**NIPS 2017** 

**Attention Is All You Need** 

**ICLR 2021** 

AN IMAGE IS WORTH 16X16 WORDS:
TRANSFORMERS FOR IMAGE RECOGNITION AT SCALE

Micosoft 2021

Swin Transformer: Hierarchical Vision Transformer using Shifted Windows

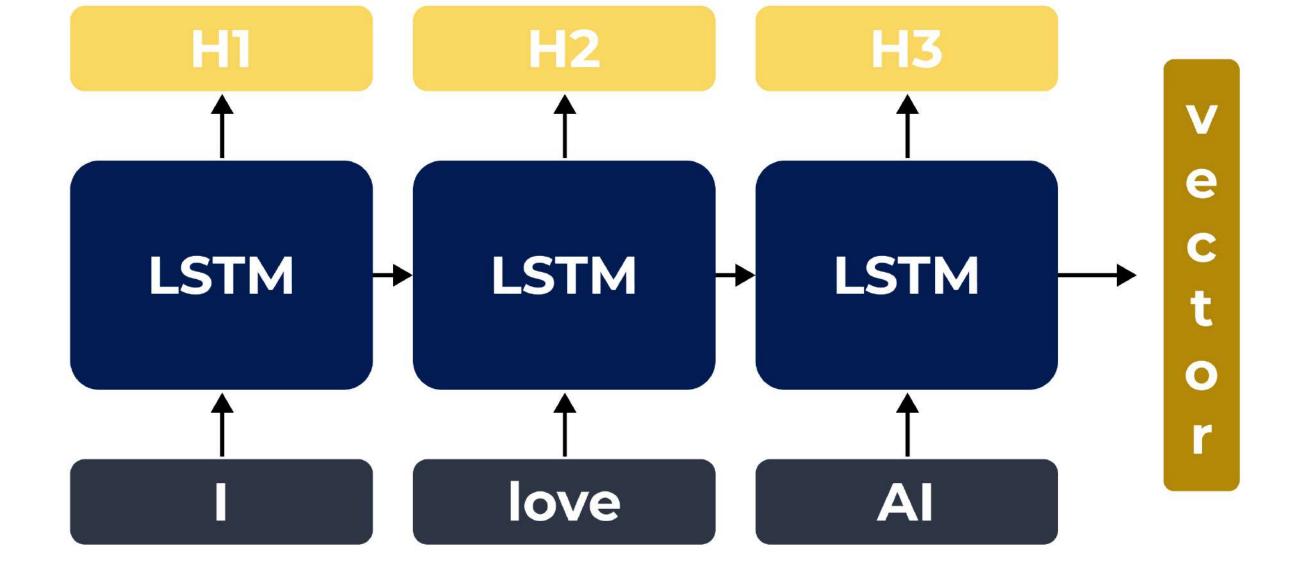
Minsu koh

## Transformer

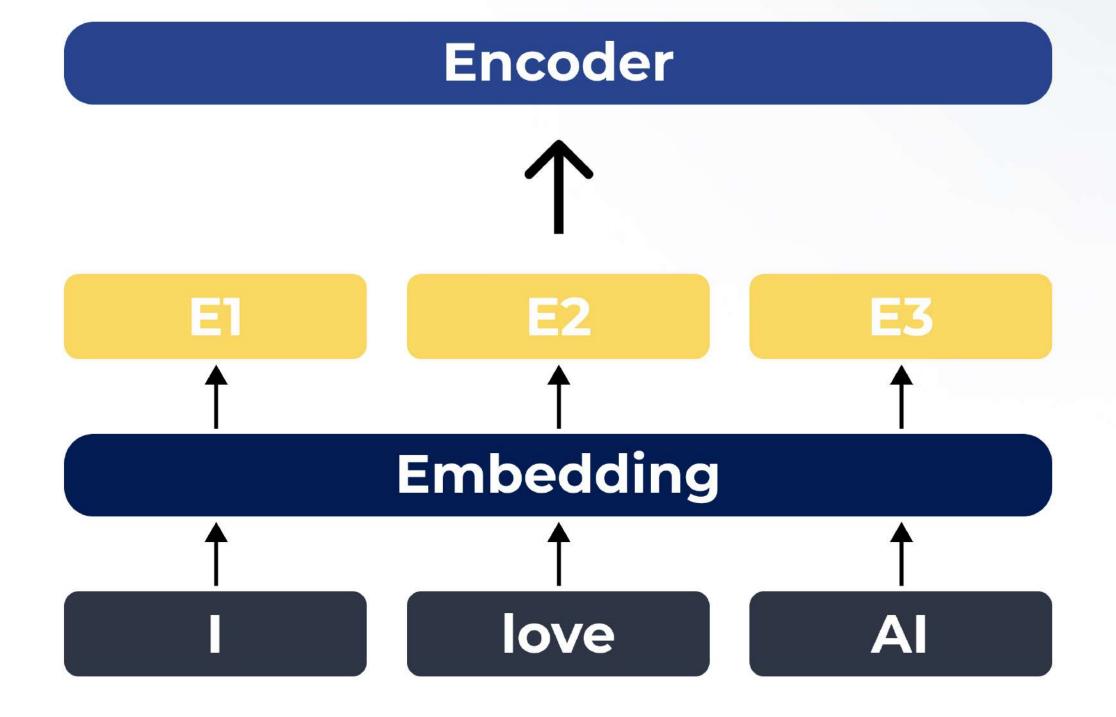
**Attention Is All You Need** 

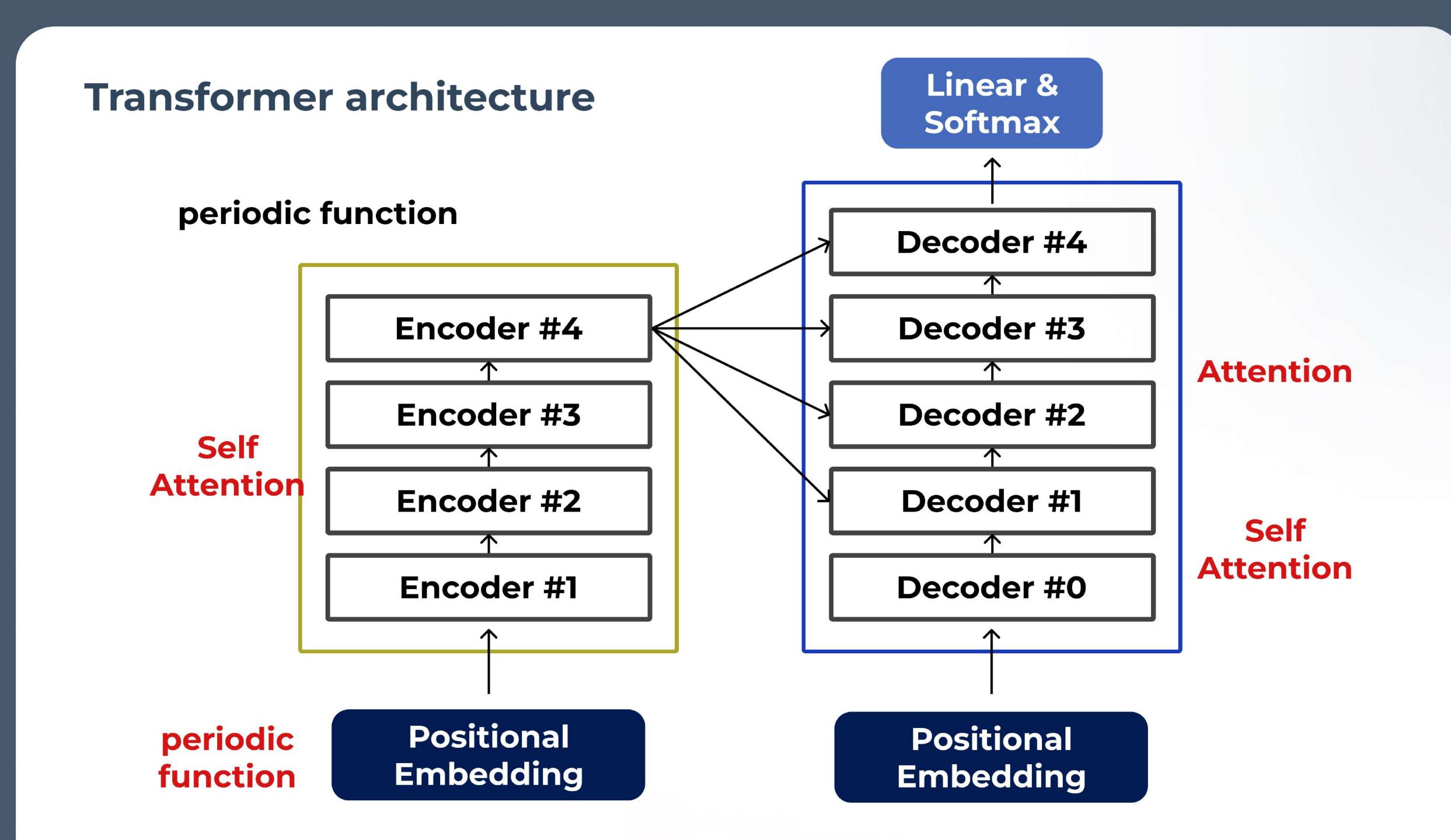
## Why we need Transformer?

### Seq2Seq

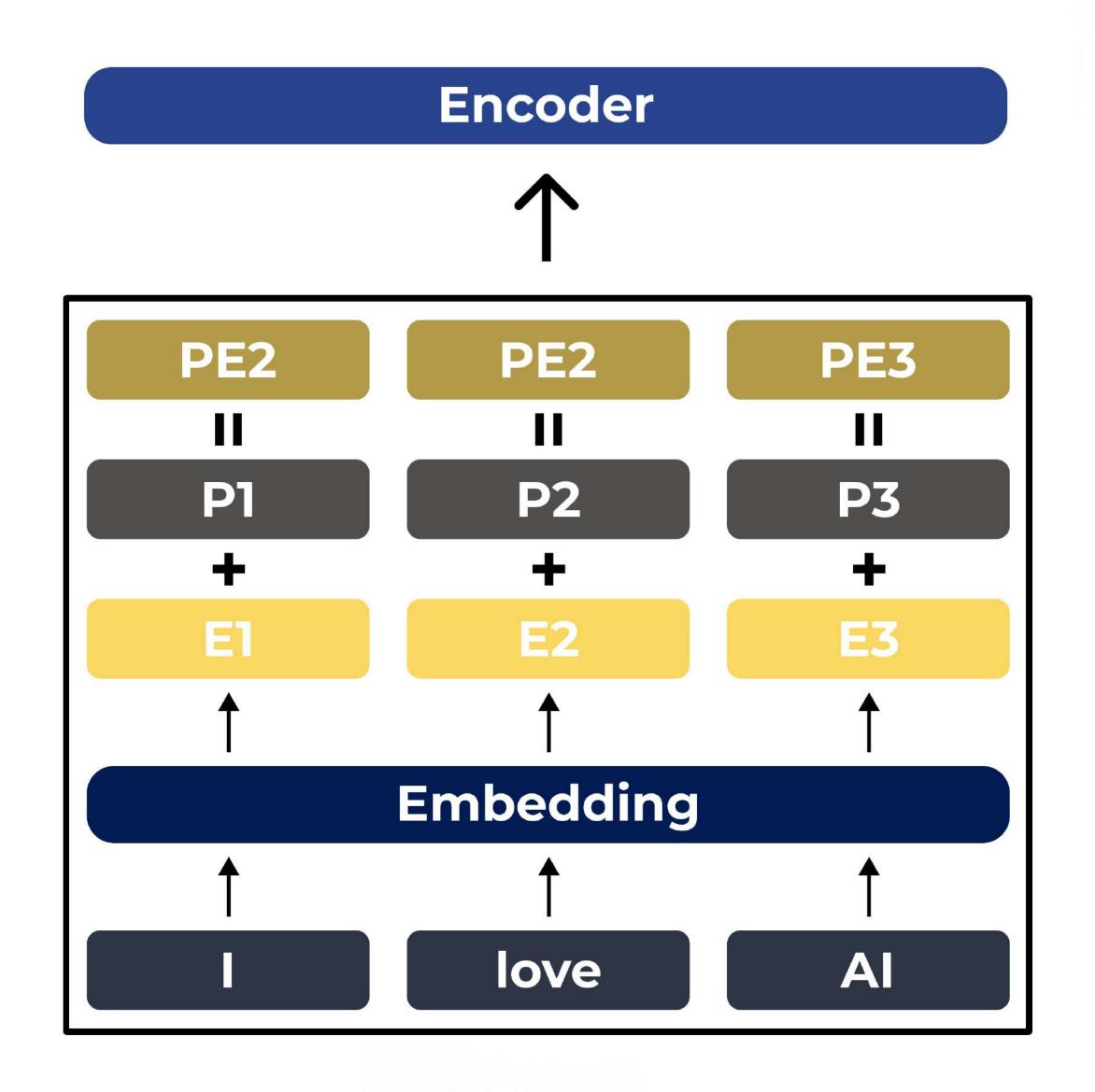


#### Transformer

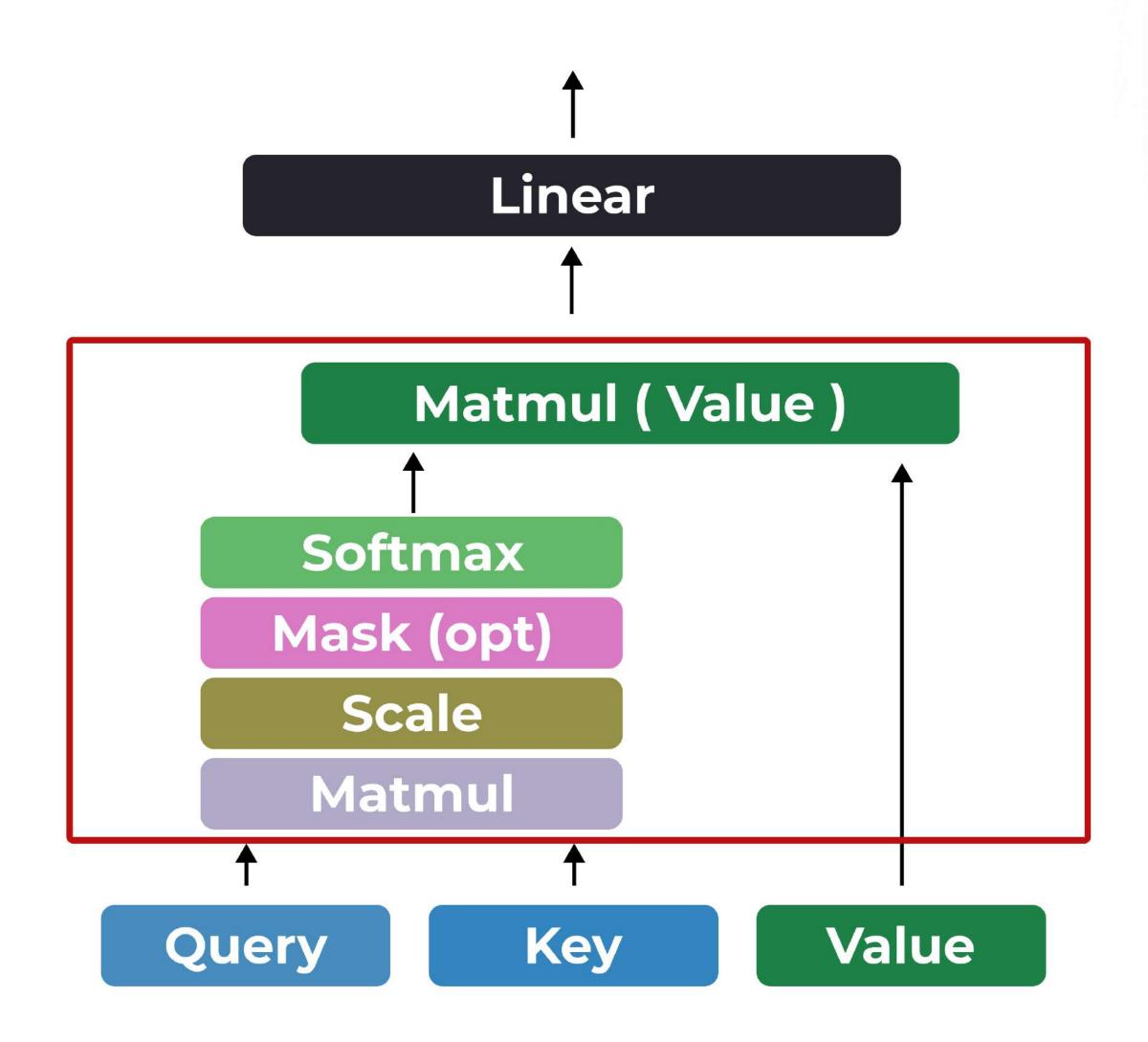




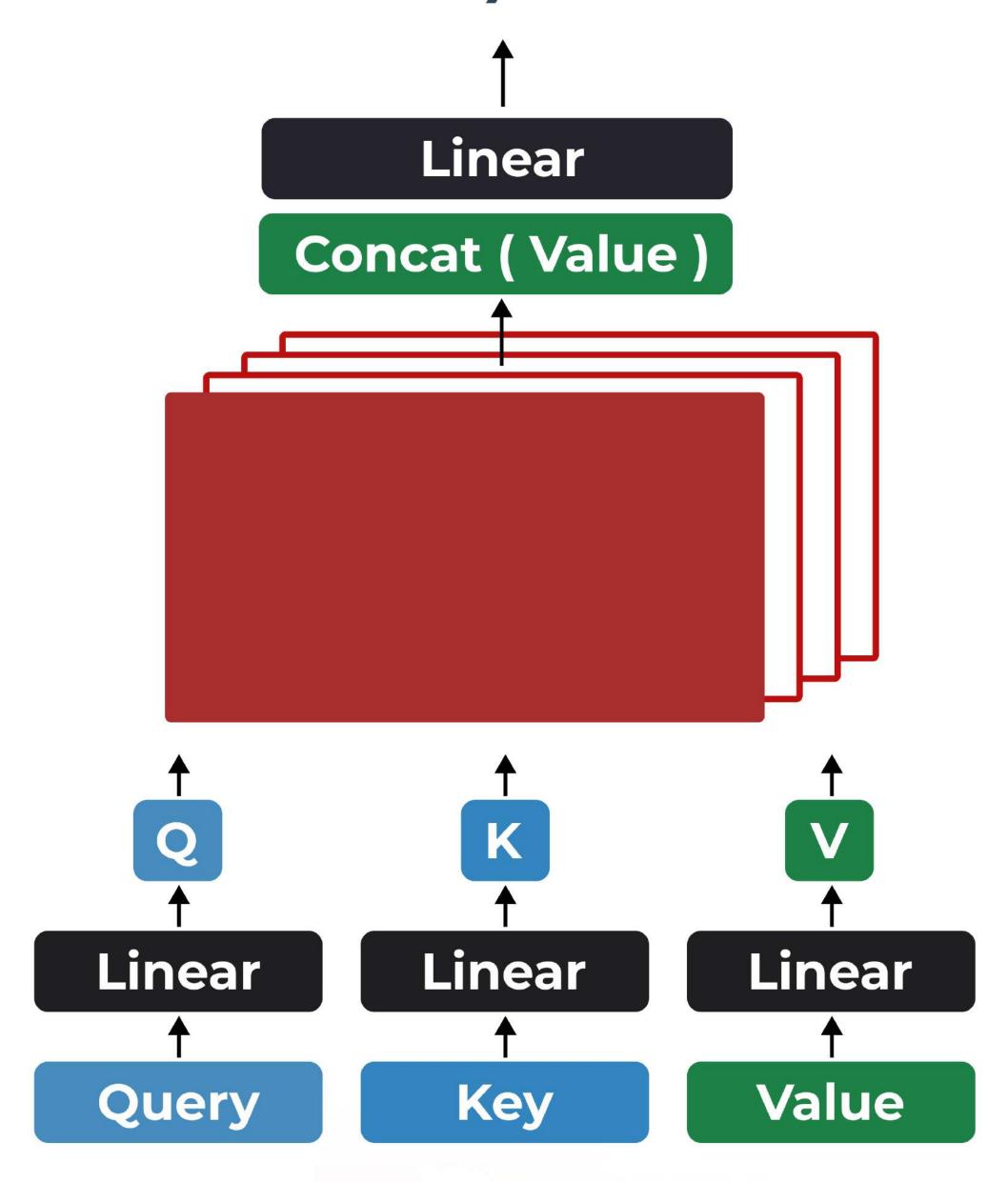
## Positional Embedding



## **Attention (One head Attention)**



## Attention (Multi head Attention)



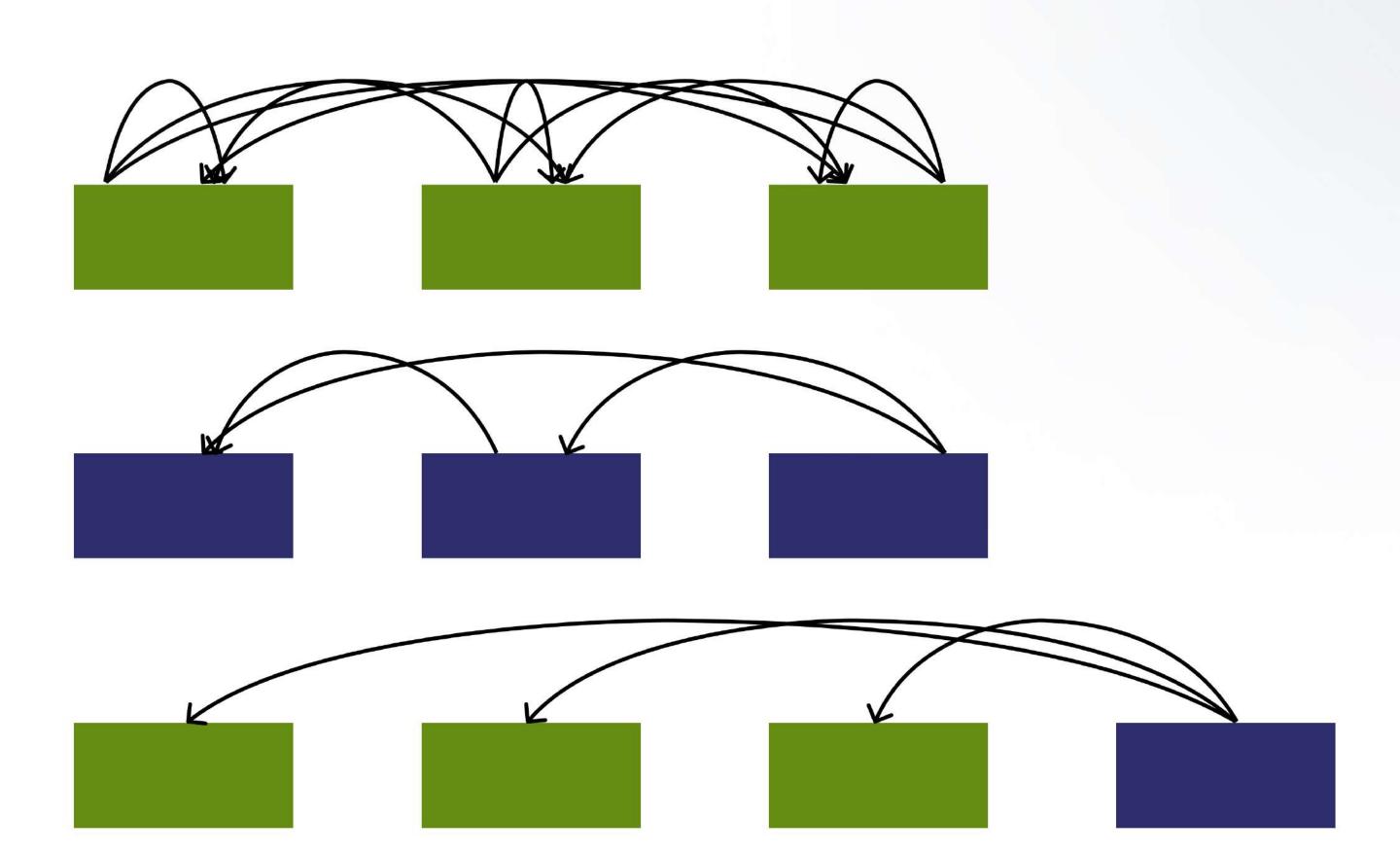
# **Self Attention** Linear Linear Linear Value Key Key

## Attention type

**Encoder Self Attention** 

**Masked Decoder Self Attention** 

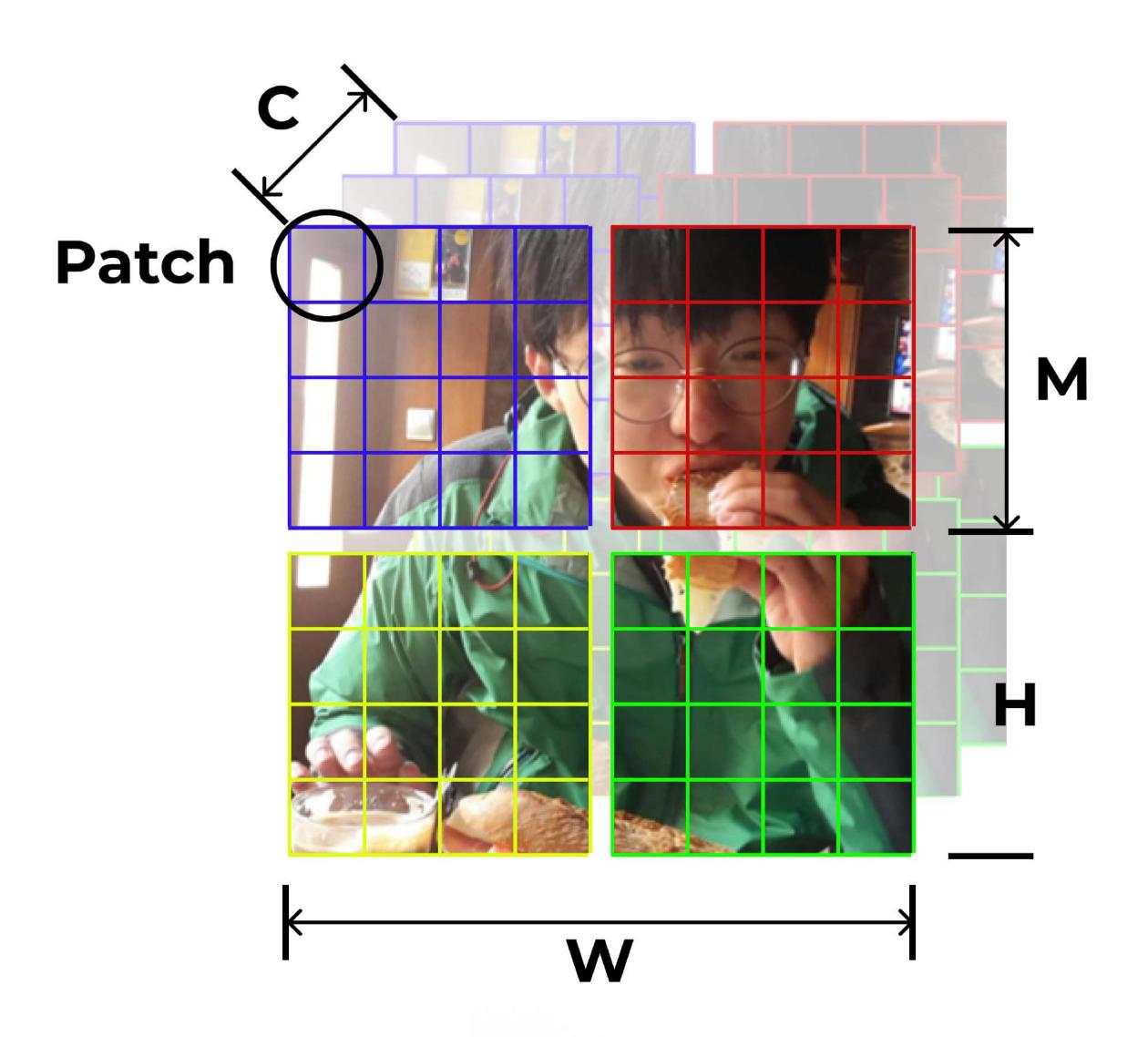
**Encoder-Decoder Attention** 



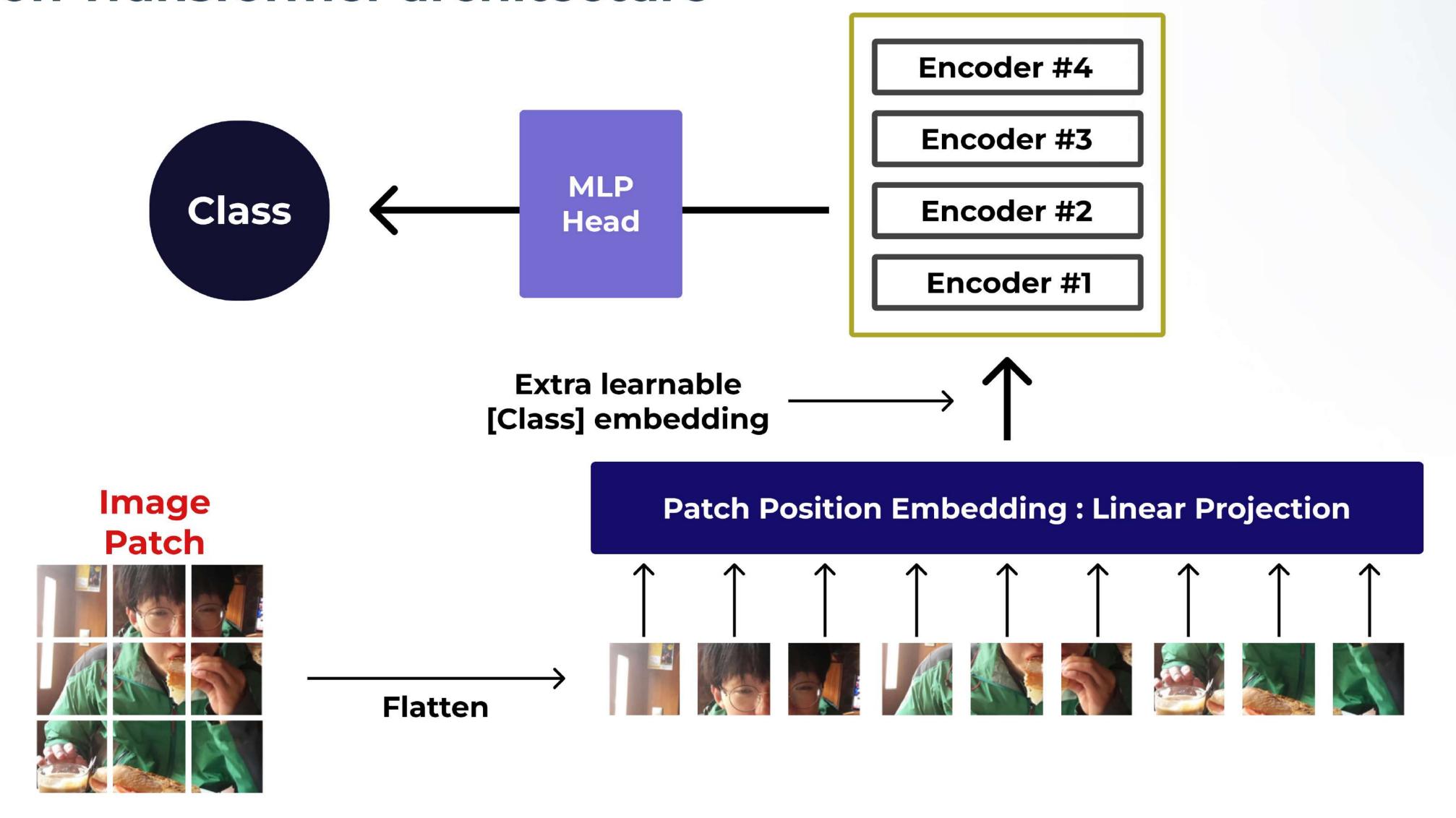
## Vision Transformer

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## Notice



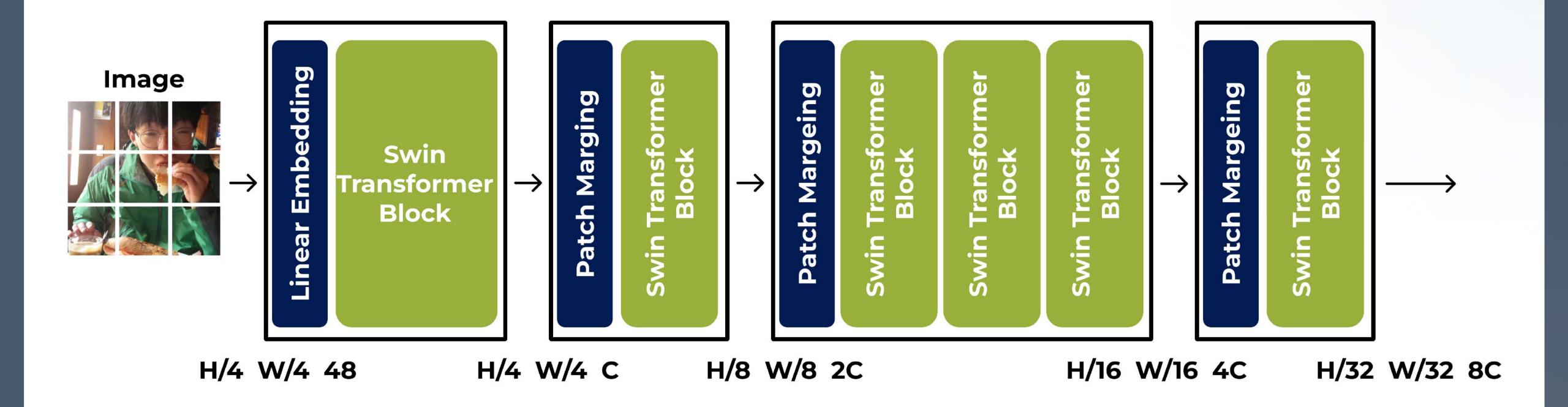
#### Vision Transformer architecture



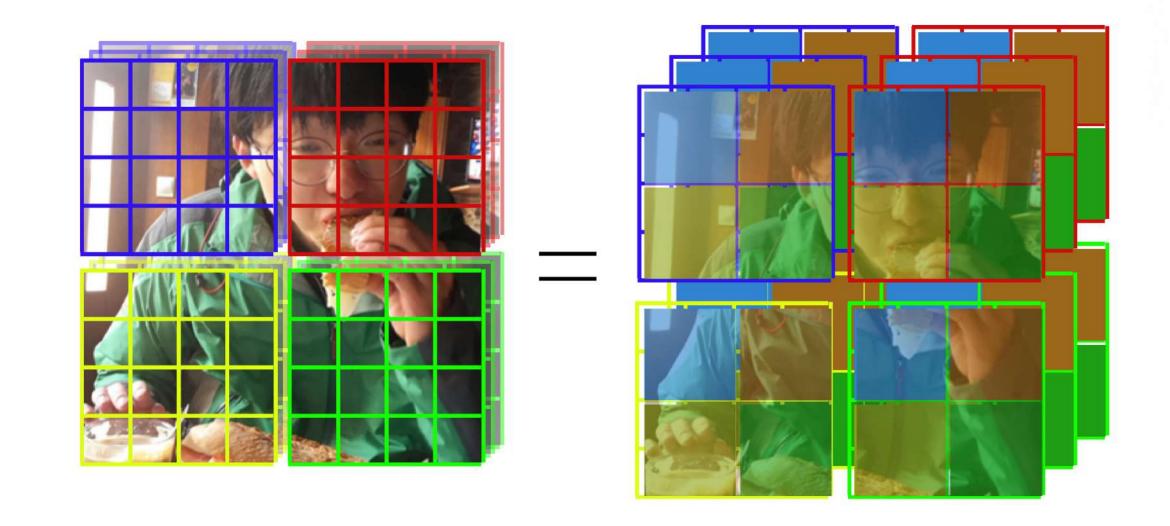
## Swin Transformer

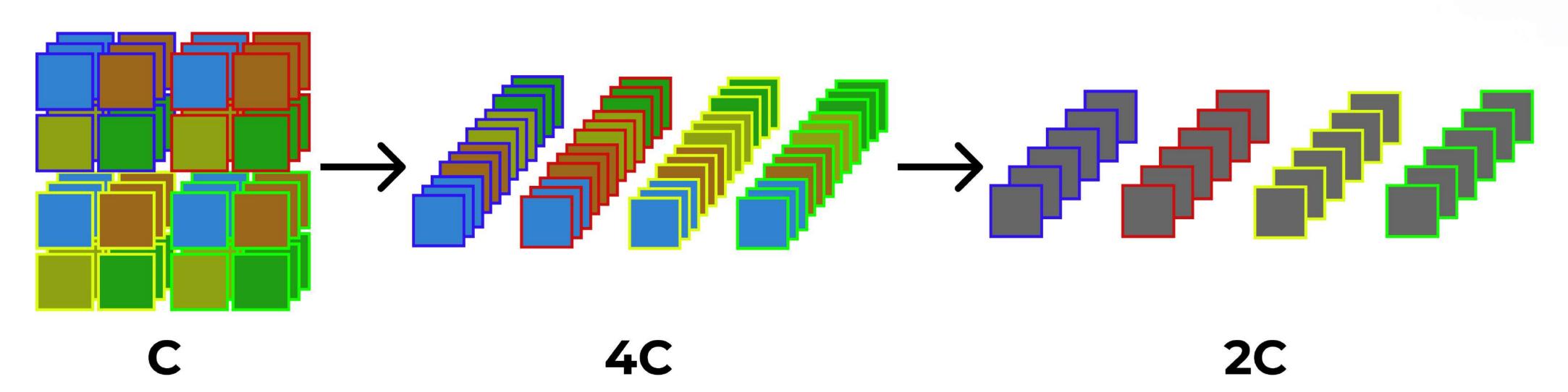
Swin Transformer:
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#### Swin Transformer achitecture



## Patch Marging

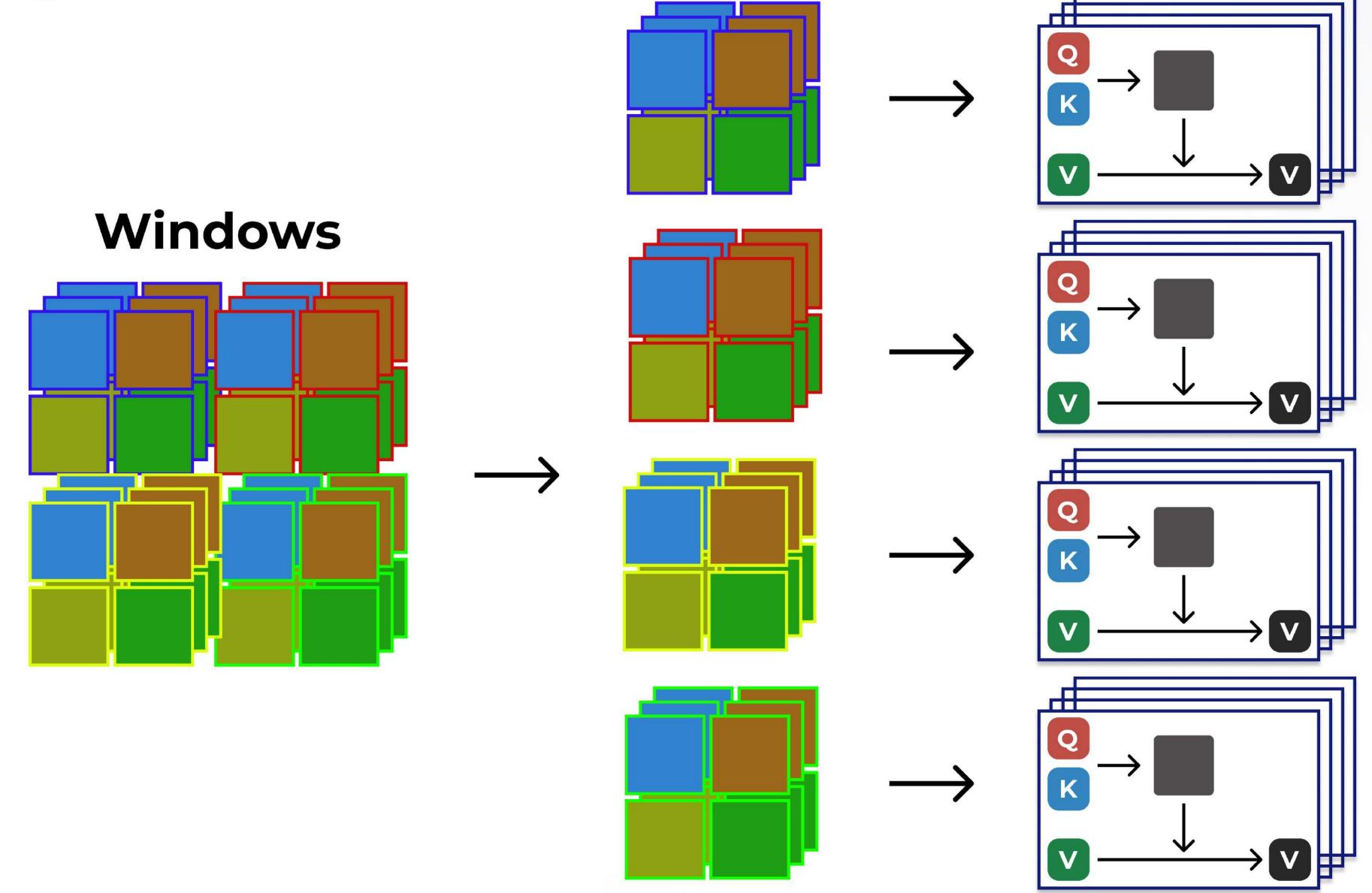




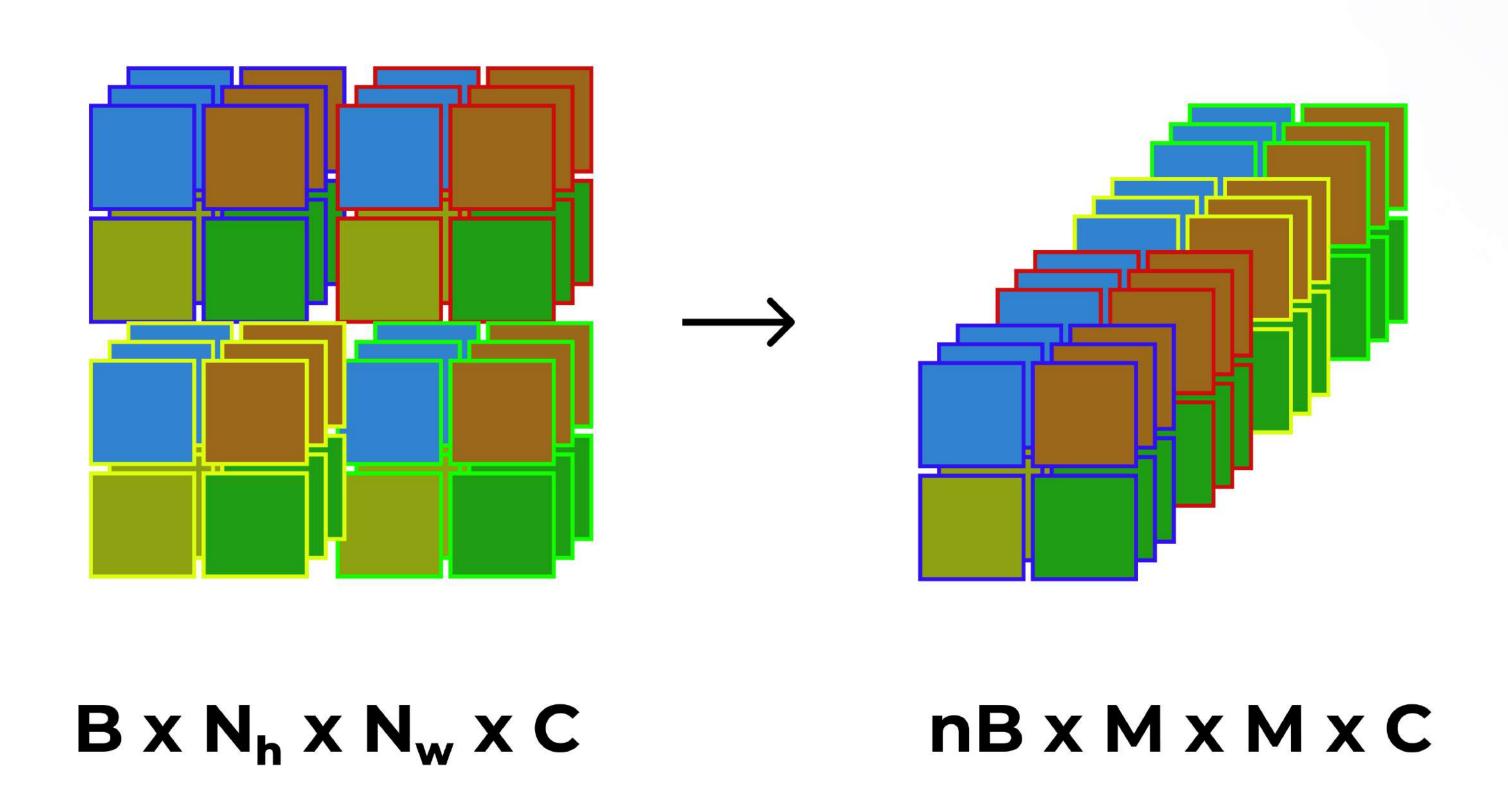
# **Swin Transformer Block MLP MLP Layer Normalization Layer Normalization** W-MSA **SW-MSA Layer Normalization Laver Normalization**

#### W-MSA

### **Multi head Self Attention**

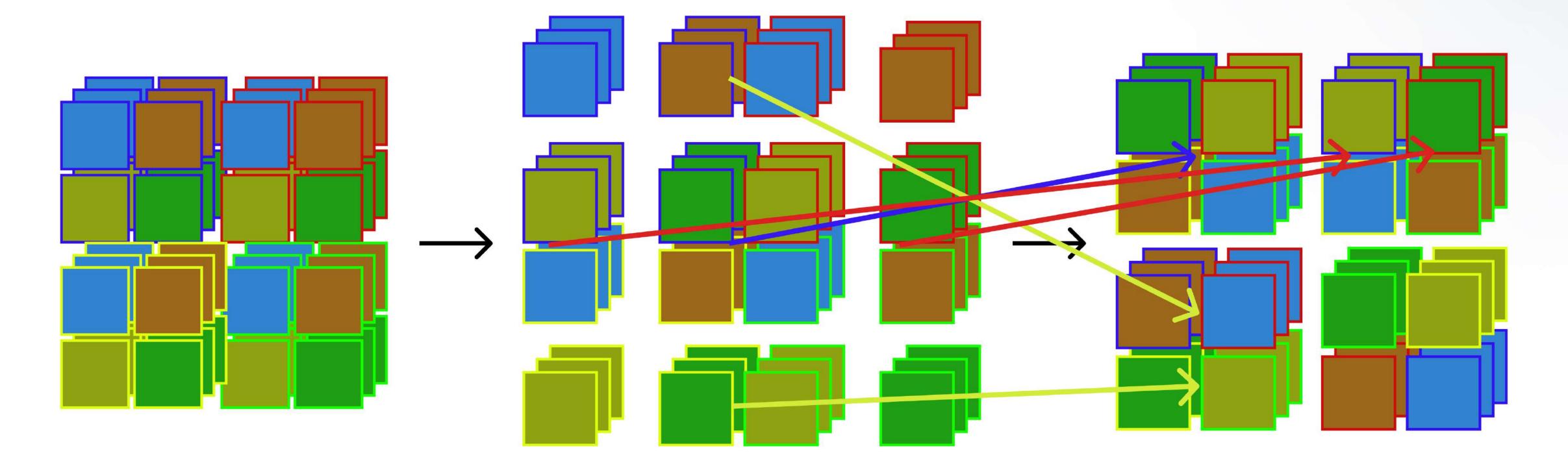


## SW-MSA - (Efficient Batch Computation)



## SW-MSA - (Cycliv Shift)

## **Cyclic Shift**



## SW-MSA - (Relative Position Bias)

#### if window\_size = 3

Attention $(Q, K, V) = \text{SoftMax}(QK^T/\sqrt{d} + B)V$ ,

#### x-axis Matrix

|   | 952 |   |   |   |    |    |    |    |    |    |
|---|-----|---|---|---|----|----|----|----|----|----|
| 3 |     | 0 | 0 | 0 | -1 | -1 | -1 | -2 | -2 | -2 |
| 6 |     | 0 | 0 | 0 | -1 | -1 | -1 | -2 | -2 | -2 |
| 9 |     | 0 | 0 | 0 | -1 | -1 | -1 | -2 | -2 | -2 |
|   |     | 1 | 1 | 1 | 0  | 0  | 0  | -1 | -1 | -1 |
|   |     | 1 | 1 | 1 | 0  | 0  | 0  | -1 | -1 | -1 |
|   |     | 1 | 1 | 1 | 0  | 0  | 0  | -1 | -1 | -1 |
|   |     | 2 | 2 | 2 | 1  | 1  | 1  | 0  | 0  | 0  |
|   |     | 2 | 2 | 2 | 1  | 1  | 1  | 0  | 0  | 0  |
|   |     | 2 | 2 | 2 | 1  | 1  | 1  | 0  | 0  | 0  |
|   |     |   |   |   |    |    |    |    |    |    |

## \*= 2\*window\_size - 1

+= window\_size - 1

0

+= window\_size - 1

#### y-axis Matrix

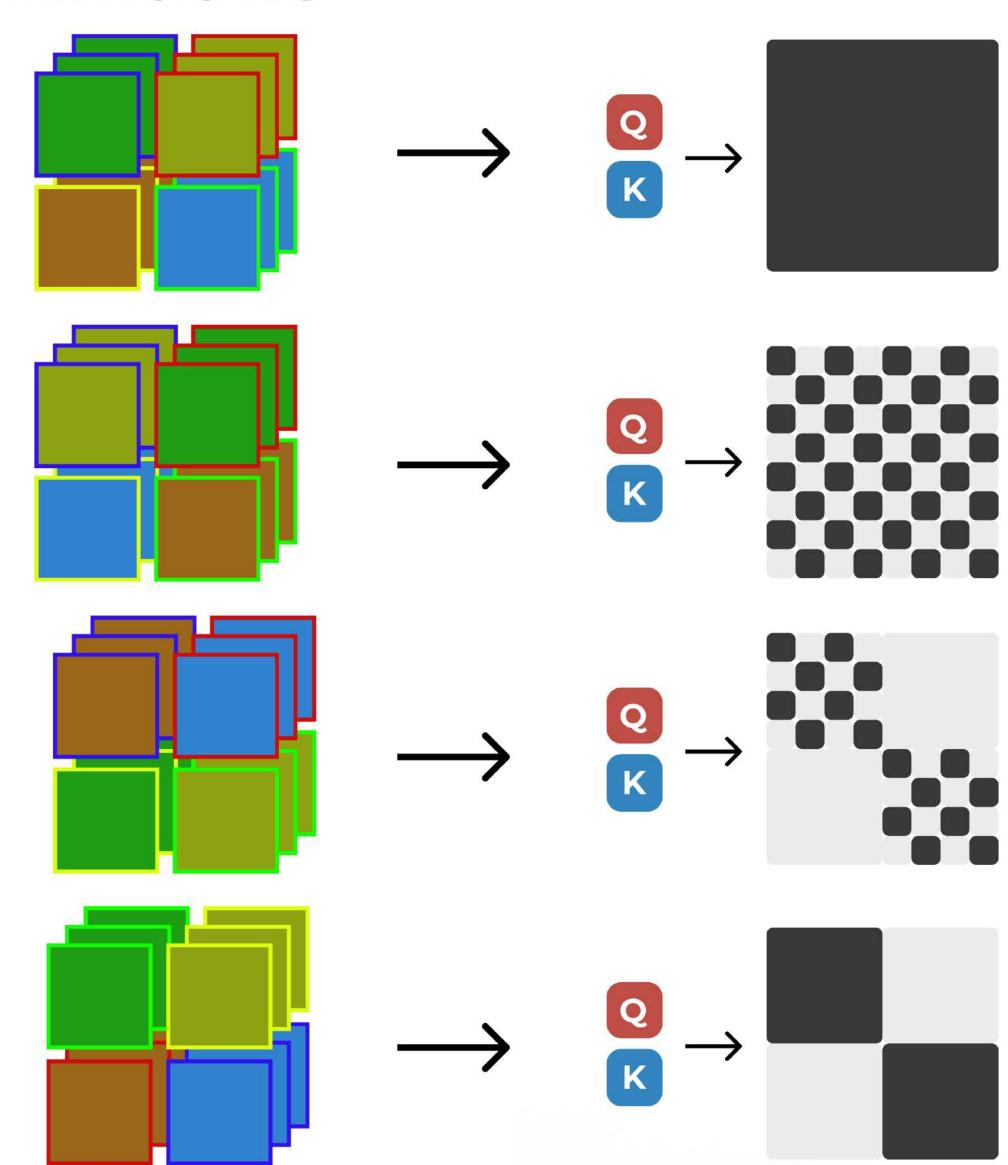
| _ | 6-2-10-3 |    |    |   |    |    |   |    |    |
|---|----------|----|----|---|----|----|---|----|----|
|   | 0        | -1 | -2 | 0 | -1 | -2 | 0 | -1 | -2 |
|   | 1        | 0  | -1 | 1 | 0  | -1 | 1 | 0  | -1 |
|   | 2        | 1  | 0  | 2 | 1  | 0  | 2 | 1  | 0  |
| _ | 0        | -1 | -2 | 0 | -1 | -2 | 0 | -1 | -2 |
|   | 1        | 0  | -1 | 1 | 0  | -1 | 1 | 0  | -1 |
|   | 2        | 1  | 0  | 2 | 1  | 0  | 2 | 1  | 0  |
|   | 0        | -1 | -2 | 0 | -1 | -2 | 0 | -1 | -2 |
|   | 1        | 0  | -1 | 1 | 0  | -1 | 1 | 0  | -1 |
|   | 2        | 1  | 0  | 2 | 1  | 0  | 2 | 1  | 0  |
|   | 3477     |    |    |   |    |    |   |    |    |

#### **Relative Position Bias**

| 12 | 11 | 10 | 7  | 6  | 5  | 2  | 1  | 0  |
|----|----|----|----|----|----|----|----|----|
| 13 | 12 | 11 | 8  | 7  | 6  | 3  | 2  | 1  |
| 14 | 13 | 12 | 9  | 8  | 7  | 4  | 3  | 2  |
| 17 | 16 | 15 | 12 | 11 | 10 | 7  | 6  | 5  |
| 18 | 17 | 16 | 13 | 12 | 11 | 8  | 7  | 6  |
| 19 | 18 | 17 | 14 | 13 | 12 | 9  | 8  | 7  |
| 22 | 21 | 20 | 17 | 16 | 15 | 12 | 11 | 10 |
| 23 | 22 | 21 | 18 | 17 | 16 | 13 | 12 | 11 |
| 24 | 23 | 22 | 19 | 18 | 17 | 14 | 13 | 12 |

### **SW-MSA**

# Shifted Windows



**Black: Self Attention** 

Whilt: Self Attention (x)

## Thank you

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