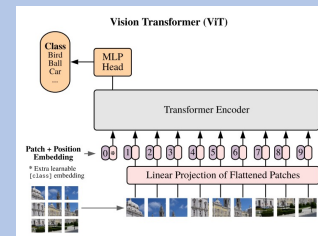


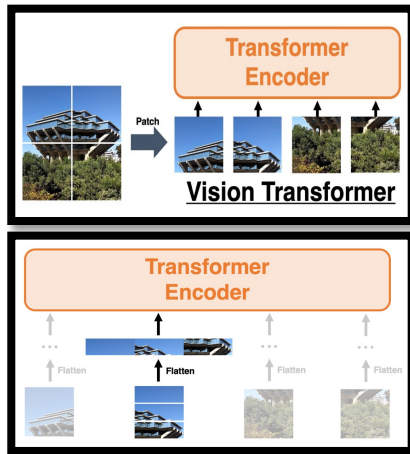
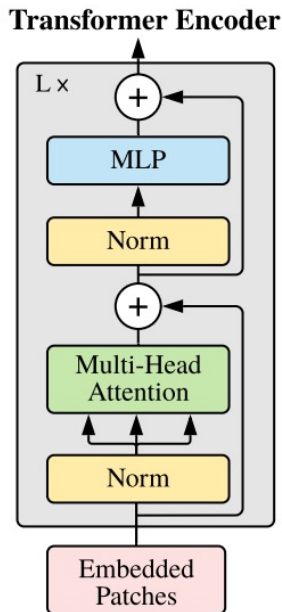
# Outline

## 1. Attention and Vision Transformers (ViT)

- NLP: Attention is all you need
- Transformer Encoder ViT with Self Attention for image classification



# Attention process in Vision



$$\mathbf{x} \in \mathbb{R}^{H \times W \times C}$$

$$\mathbf{x}_p \in \mathbb{R}^{N \times (P^2 \cdot C)}$$

$$N = HW/P^2$$

$$\mathbf{z}_0 = [\mathbf{x}_{\text{class}}; \mathbf{x}_p^1 \mathbf{E}; \mathbf{x}_p^2 \mathbf{E}; \dots; \mathbf{x}_p^N \mathbf{E}] + \mathbf{E}_{\text{pos}},$$

$$\mathbf{z}'_\ell = \text{MSA}(\text{LN}(\mathbf{z}_{\ell-1})) + \mathbf{z}_{\ell-1},$$

$$\mathbf{z}_\ell = \text{MLP}(\text{LN}(\mathbf{z}'_\ell)) + \mathbf{z}'_\ell,$$

$$\mathbf{y} = \text{LN}(\mathbf{z}_L^0)$$

$$\mathbf{E} \in \mathbb{R}^{(P^2 \cdot C) \times D}, \mathbf{E}_{\text{pos}} \in \mathbb{R}^{(N+1) \times D}$$

CLS token

(N+1)

D

D

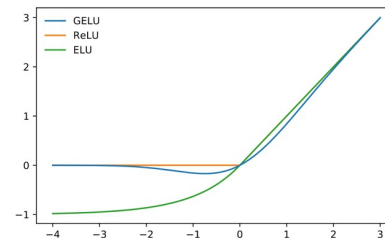
[class=CLS] token: a learnable embedding to the sequence of embedded patches

LayerNorm (LN) before every block, and residual connections after every block

MSA: Multi Head Self Attention

MLP: two layers with a GELU non-linearity

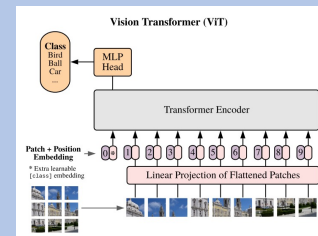
Hybrid Architecture : Raw image patches --> Feature map of a CNN



# Outline

## 1. Attention and Vision Transformers (ViT)

- NLP: Attention is all you need
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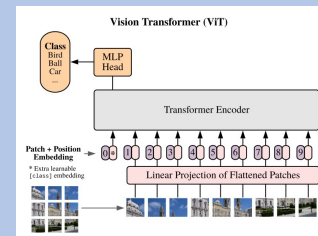


## 2. Transformer Decoder for downstream tasks

# Outline

## 1. Attention and Vision Transformers (ViT)

- NLP: Attention is all you need
- Transformer Encoder ViT with Self Attention for image classification

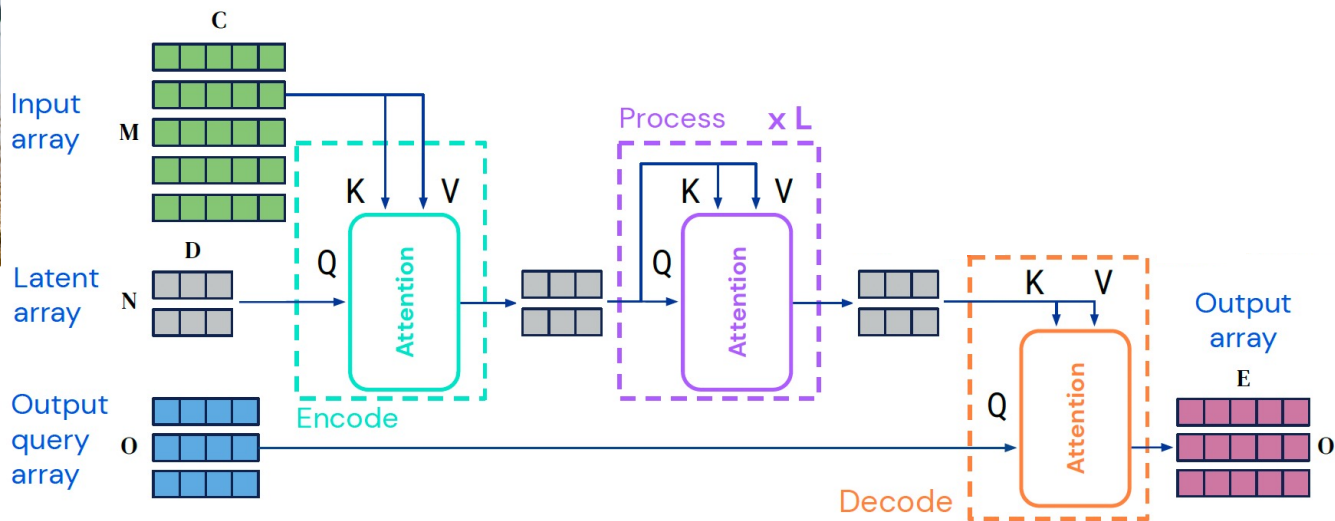


## 2. Transformer Decoder for downstream tasks

- Detection
- Segmentation
- Continual Learning, ...

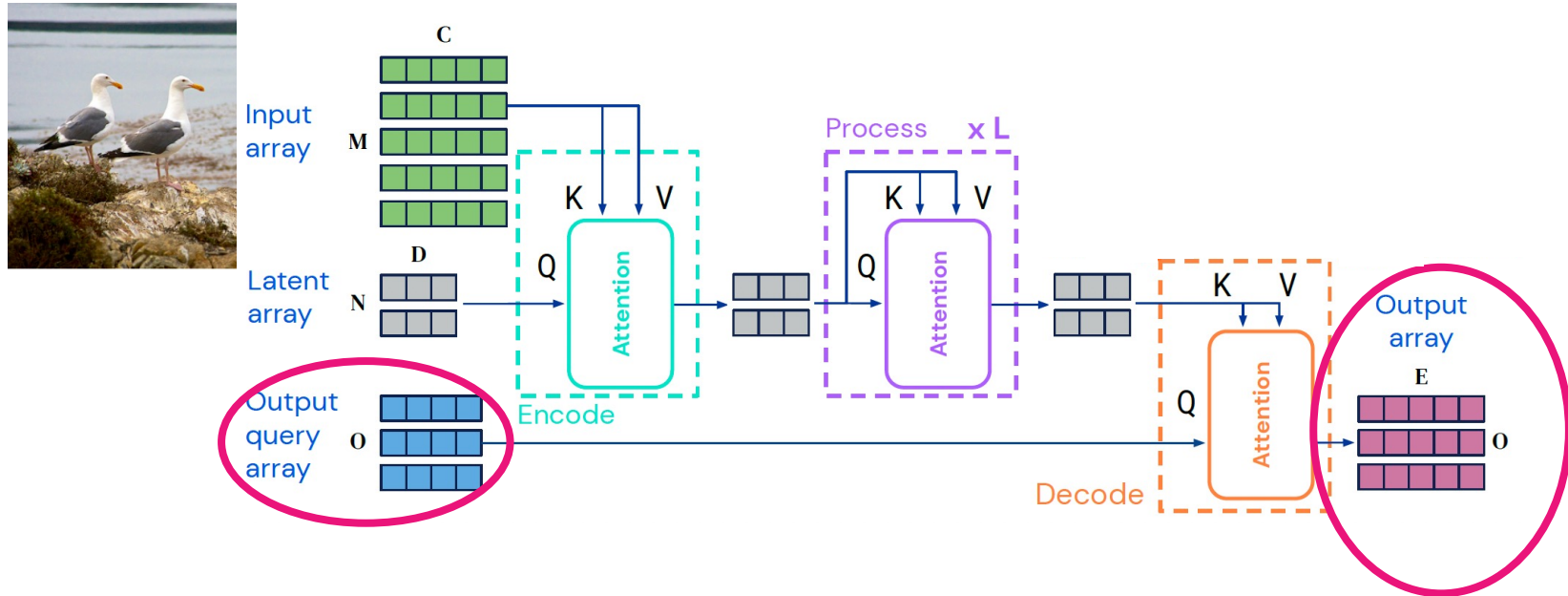
# General Decoder

[Perceiver IO A General Architecture for Structured Inputs & Outputs ICLR22]



# General Decoder

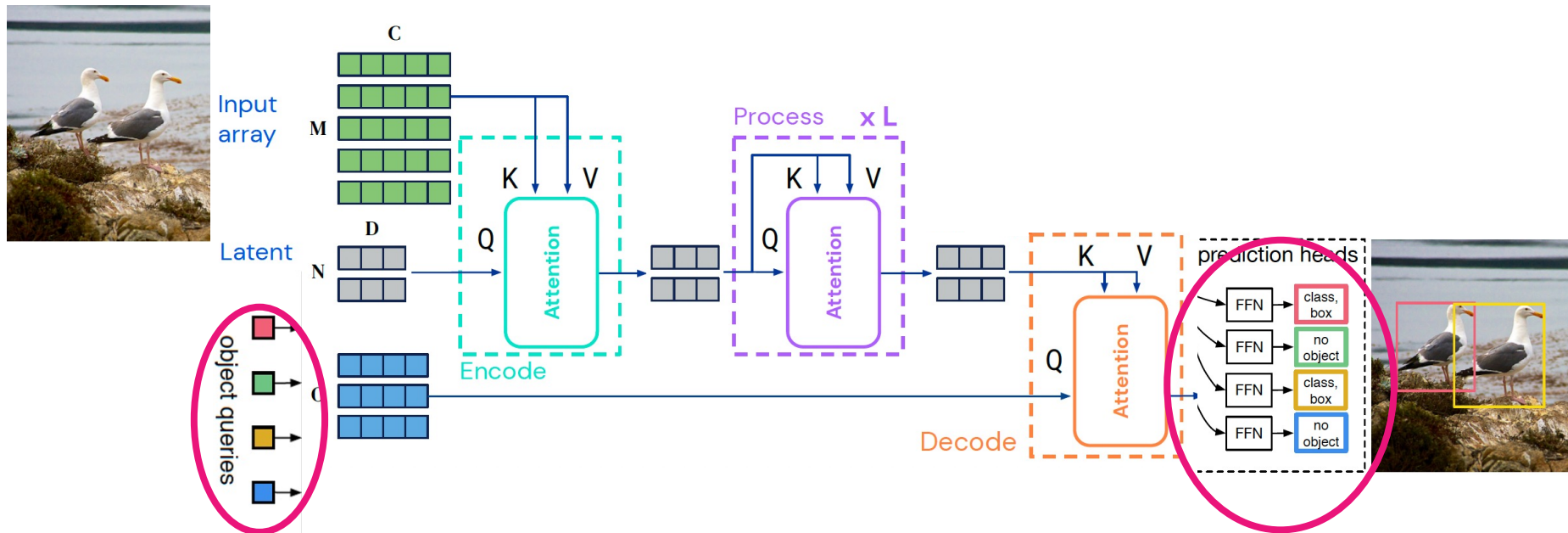
[Perceiver IO A General Architecture for Structured Inputs & Outputs ICLR22]



Output query array / Output array defines the downstream task: **detection, segmentation ...**

# General Decoder

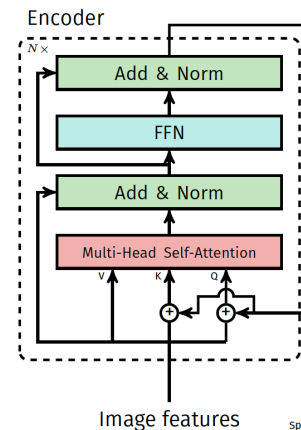
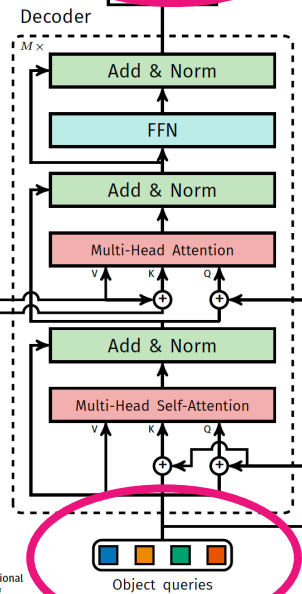
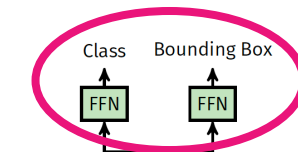
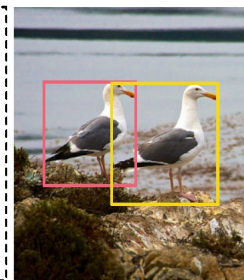
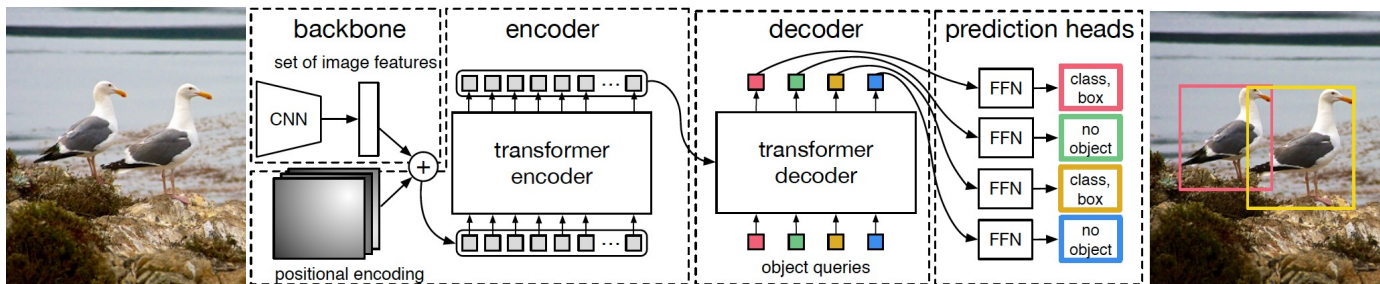
[Perceiver IO A General Architecture for Structured Inputs & Outputs ICLR22]



Output query array / Output array defines the downstream task: **detection**

# Transformer Decoder for detection

Just another scheme for DETR model



Cornell University

arXiv > cs > arXiv:2005.12872

Computer Science > Computer Vision and Pattern Recognition

[Submitted on 26 May 2020 (v1), last revised 28 May 2020 (this version, v3)]

## End-to-End Object Detection with Transformers

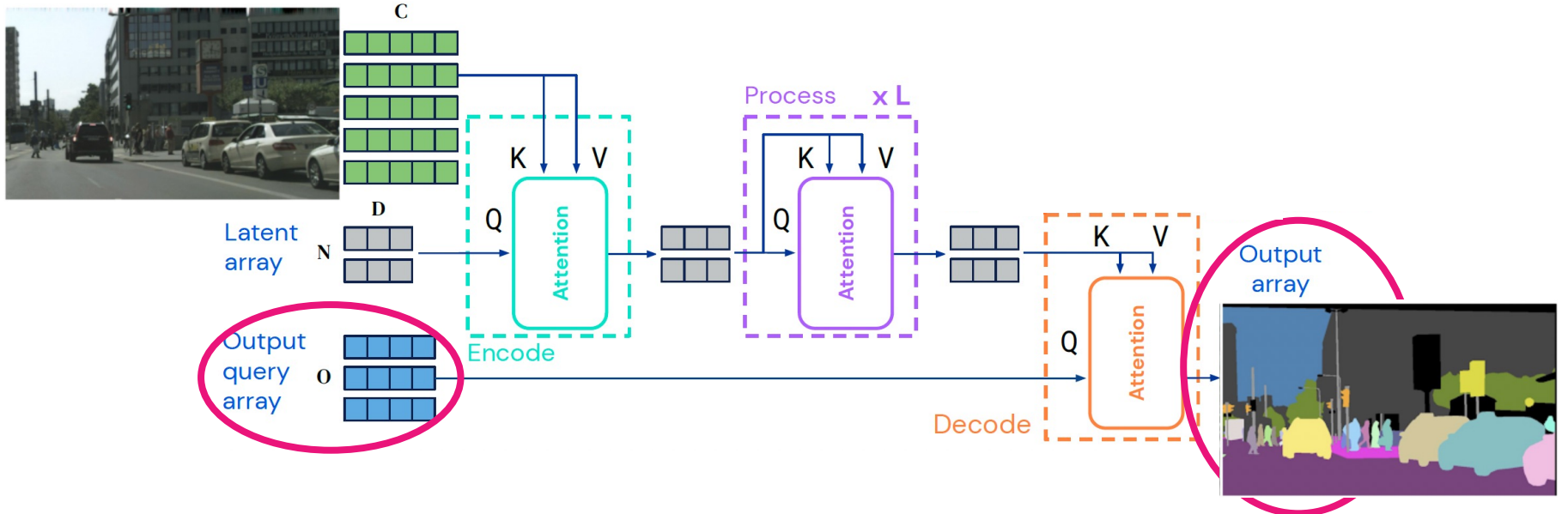
Nicolas Carion, Francisco Massa, Gabriel Synnaeve, Nicolas Usunier, Alexander Kirillov, Sergey Zagoruyko

We present a new method that views object detection as a direct set prediction problem. Our approach streamlines the detection pipeline by removing hand-designed components like a non-maximum suppression procedure or anchor generation that explicitly encode our prior knowledge. The new framework, called DETection TRansformer or DETR, is a set-based global loss that forces unique predictions via bipartite matching.



# General Decoder

[Perceiver IO A General Architecture for Structured Inputs & Outputs ICLR22]



Output query array / Output array defines the downstream task: **segmentation ...**

# General Decoder: or not!



This ICCV paper is the Open Access version, provided by the Computer Vision Foundation.  
Except for this watermark, it is identical to the accepted version;  
the final published version of the proceedings is available on IEEE Xplore.

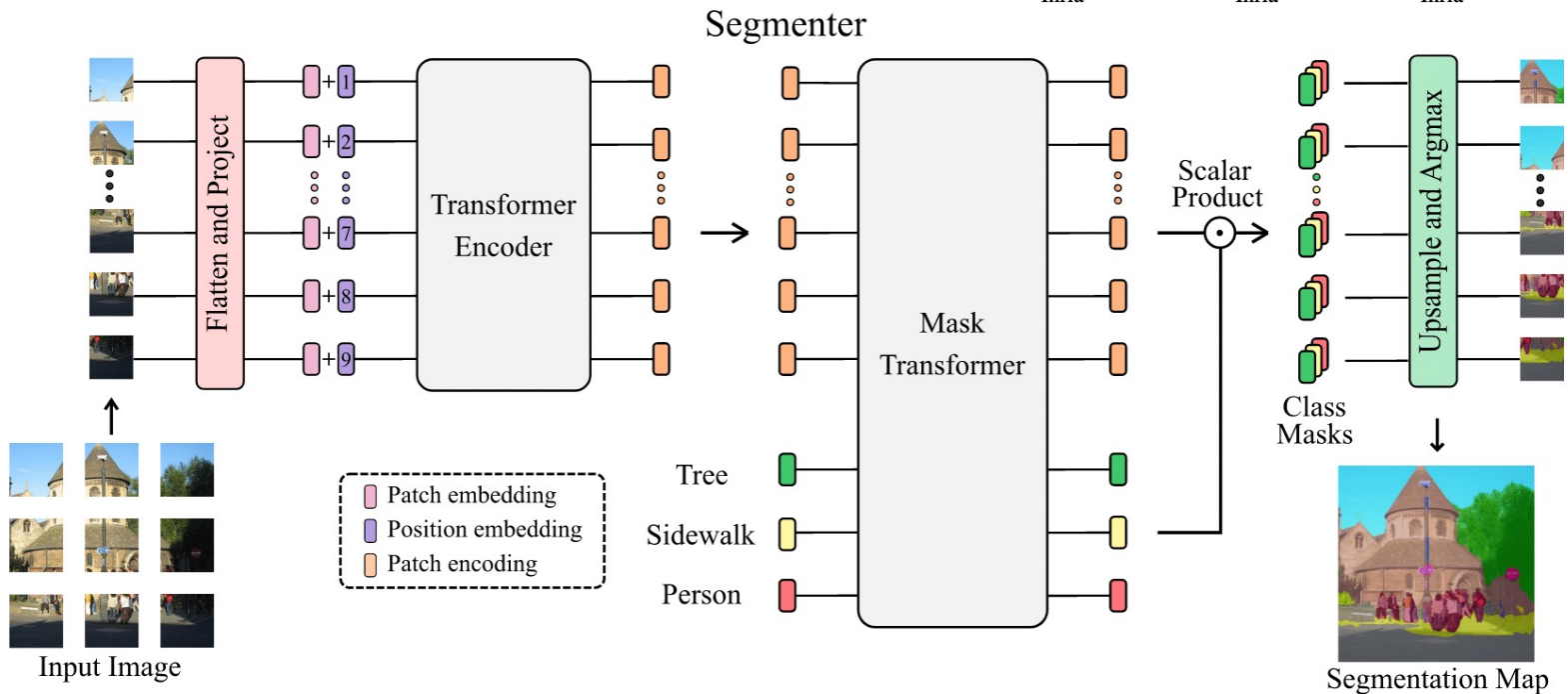
## Segmenter: Transformer for Semantic Segmentation

Robin Strudel\*  
Inria†

Ricardo Garcia\*  
Inria†

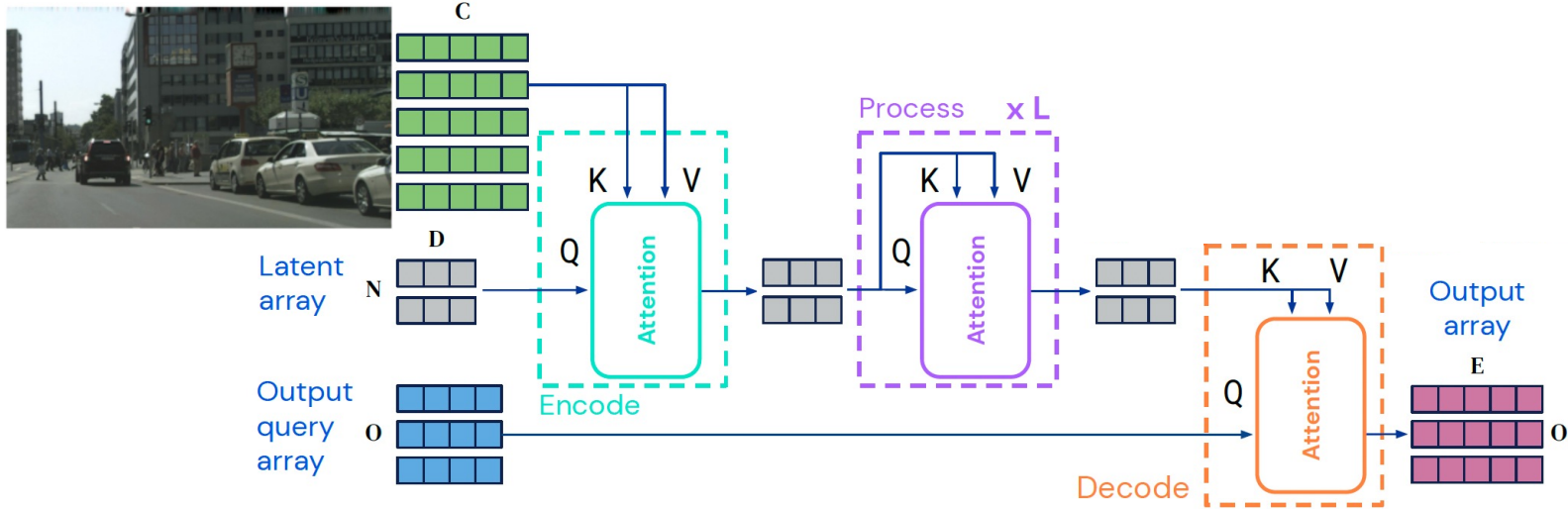
Ivan Laptev  
Inria†

Cordelia Schmid  
Inria†



# General Decoder

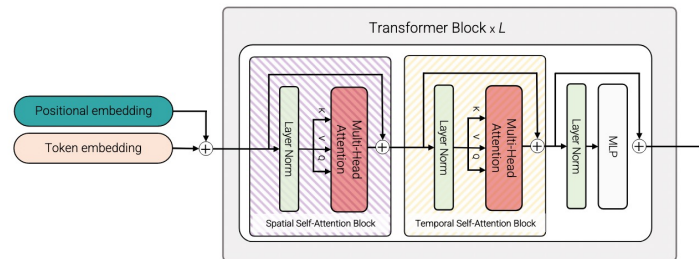
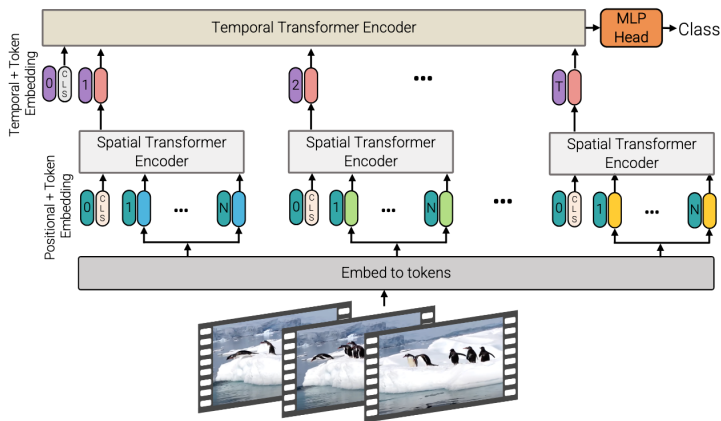
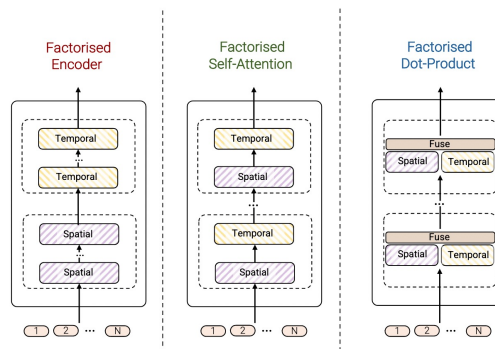
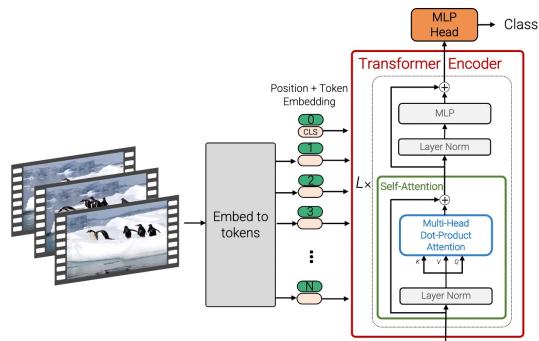
[Perceiver IO A General Architecture for Structured Inputs & Outputs ICLR22]



Output query array / Output array defines the downstream task: **continual learning**

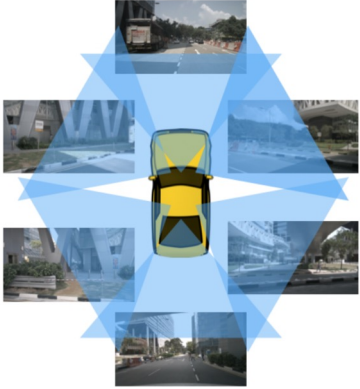
# Video Transformer

[ViViT: A Video Vision Transformer ICCV 2021]



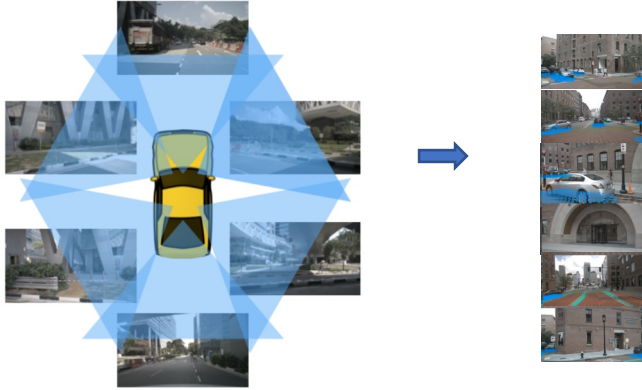
# General Encoder / Decoder

Input array = N cameras



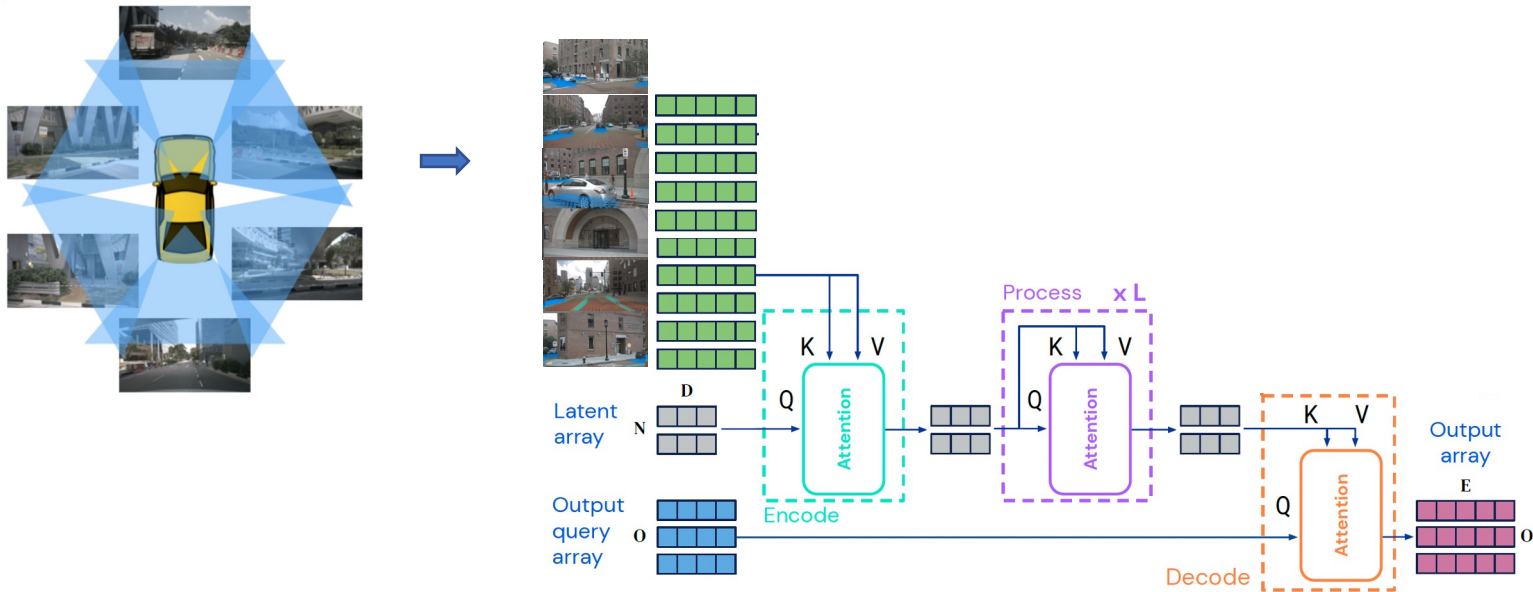
# General Encoder / Decoder

Input array = N cameras



# General Encoder / Decoder

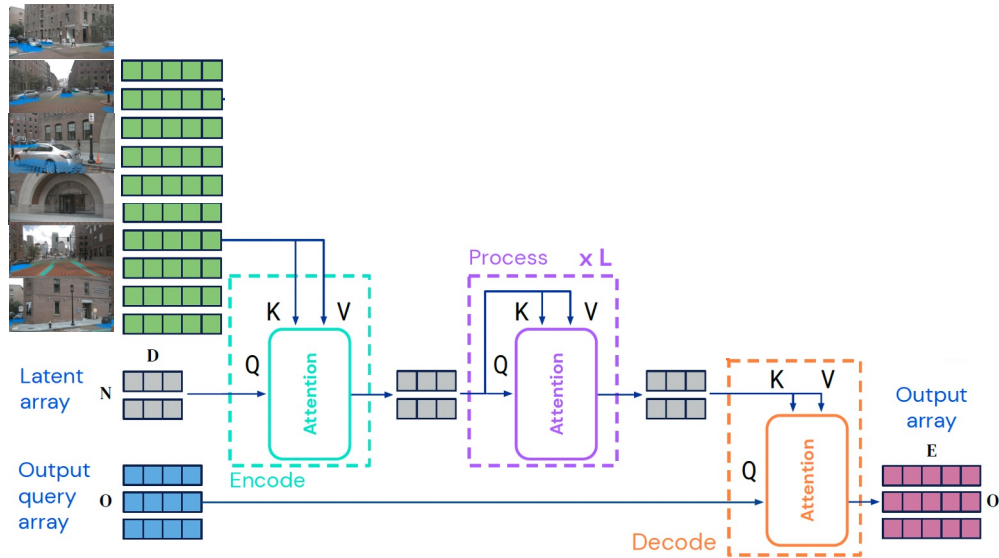
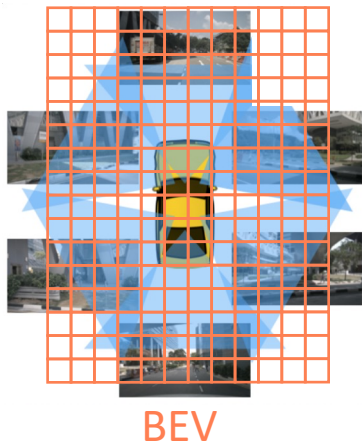
Input array = N cameras



# General Encoder / Decoder

Input array = N cameras

Output array = Bird Eye View (BEV) representation

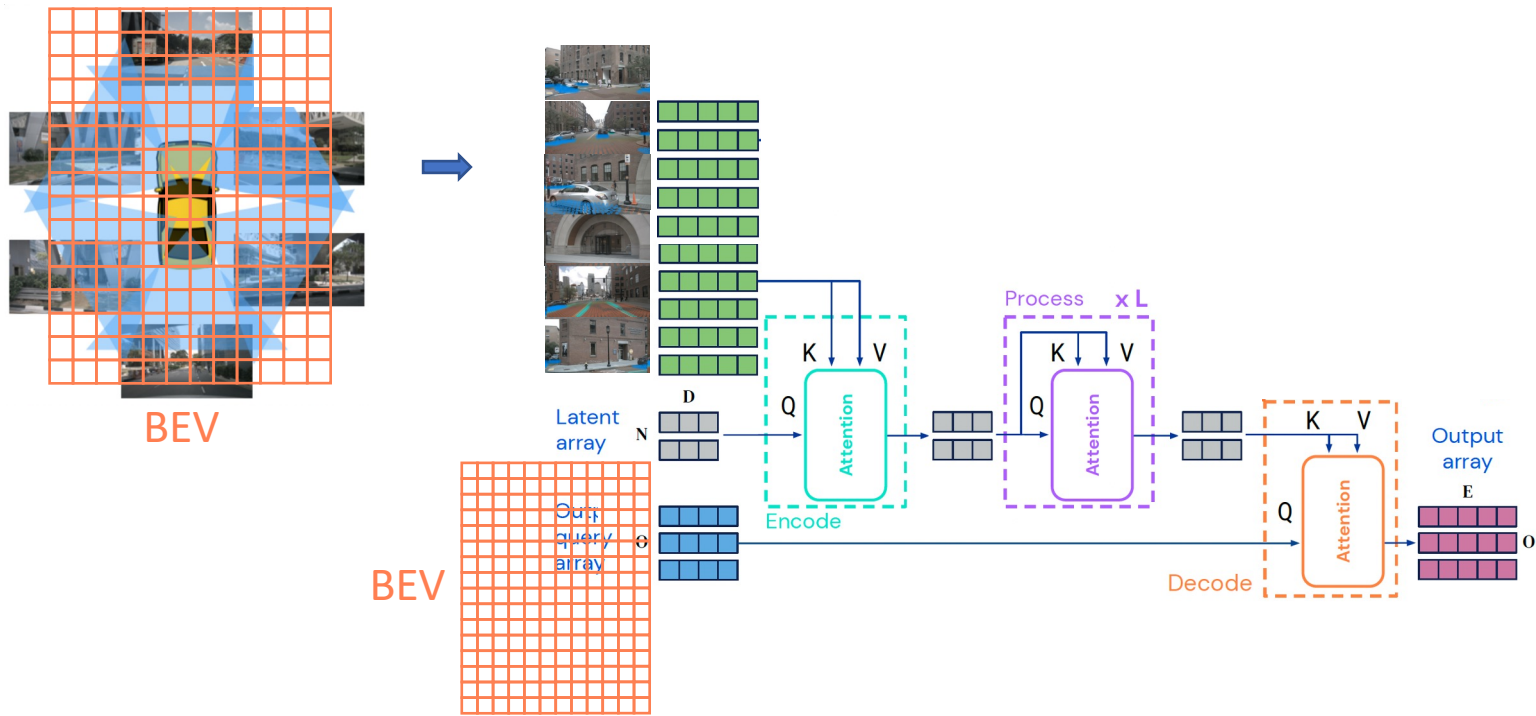




# General Encoder / Decoder

Input array = N cameras

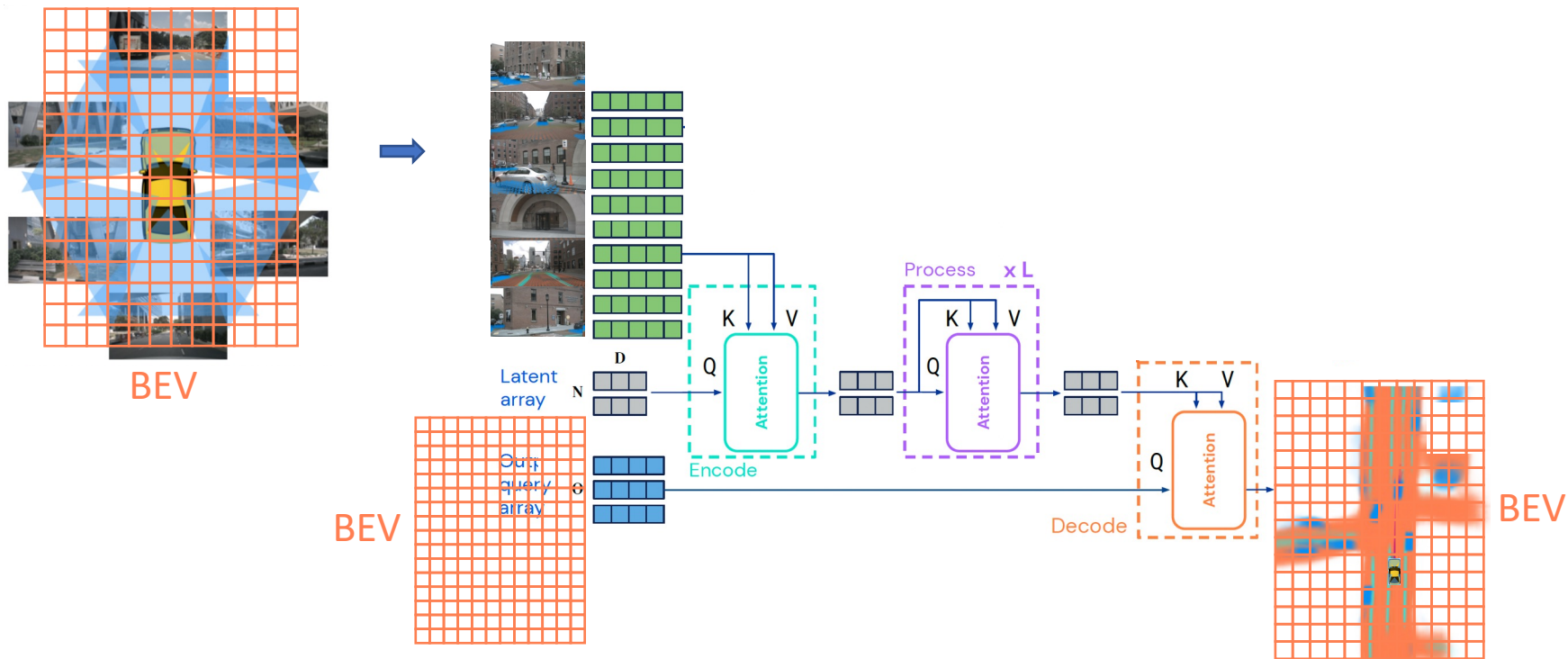
Output array = Bird Eye View (BEV) representation



# General Encoder / Decoder

Input array = N cameras

Output array = Bird Eye View (BEV) representation



# Vision Transformers

**Global Attention** mechanism at every layer of the deep archi

Very **competitive architectures** in image classification with the best Convnets

**Fusion/Merging by mixing** thanks to cross attention process

**Somehow universal** deep structure around encoding/decoding for many vision tasks as classification (1 class token), object detection, segmentation, ...

