

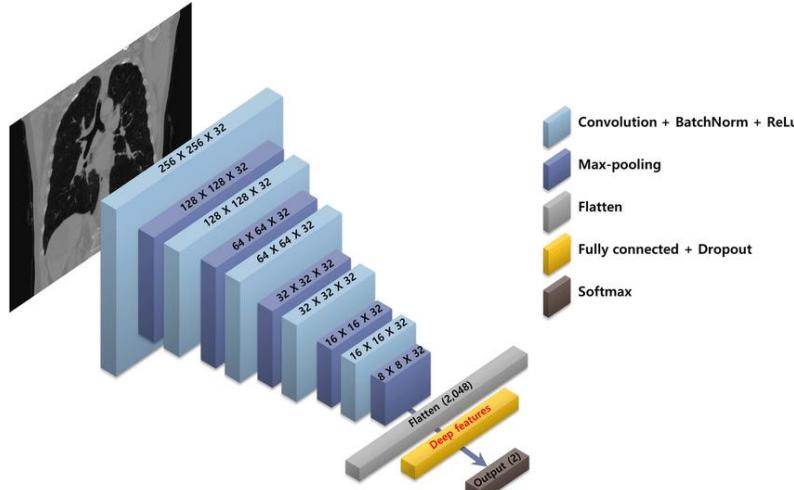
OuroMamba: A Data-Free Quantization Framework for Vision Mamba Models

ICCV 2025

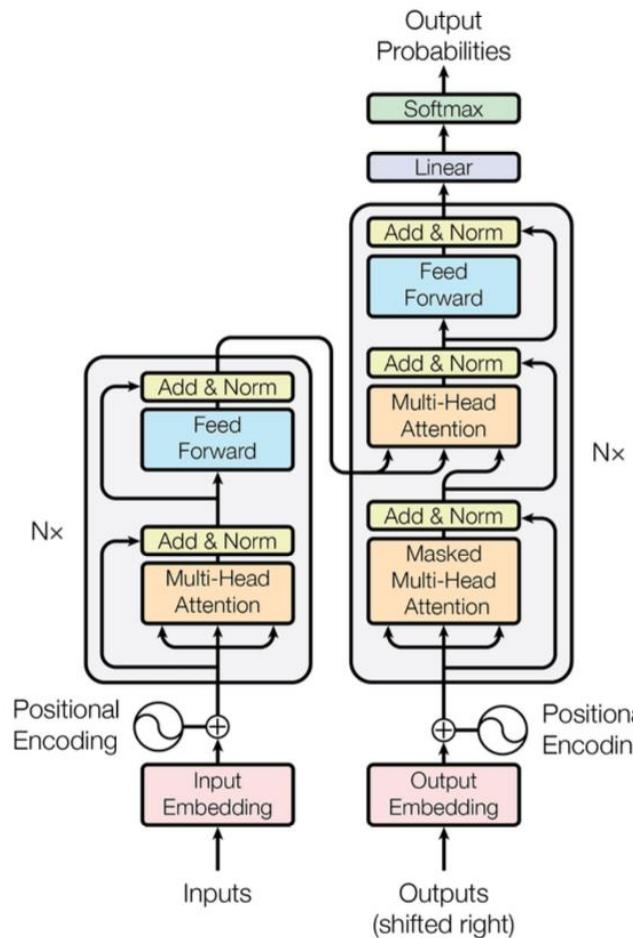
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*Equal Contribution

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Contact: akshat.r@gatech.edu

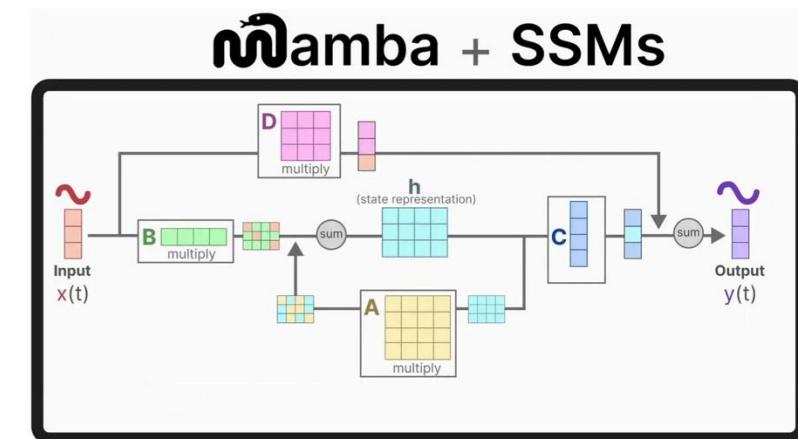
Diverse Space of AI Models



Convolutional Neural Networks

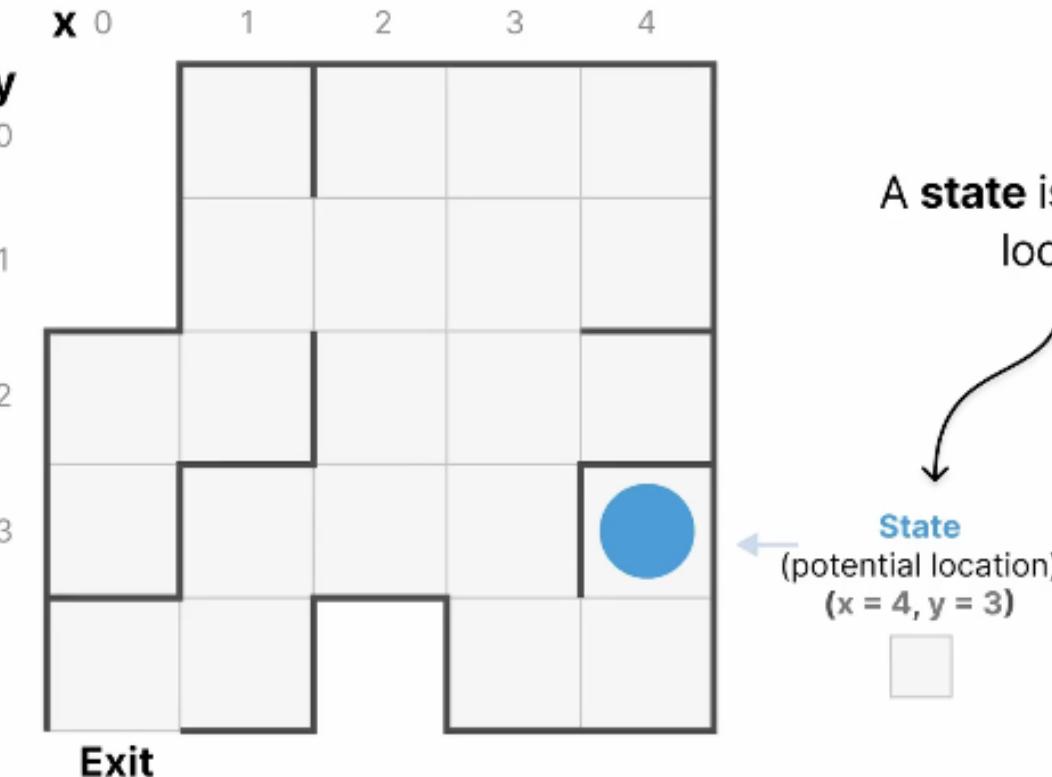


Transformer-based Models



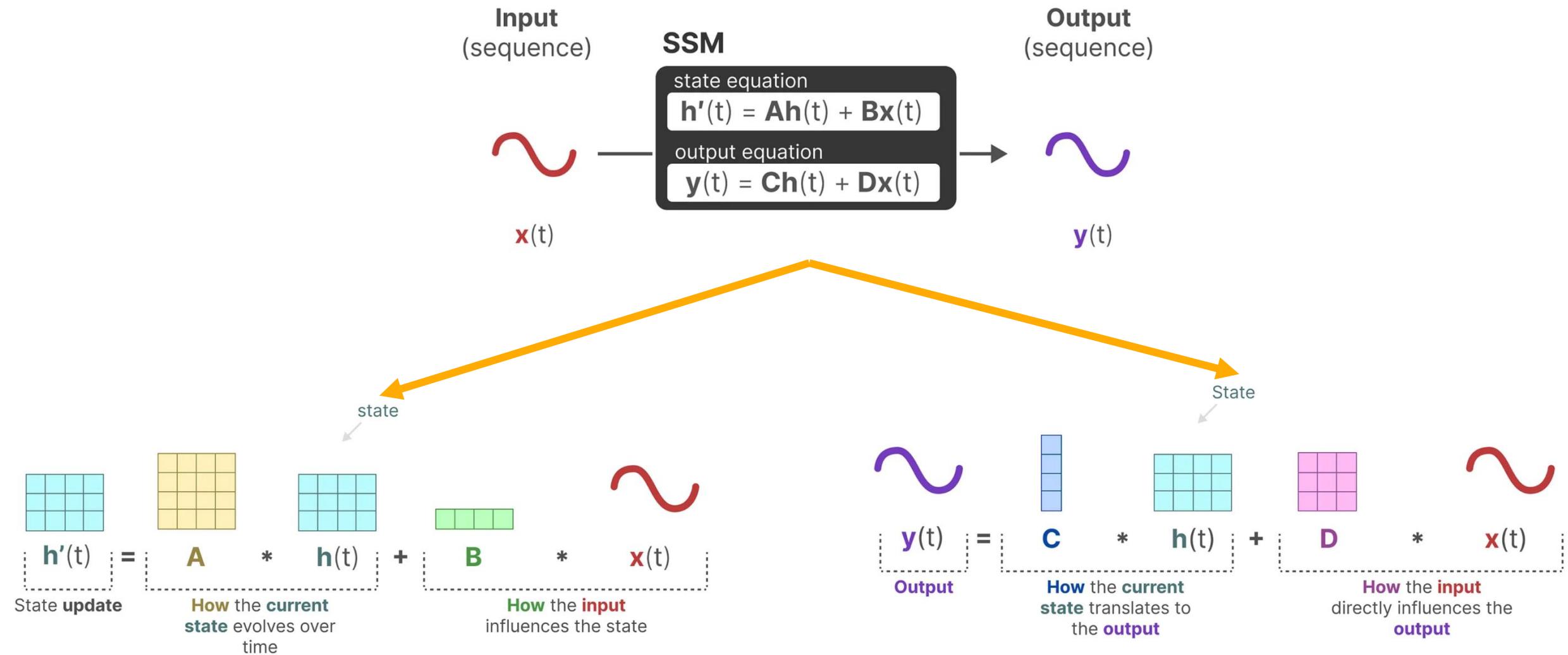
State-Space Models

State Space Model (SSM) Family

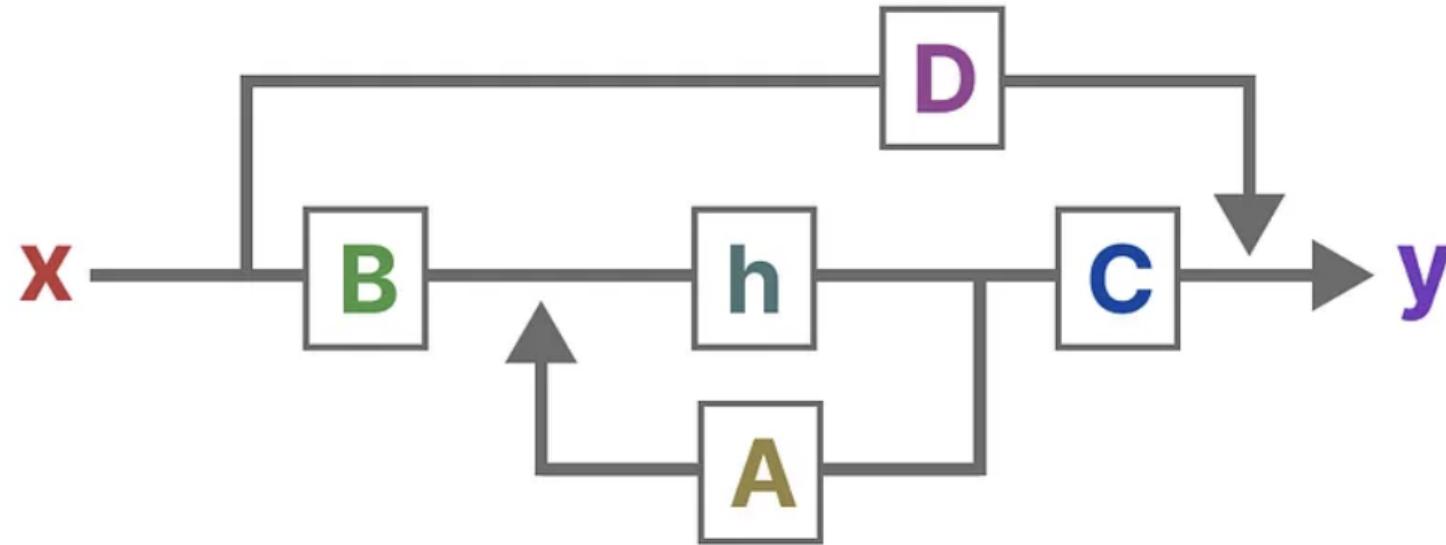


A State Space contains the minimum number of variables that fully describe a system. It is a way to mathematically represent a problem by defining a system's possible states.

The Two Core Equations of SSMs

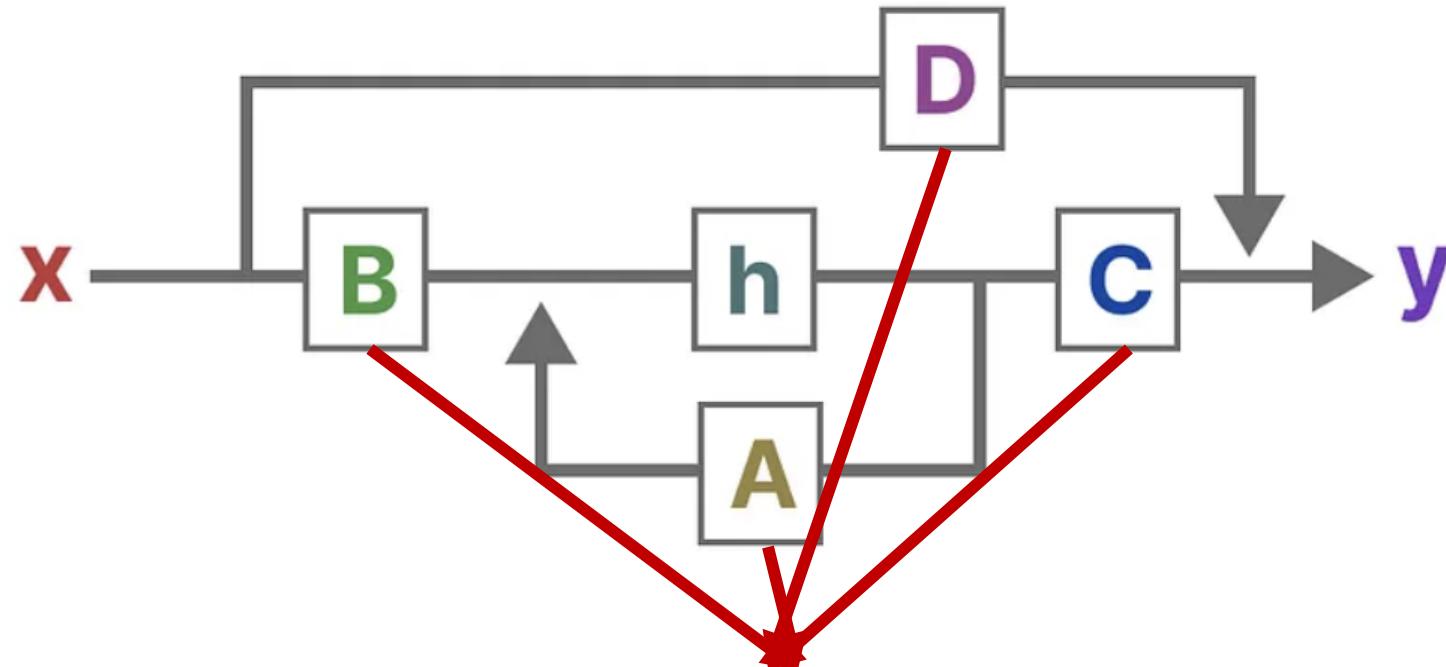


Simplified SSM Model Representation

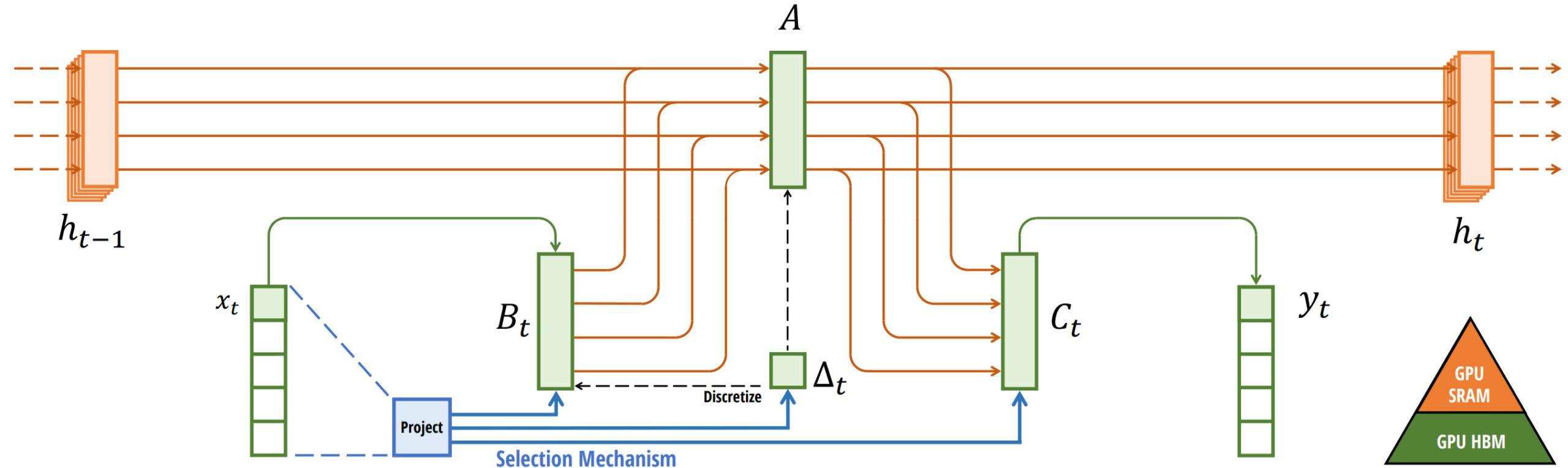


Continuous domain representation of SSMs

Simplified SSM Model Representation



Selective SSM or S6



Convert to discrete domain for LLMs!

Selective SSM or S6

State space equation

$$h(t) = \bar{A}(t) \odot h(t-1) + \bar{B}(t) \odot u(t); \quad o(t) = C(t)h(t)$$

Discretization

$$\bar{A}(t) = e^{(A\Delta(t))}; \quad B(t) = W_B(u(t)); \quad \bar{B}(t) = B\Delta(t)$$

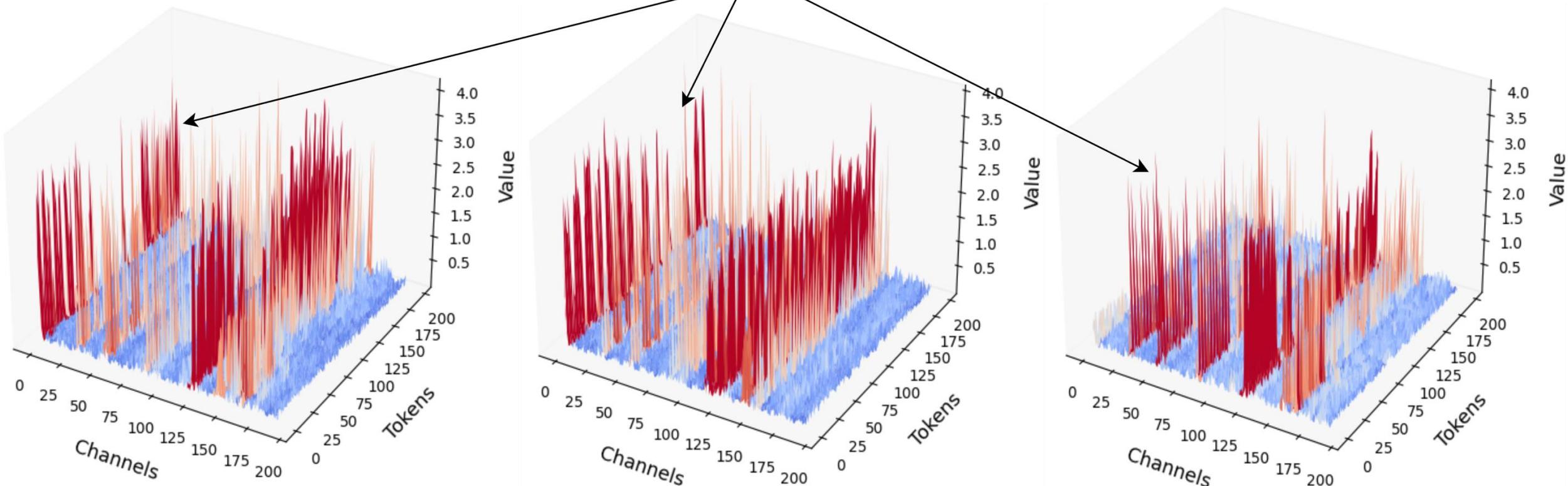
$$\Delta(t) = S^+(u(t)\Delta(t)_{\text{proj}}); \quad C(t) = (W_C(z(t)))^T$$

Selection Mechanism

Convert to discrete domain for LLMs!

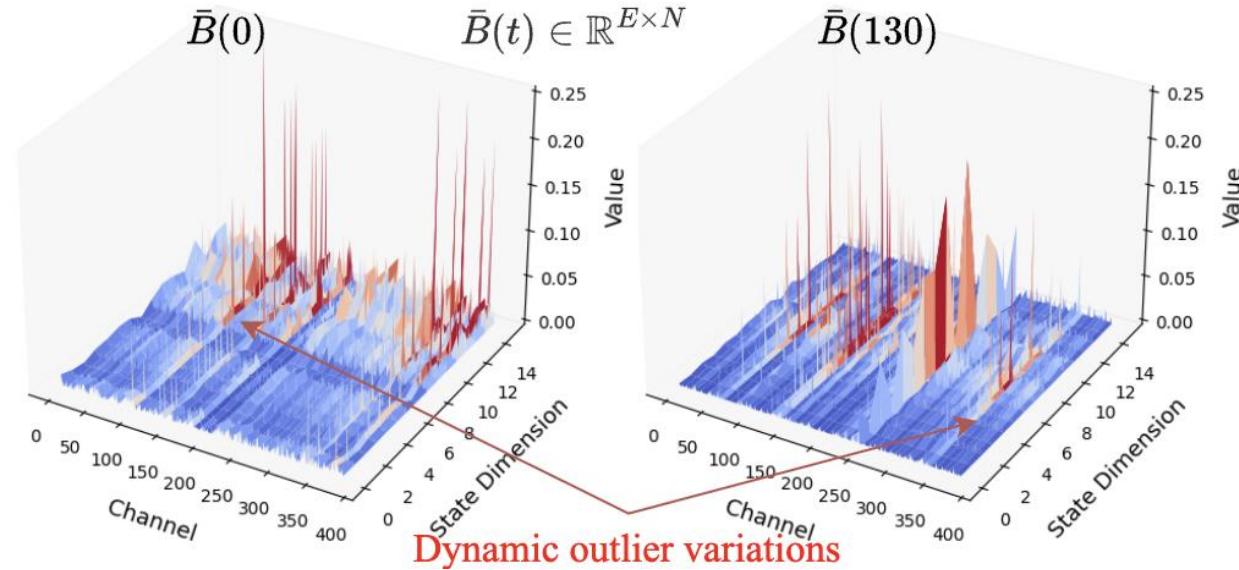
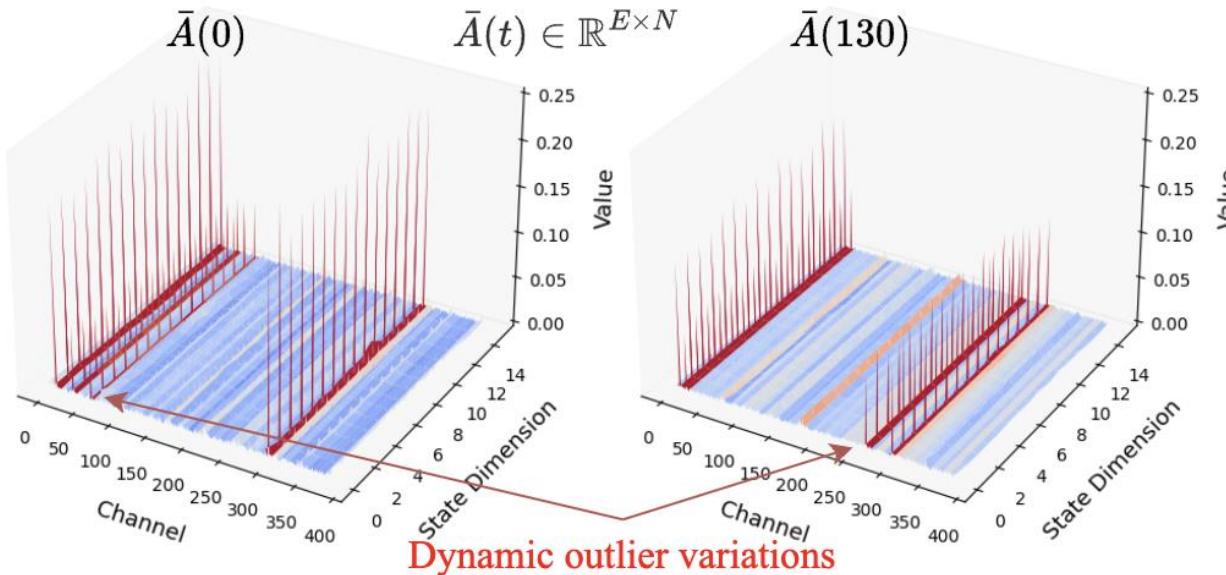
Transformers' Static Outlier Patterns

Static outlier
channels: 31, 55, 86 ...



DeiT Layer 9 for different inputs

Dynamic Activation Variations



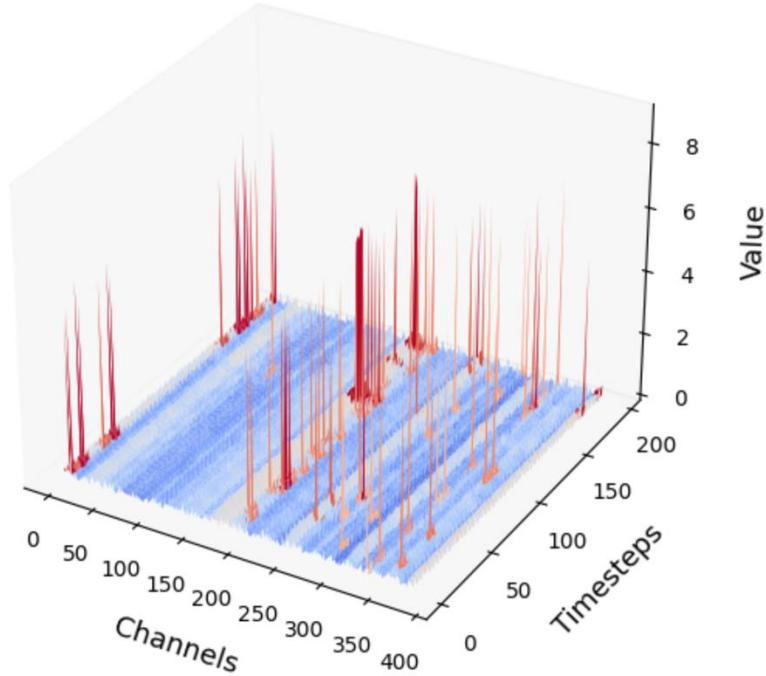
$$\bar{A}(t) = e^{(A\Delta(t))}; B(t) = W_B(u(t)); \bar{B}(t) = B\Delta(t)$$

$$\Delta(t) = S^+(u(t)\Delta(t)_{\text{proj}}); C(t) = (W_C(z(t)))^T$$

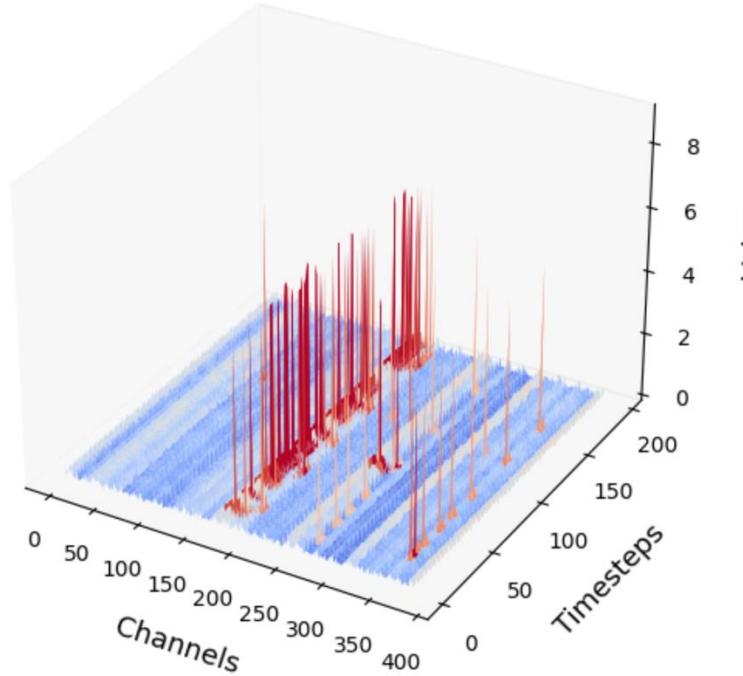
VMM activations exhibit dynamic inter-time-step channel variations

Dynamic Activation Variations

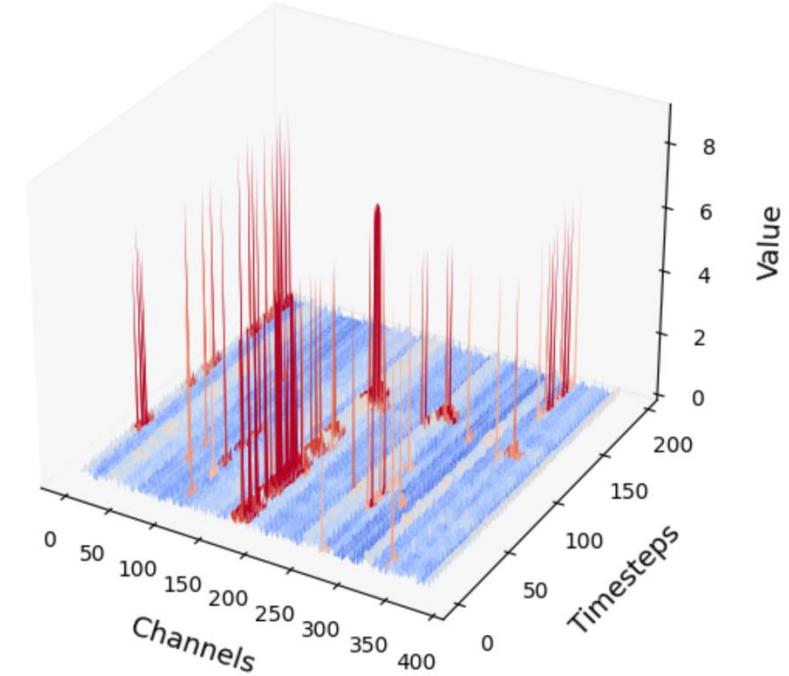
ViM-S Layer 3



Δ



A



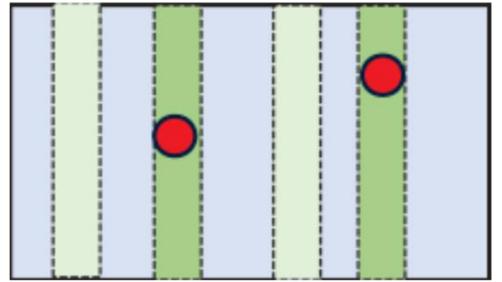
B

Dynamic variations across different activations too!

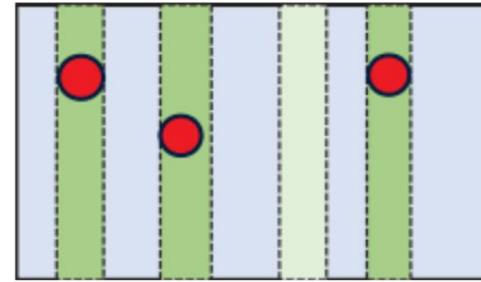
Shortcoming of Static Determination

Overprovisioning for outliers during calibration

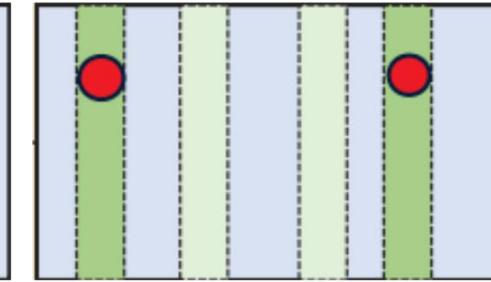
During Calibration



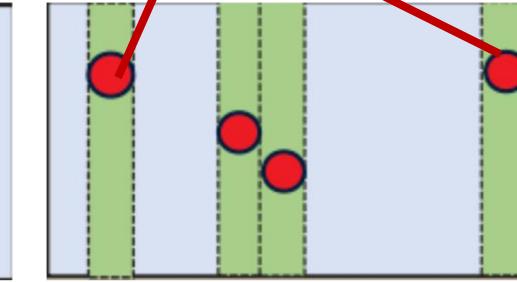
Time step 1



Time step 2

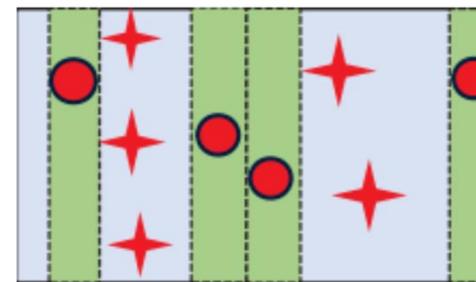


Time step 3

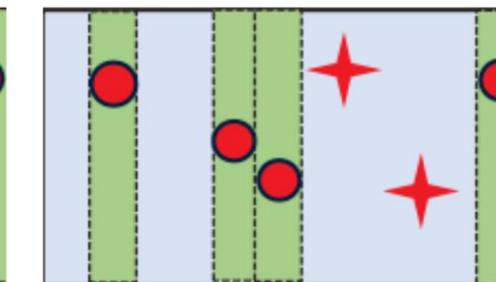


Time step 4

During inference



Time step 1



Time step 2

Misses new outlier channels

OuroMamba-Quant

Input : Activation $X(t) \in \mathbb{R}^{N \times E}$, Static scale $S^I(t)$, Threshold θ , Refresh rate n_{refresh} , Outlier list O_{list} , Inlier and outlier bit-precision b_a^I, b_a^O

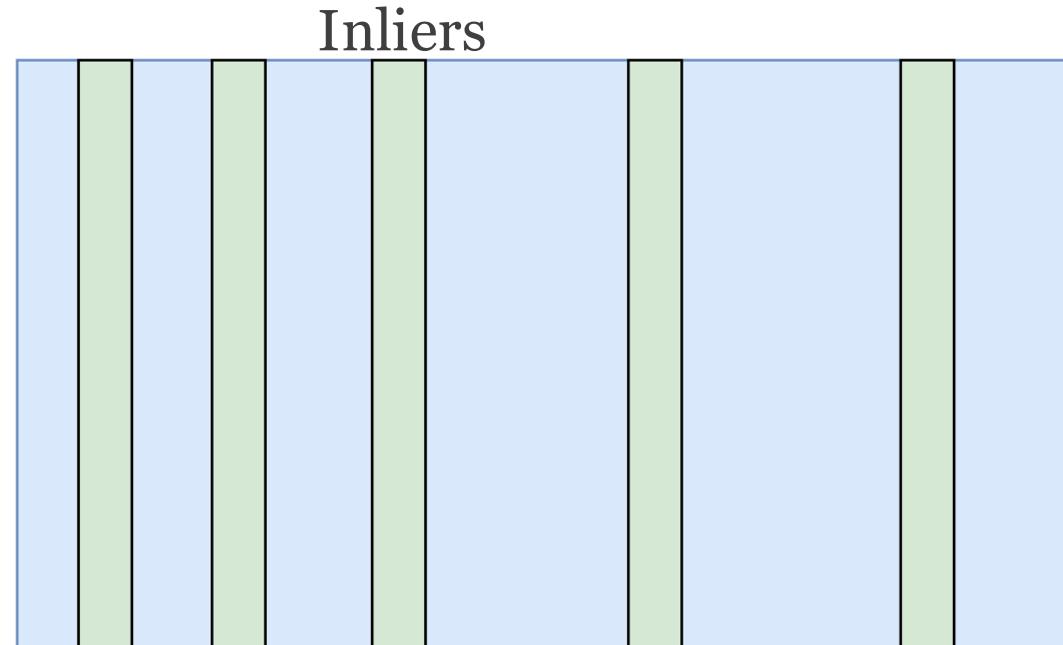
Output: Quantized activation $X_q(t)$, Updated outlier list O_{list}

```

if  $t \% n_{\text{refresh}} == 0$  then
     $O_{\text{list}} = \{\phi\}$ 
end
 $S^D(t) = \text{ComputeScale}(X(t)[:, c] \forall c \notin O_{\text{list}})$ 
if  $S^D(t) > S^I(t)$  then
    for each channel  $c$  in  $X(t)$  not in  $O_{\text{list}}$  do
        if  $\max(|X(t)[:, c]|) \geq \theta$  then
             $O_{\text{list}} = O_{\text{list}} \cup \{c\}$ 
        end
    end
end
 $I(t), O(t) = \text{Separate}(X(t), O_{\text{list}})$ 
 $I_q(t) = \text{InlierQuant}(I(t), S^I(t), b_a^I)$ 
 $O_q(t) = \text{OutlierQuant}(O(t), b_a^O)$ 
 $X_q(t) = \text{Merge}(I_q(t), O_q(t))$ 
return  $X_q(t), O_{\text{list}}$ 

```

Outlier List: {}



Offline: Determine threshold to detect outliers and inlier scale factor

OuroMamba-Quant

Input : Activation $X(t) \in \mathbb{R}^{N \times E}$, Static scale $S^I(t)$, Threshold θ , Refresh rate n_{refresh} ,
Outlier list O_{list} , Inlier and outlier bit-precision b_a^I, b_a^O

Output: Quantized activation $X_q(t)$, Updated outlier list O_{list}

if $t \% n_{\text{refresh}} == 0$ **then**

 | $O_{\text{list}} = \{\phi\}$

end

$S^D(t) = \text{ComputeScale}(X(t)[:, c] \forall c \notin O_{\text{list}})$

if $S^D(t) > S^I(t)$ **then**

 | **for** each channel c in $X(t)$ **not in** O_{list} **do**

 | **if** $\max(|X(t)[:, c]|) \geq \theta$ **then**

 | $O_{\text{list}} = O_{\text{list}} \cup \{c\}$

 | **end**

 | **end**

end

$I(t), O(t) = \text{Separate}(X(t), O_{\text{list}})$

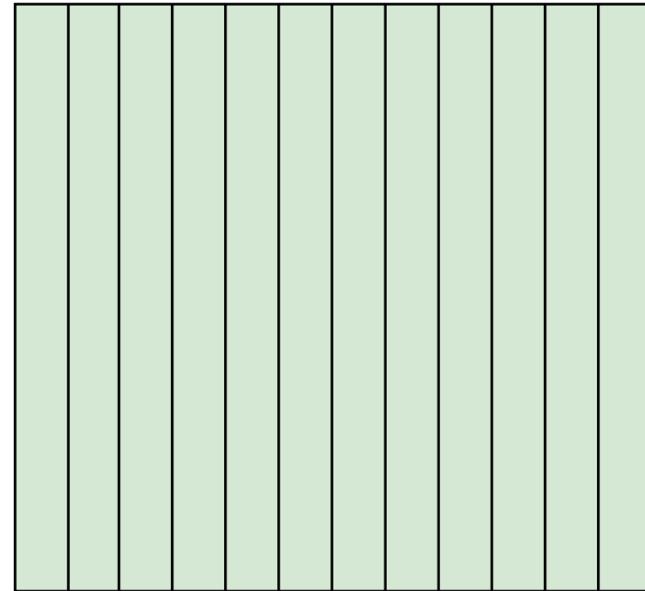
$I_q(t) = \text{InlierQuant}(I(t), S^I(t), b_a^I)$

$O_q(t) = \text{OutlierQuant}(O(t), b_a^O)$

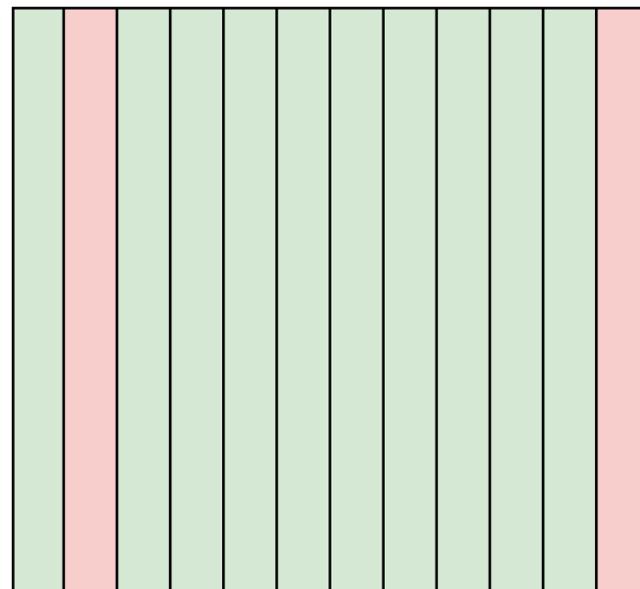
$X_q(t) = \text{Merge}(I_q(t), O_q(t))$

return $X_q(t), O_{\text{list}}$

Outlier List: {}



Outlier List: {1,11}



OuroMamba-Quant

Input : Activation $X(t) \in \mathbb{R}^{N \times E}$, Static scale $S^I(t)$, Threshold θ , Refresh rate n_{refresh} ,
Outlier list O_{list} , Inlier and outlier bit-precision b_a^I, b_a^O

Output: Quantized activation $X_q(t)$, Updated outlier list O_{list}

if $t \% n_{\text{refresh}} == 0$ **then**

 | $O_{\text{list}} = \{\phi\}$

end

$S^D(t) = \text{ComputeScale}(X(t)[:, c] \forall c \notin O_{\text{list}})$

if $S^D(t) > S^I(t)$ **then**

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 | **if** $\max(|X(t)[:, c]|) \geq \theta$ **then**

 | $O_{\text{list}} = O_{\text{list}} \cup \{c\}$

 | **end**

 | **end**

end

$I(t), O(t) = \text{Separate}(X(t), O_{\text{list}})$

$I_q(t) = \text{InlierQuant}(I(t), S^I(t), b_a^I)$

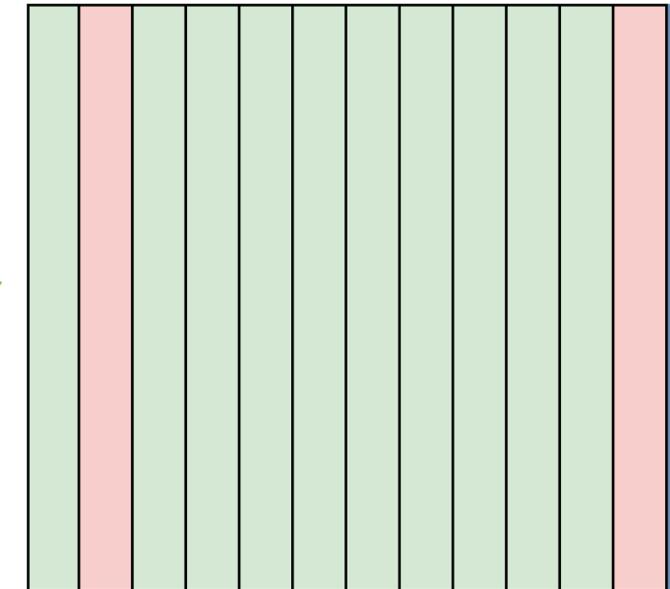
$O_q(t) = \text{OutlierQuant}(O(t), b_a^O)$

$X_q(t) = \text{Merge}(I_q(t), O_q(t))$

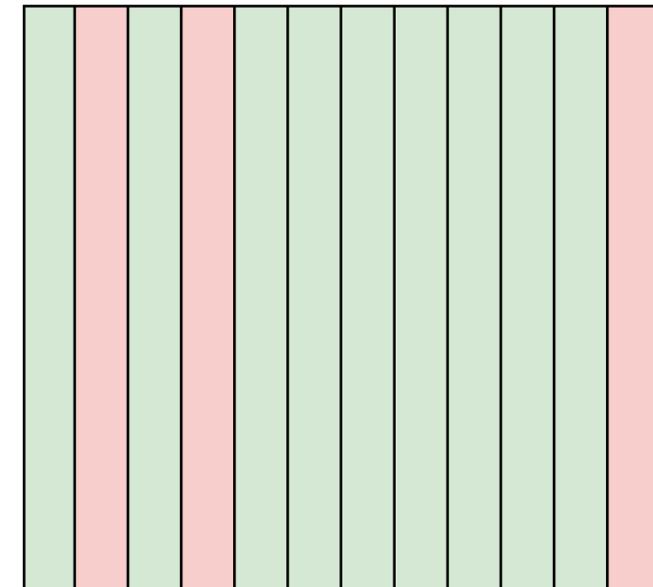
return $X_q(t), O_{\text{list}}$

We must now identify which channel is the outlier channel

Outlier List: {1,11}



Outlier List: {1,3,11}



OuroMamba-Quant

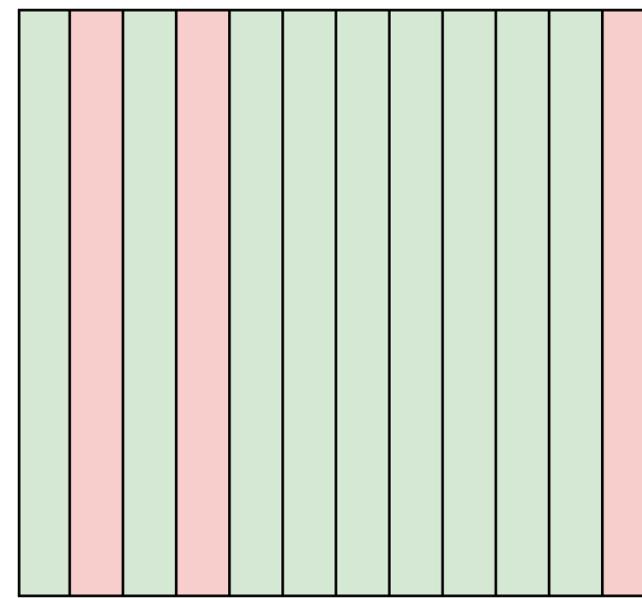
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Output: Quantized activation $X_q(t)$, Updated outlier list O_{list}

```

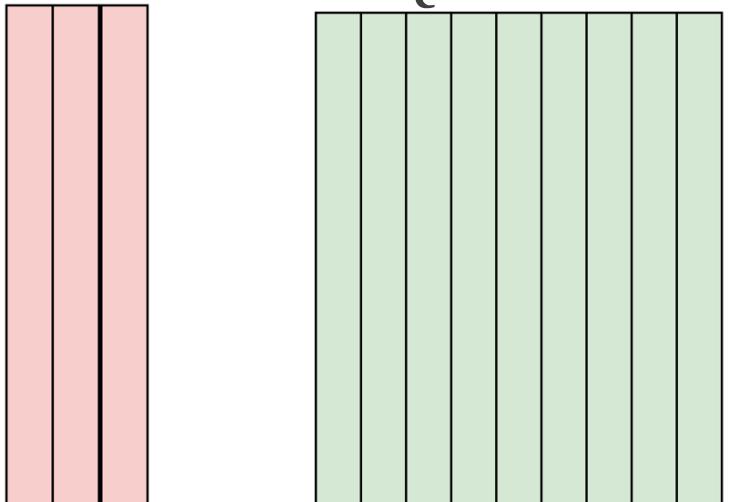
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|   |   end
|   end
end
|    $I(t), O(t) = \text{Separate}(X(t), O_{\text{list}})$ 
|    $I_q(t) = \text{InlierQuant}(I(t), S^I(t), b_a^I)$ 
|    $O_q(t) = \text{OutlierQuant}(O(t), b_a^O)$ 
|    $X_q(t) = \text{Merge}(I_q(t), O_q(t))$ 
return  $X_q(t), O_{\text{list}}$ 

```



Outlier Quantization

Inlier Quantization



OuroMamba-Quant

Input : Activation $X(t) \in \mathbb{R}^{N \times E}$, Static scale $S^I(t)$, Threshold θ , Refresh rate n_{refresh} ,
Outlier list O_{list} , Inlier and outlier bit-precision b_a^I, b_a^O

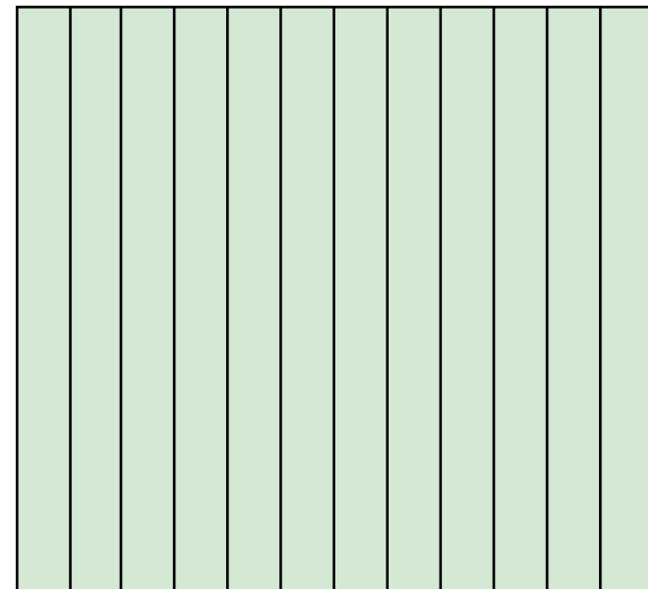
Output: Quantized activation $X_q(t)$, Updated outlier list O_{list}

```

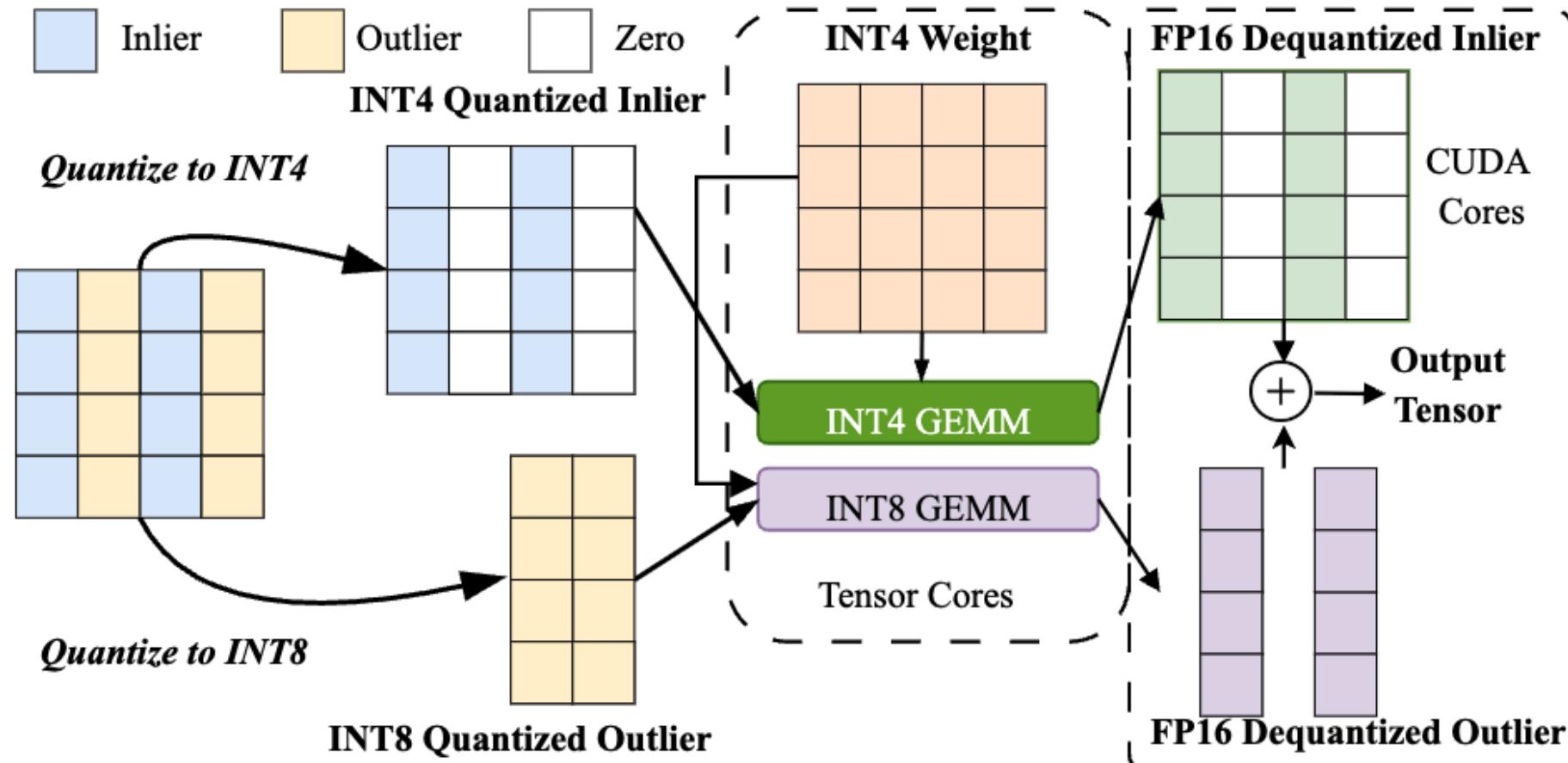
if  $t \% n_{\text{refresh}} == 0$  then
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|   |   end
|   end
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 $X_q(t) = \text{Merge}(I_q(t), O_q(t))$ 
return  $X_q(t), O_{\text{list}}$ 

```

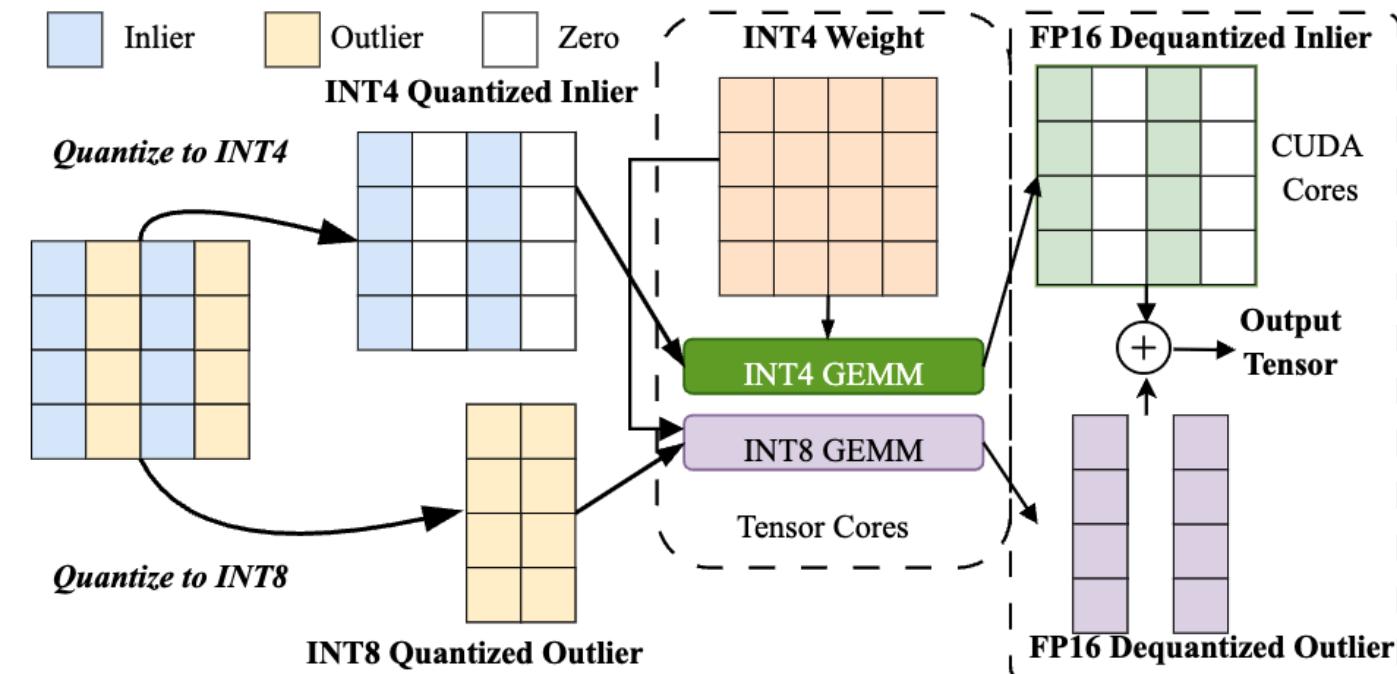
Outlier List: {}



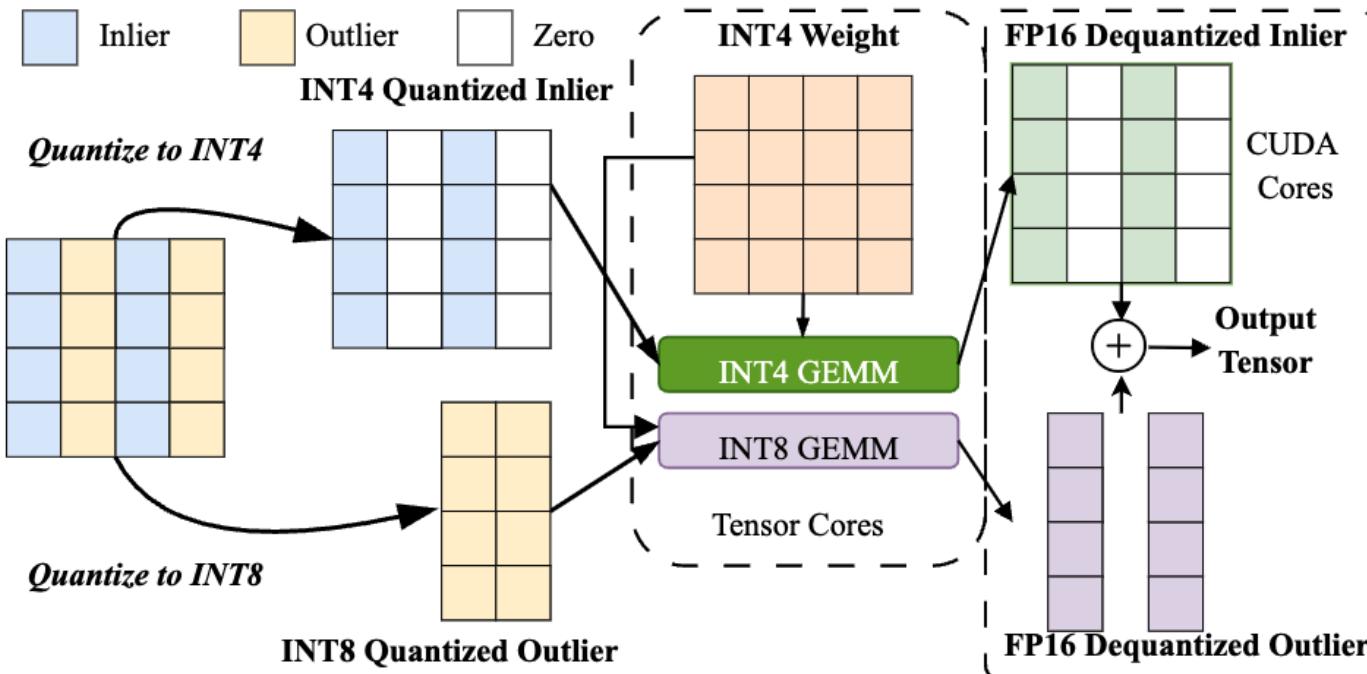
OuroMamba-Quant: W4A4 Hybrid GEMM



OuroMamba-Quant: W4A4 Hybrid GEMM

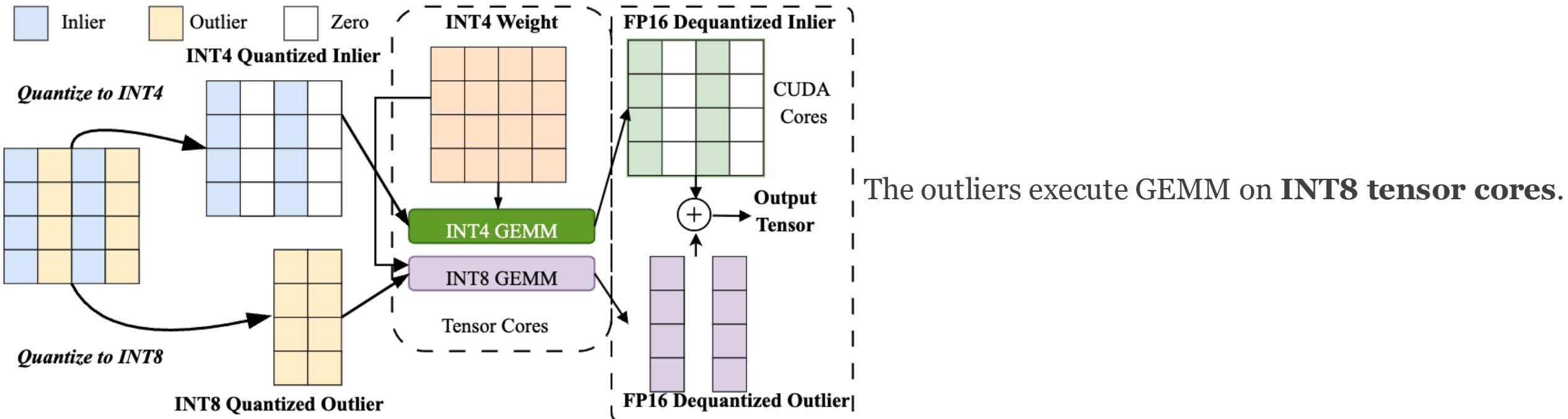


OuroMamba-Quant: W4A4 Hybrid GEMM

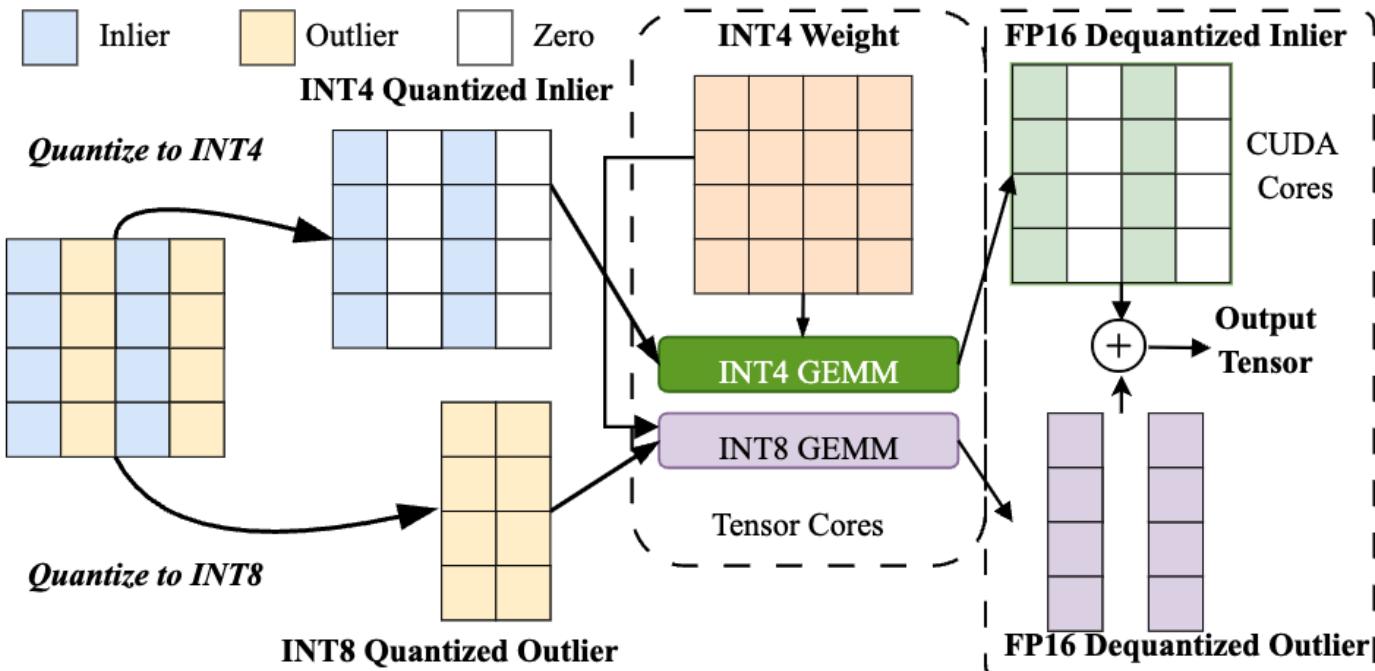


We pack two consecutive 4-bit inlier activations into one byte, with outlier positions set to zero, and leverage the **INT4 tensor cores for inlier GEMM**.

OuroMamba-Quant: W4A4 Hybrid GEMM

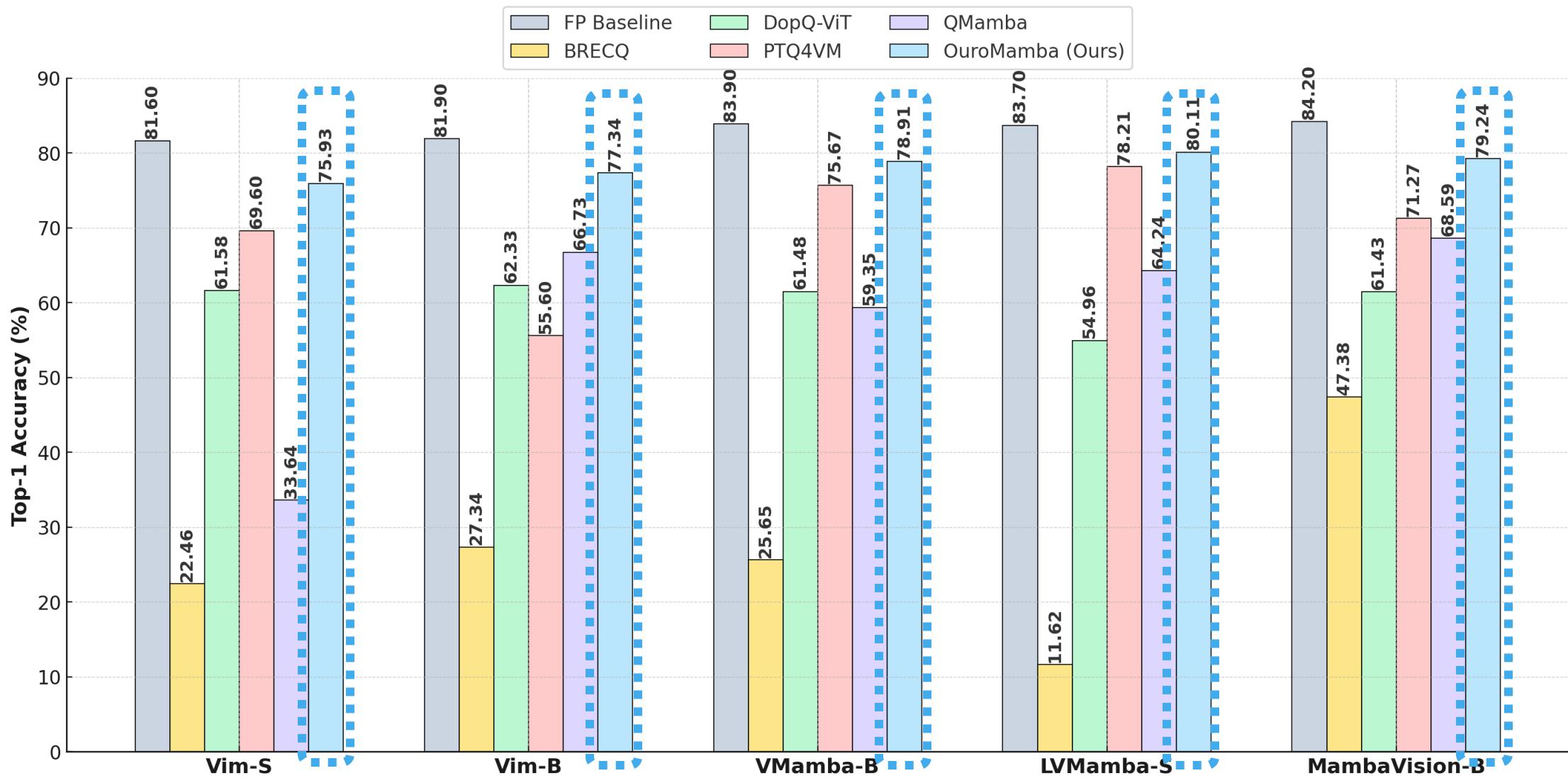


OuroMamba-Quant: W4A4 Hybrid GEMM

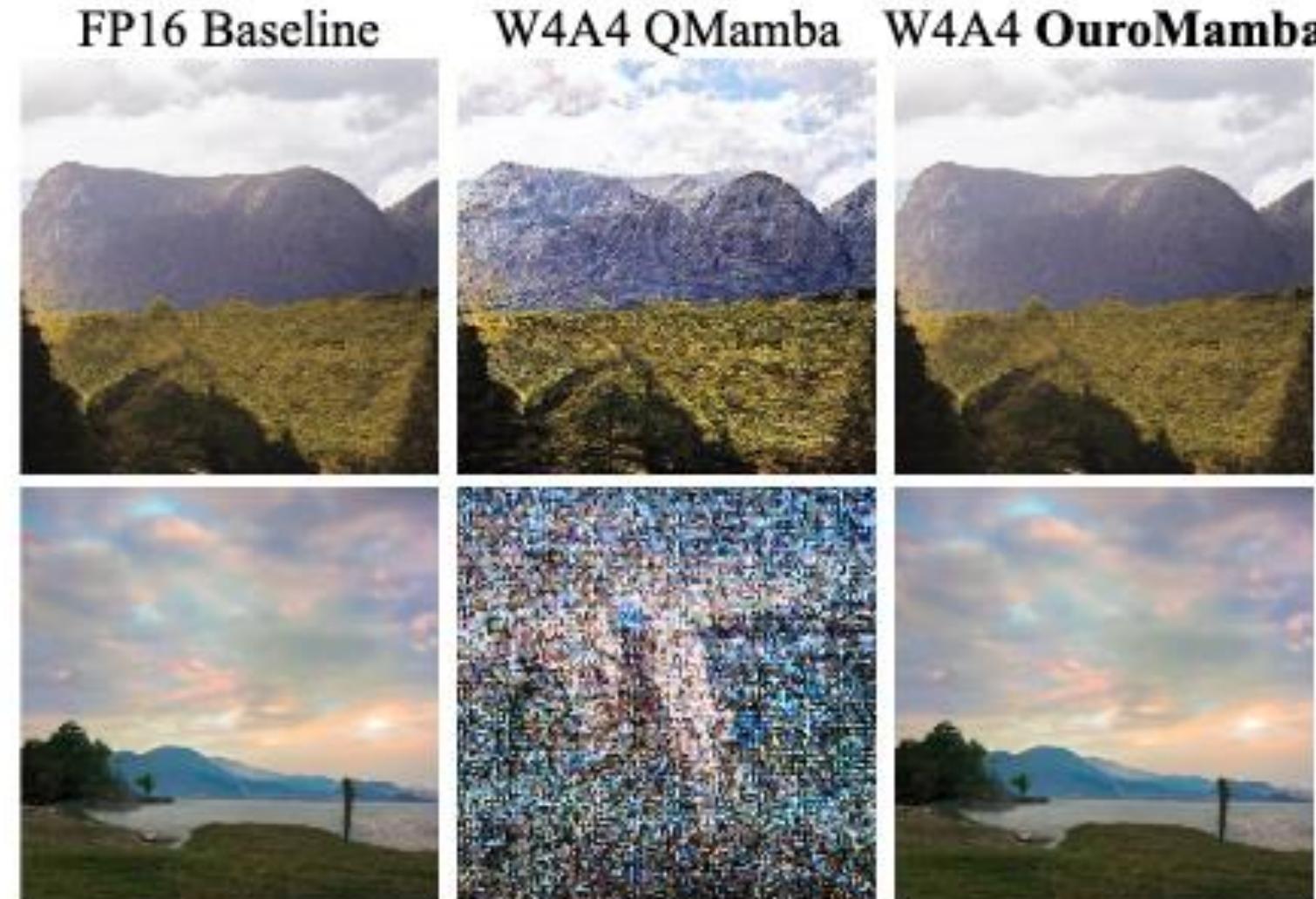


The outputs from the inlier and outlier GEMMs are dequantized to FP16 and summed together.

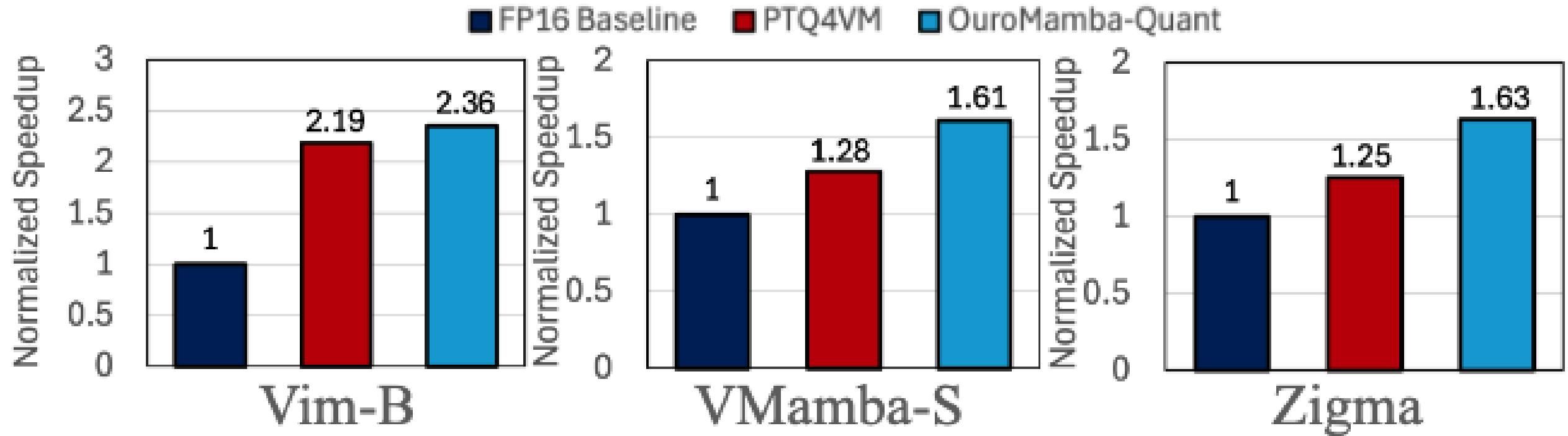
Quantization Results: W4A4



Quantization Results for Diffusion Models



End-to-End Latency Comparison

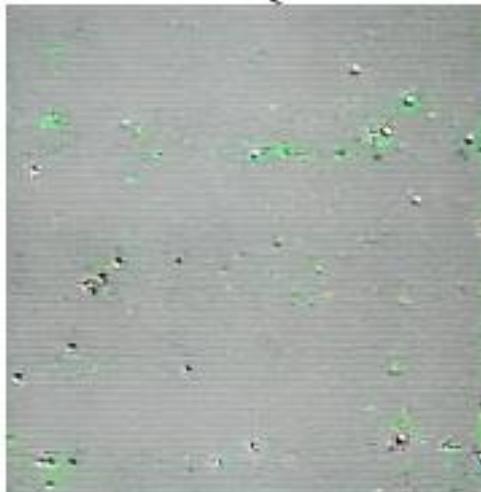


Generalization to ViTs

FP16 Baseline



W4A8 Q-DiT



W4A8 PTQ4DiT



W4A8 OuroMamba



W4A4 OuroMamba



Prompt: An astronaut relaxing on a beach chair, sipping coffee on Mars, with Earth visible in the sky