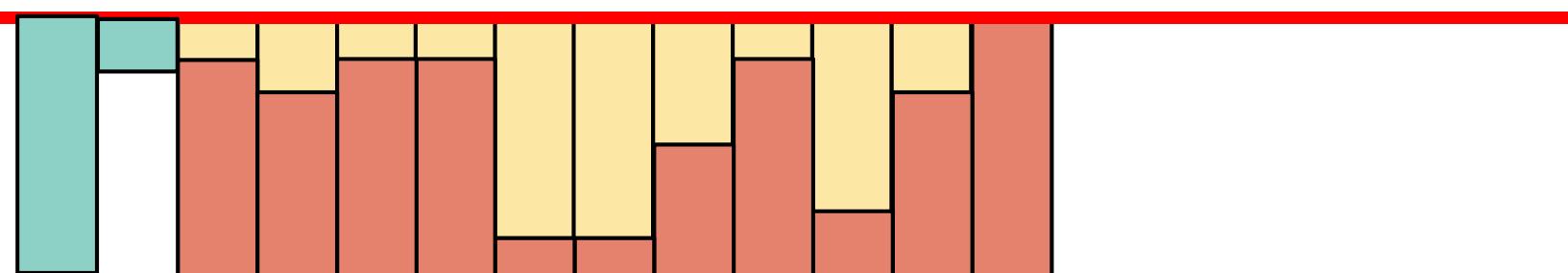


# Per Group Precision Detection/Encoding

**Detect Precisions:** On-the-fly for Activations + Statically for Weights



In Memory



# Takeaways

Don't go after all  $W = 0$

Go for enough

Software Scheduler + Restricted Motions are enough

Don't go after  $A = 0$

Go for bit sparsity 50% vs 90%+

Don't let the loud values dominate the data type

Encode in groups

Bit-Tactical: Exploiting Ineffectual Computations in Convolutional Neural Networks: Which, Why, and How, [Alberto Delmas](#), [Patrick Judd](#), [Dylan Malone Stuart](#), [Zisis Poulos](#), [Mostafa Mahmoud](#), [Sayeh Sharify](#), [Milos Nikolic](#), [Andreas Moshovos](#), arXiv:1803.03688

## Laconic: Goal

W 0001 0100

A 0010 1010

---

0000 0000

0010 1010

0000 0000

0010 1010

0000 0000

0010 1010

0000 0000

0000 0000

# Laconic: Goal

W 0001 0100

A 0010 1010



0010 1010

0000 0000

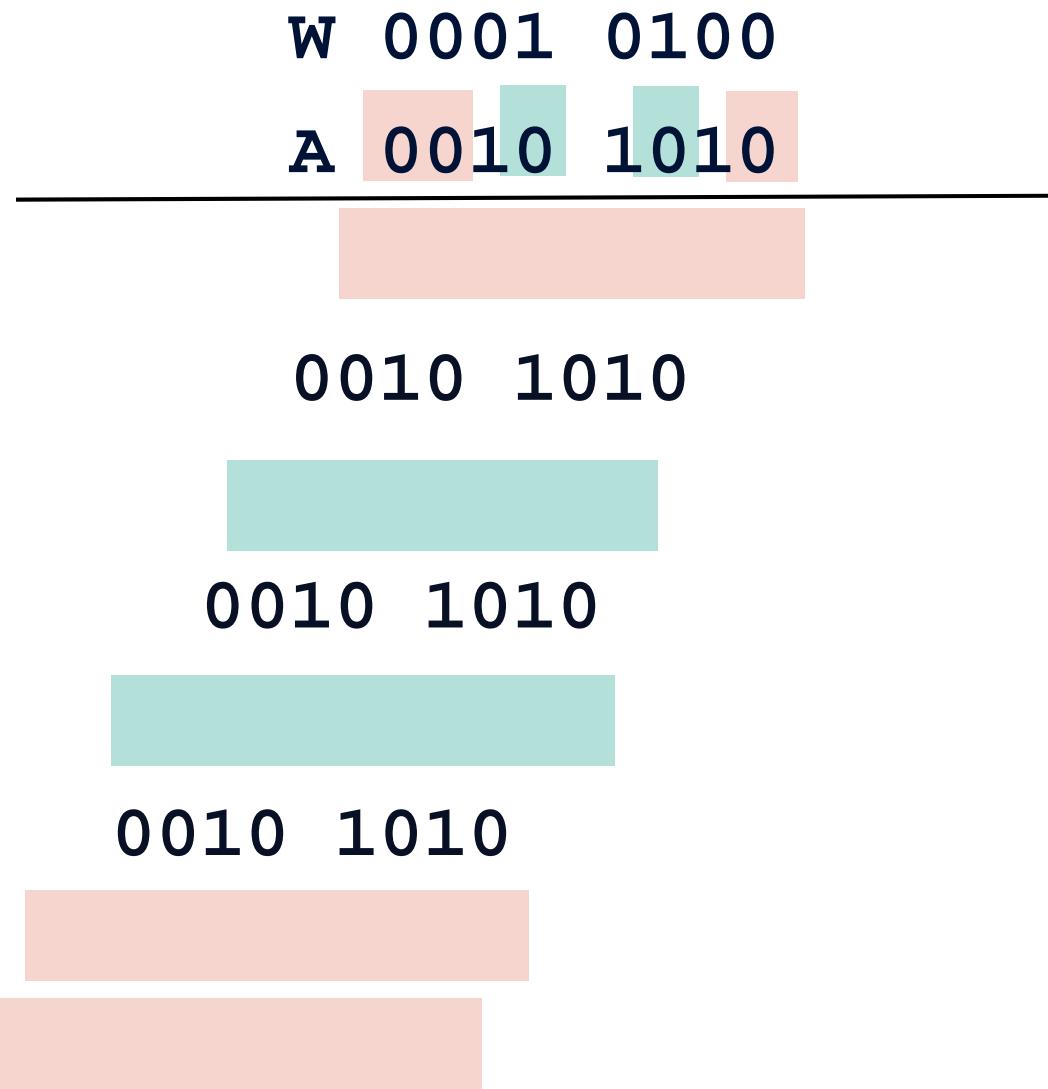
0010 1010

0000 0000

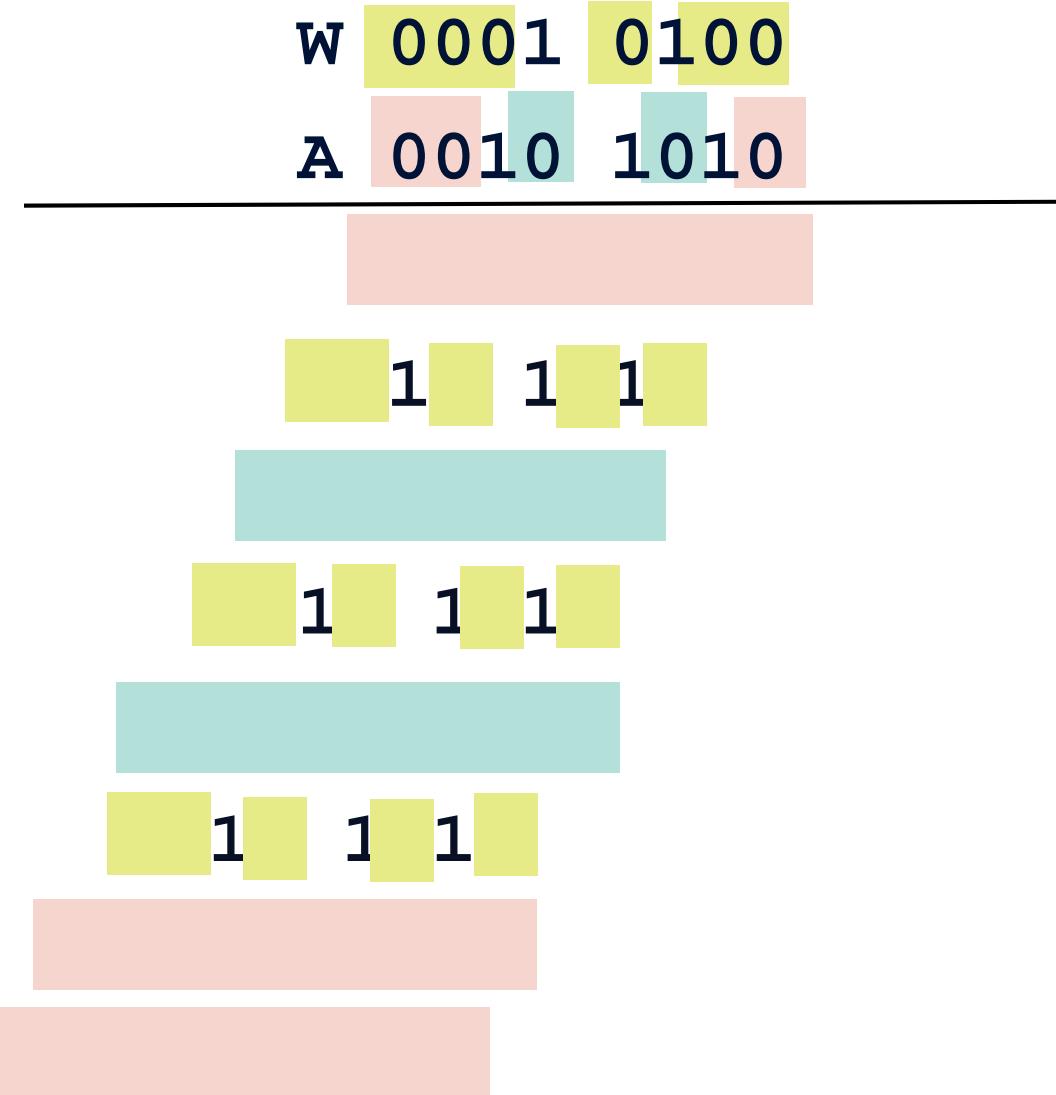
0010 1010



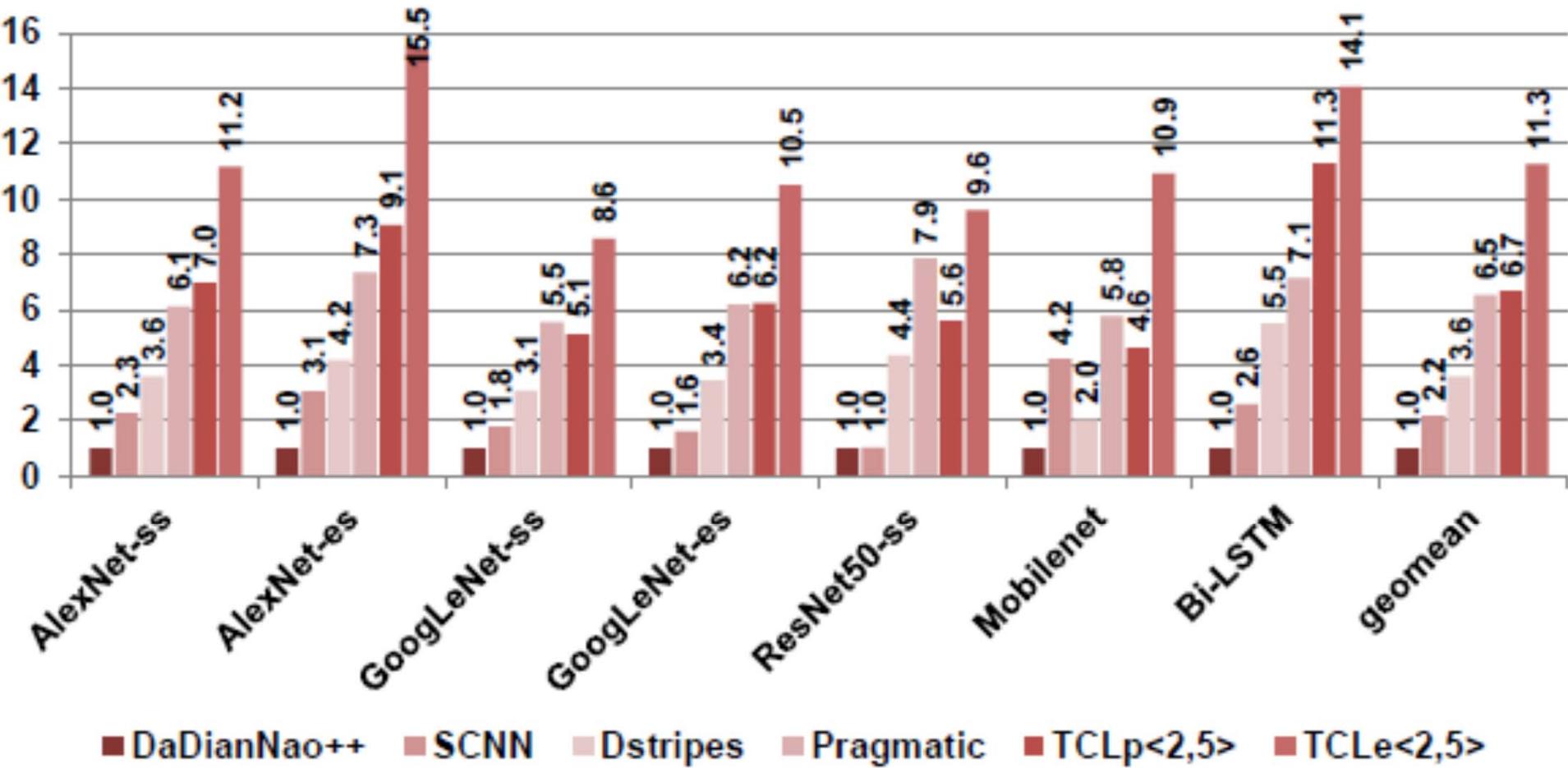
# Laconic: Goal

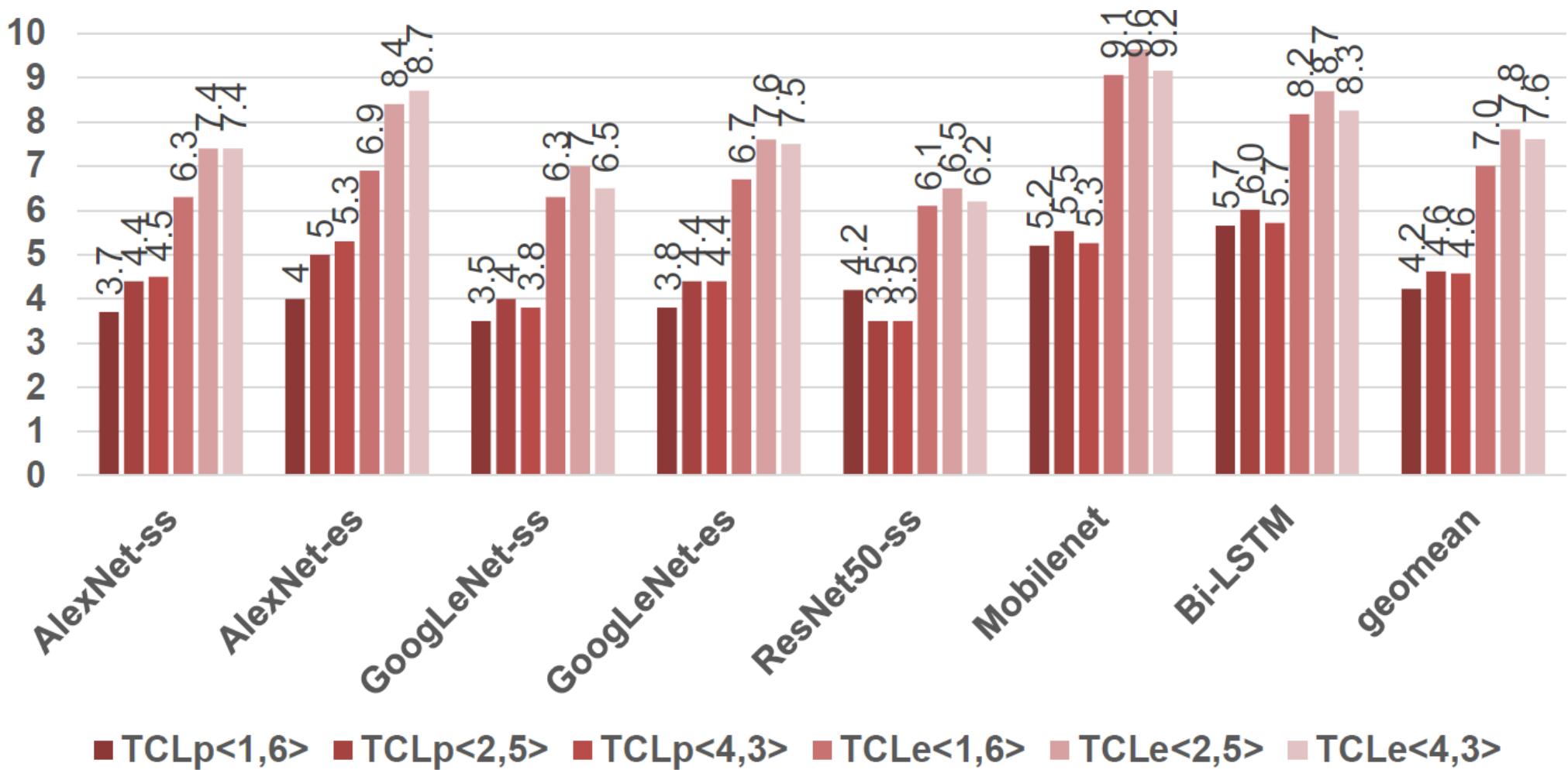


# Laconic: Goal



- Best for edge devices
- Laconic Deep Learning Computing, Sayeh Sharify, Mostafa Mahmoud, Alberto Delmas Lascorz, Milos Nikolic, Andreas Moshovos, Arxiv, arXiv:1805.04513





**Figure 12: Speedup with 8b quantization**