

Calculate the FLOPs for V2X models

General Formula for FLOPs in a Convolutional Layer

For a convolutional layer, the number of FLOPs can be calculated as:

$$\text{FLOPs} = 2 \times K^2 \times C_{in} \times C_{out} \times H_{out} \times W_{out}$$

Where:

- K is the kernel size
- C_{in} and C_{out} are the number of input and output channels, respectively
- H_{out} and W_{out} are the height and width of the output feature map

Note: You can find the code for FLOPs Calculation is on GitHub

V2X-ViT

Table T2: Detailed architectural specifications for V2X-ViT.

	Output size	V2X-ViT framework
PointPillar Encoder	$M \times 352 \times 96 \times 256$	$\begin{bmatrix} \text{Voxel samp. reso. 0.4m, Scatter, 64} \\ \text{Conv3x3, 64, stride 2, BN, ReLU} \end{bmatrix} \times 3$ $\begin{bmatrix} \text{Conv3x3, 128, stride 2, BN, ReLU} \end{bmatrix} \times 5$ $\begin{bmatrix} \text{Conv3x3, 256, stride 2, BN, ReLU} \end{bmatrix} \times 8$ $\begin{bmatrix} \text{ConvT3x3, 128, stride 1, BN, ReLU} \end{bmatrix} \times 1$ $\begin{bmatrix} \text{ConvT3x3, 128, stride 2, BN, ReLU} \end{bmatrix} \times 1$ $\begin{bmatrix} \text{ConvT3x3, 128, stride 4, BN, ReLU} \end{bmatrix} \times 1$
	$M \times 176 \times 48 \times 256$	$\begin{bmatrix} \text{Concat3, 384} \\ \text{Conv3x3, 256, stride 2, ReLU} \\ \text{Conv3x3, 256, stride 1, ReLU} \end{bmatrix} \times 1$
Delay-aware Pos. Encoding	$M \times 176 \times 48 \times 256$	$\begin{bmatrix} \text{sin-cos pos. encoding} \\ \text{Linear, 256} \end{bmatrix} \times 1$
Transformer Backbone	$M \times 176 \times 48 \times 256$	$\begin{bmatrix} \text{HSMA, dim 256, head 8} \\ \text{MSwin, dim 256, head \{16, 8, 4\},} \\ \text{ws. \{4 \times 4, 8 \times 8, 16 \times 16\}} \\ \text{MLP, dim 256} \end{bmatrix} \times 3$
Detection Head	$176 \times 48 \times 16$	Cls. head: $\begin{bmatrix} \text{Conv1x1, 2, stride 1} \end{bmatrix}$ Regr. head: $\begin{bmatrix} \text{Conv1x1, 14, stride 1} \end{bmatrix}$

V2X-ViT has 4 modules, we will need to calculate for each module and sum them up for total FLOPs

PointPillar Encoder:

- $\begin{bmatrix} \text{Conv3x3, 64, stride 2, BN, ReLU} \end{bmatrix} \times 3$:
 - Kernel size (K) = 3x3
 - Input channels (C_{in}) : 256
 - Output channels (C_{out}) = 64
 - Stride = 2

- Output feature map size (using stride 2 and output size)
 - $H_{out} = 253/2 = 176$
 - $W_{out} = 96/2 = 48$
- => FLOPs per layer = $2 \times 9 \times 256 \times 64 \times 176 \times 48$
- => Total FLOPs for 3 repetitions = $3 \times (\text{FLOPs per layer})$
- => With M agents, total FLOPs = $M \times 3 \times (\text{FLOPs per layer})$

You can apply the same method to calculate FLOPs of the rest of PointPillar Encoder layers, note that the output from previous layer is the input for current layer.

Delay-aware Pos. Encoding:

Negligible

Transformer Backbone:

- HSMA: Each attention head operation is assumed to have the following FLOPs:
 - Query, Key, Value Computation: $N \times (256/8)^2 \times 2$
 - Attention Score Calculation: $N \times N \times (256/8)$
 - Output Calculation: $N \times (256/8) \times (256/8)$
 - N is the number of tokens ($176 \times 48 = 8448$ in this case)
 - Given that there are 8 heads, we multiply the FLOPs for one head by 8
- MSwin: The calculation for each scale of MSwin is similar to HSMA, with adjustments made for window sizes. Assume the FLOPs are the same as one HSMA head for each scale
- MLP: we calculate FLOPs as $2 \times 256 \times 256 \times 176 \times 48$

Total FLOPs for Transformer Backbone is the sum of HSMA, MSwin, and MLP, then multiply by 3 to account for 3 repetitions in the transformer, then multiply by the number of agents M

Detection Head:

Negligible

My hand-written FLOPs calculation for V2X-ViT:

Assume we have 2 agents

$$\begin{aligned}
& \text{FLOPs : } M \times 2 \times 9 \times 256 \times 64 \times 176 \times 48 \times 3 (\sim 7.5 \text{ B}) \\
& + M \times 2 \times 9 \times 64 \times 128 \times 176 \times 48 \times 5 (\sim 6.2 \text{ B}) \\
& + M \times 2 \times 9 \times 128 \times 256 \times 176 \times 48 \times 8 (\sim 39.9 \text{ B}) \\
& + M \times 2 \times 9 \times 256 \times 128 \times 352 \times 96 (\sim 19.9 \text{ B}) \\
& + M \times 2 \times 9 \times 128 \times 128 \times 176 \times 48 (\sim 2.5 \text{ B}) \\
& + M \times 2 \times 9 \times 128 \times 128 \times 88 \times 24 (\sim 0.6 \text{ B}) \\
& + M \times 2 \times 9 \times 256 \times 256 \times 88 \times 24 (\sim 2.5 \text{ B}) \\
& + M \times 2 \times 9 \times 256 \times 256 \times 176 \times 48 (\sim 10.0 \text{ B}) \\
& + M \times 2 \times 1 \times 256 \times 256 \times 176 \times 48 (\sim 1.1 \text{ B}) \\
& + 3 \times \left\{ \begin{aligned} & 8 \times \left[(8448 \times (256/8)^2 \times 2) + (8448 \times 8448 \times 256/8) + (8448 \times (256/8)^2) \right] \\ & + 16 \times \left[(8448 \times (256/8)^2 \times 2) + (8448 \times 8448 \times 256/8) + (8448 \times (256/8)^2) \right] \\ & + 8 \times \left[(8448 \times (256/8)^2 \times 2) + (8448 \times 8448 \times 256/8) + (8448 \times (256/8)^2) \right] \\ & + 4 \times \left[(8448 \times (256/8)^2 \times 2) + (8448 \times 8448 \times 256/8) + (8448 \times (256/8)^2) \right] \\ & + 2 \times 256 \times 256 \times 176 \times 48 \end{aligned} \right\} (\sim 253.2 \text{ B}) \times M \\
& + 2 \times 1 \times 16 \times 2 \times 48 \times 176 \left(\text{small} \right) \\
& + 2 \times 1 \times 2 \times 14 \times 176 \times 48 \left(\text{small} \right)
\end{aligned}$$

Point Pillar Encoder

Transformer Backbone

Detection Head

$M=2 \Rightarrow \text{Total FLOPs} = 686.8 \text{ B}$

CoBEVT

Table A2: Detailed architectural specifications of CoBEVT for OPV2V camera track. M represents the number of cameras and N is the number of agents.

	Output size	CoBEVT framework
ResNet34 Encoder	$N \times M \times 64 \times 64 \times 128$	ResNet34-layer1
	$N \times M \times 32 \times 32 \times 256$	ResNet34-layer2
	$N \times M \times 16 \times 16 \times 512$	ResNet34-layer3
SinBEVT Backbone	$N \times 128 \times 128 \times 128$	FAX-CA, dim 128, head 4, bev win. sz. {16 × 16} feat win. sz. {8 × 8} MLP, dim 256 Res-Bottleneck-block × 2 × 1
	$N \times 64 \times 64 \times 128$	FAX-CA, dim 128, head 4, bev win. sz. {16 × 16} feat win. sz. {8 × 8} MLP, dim 256 Res-Bottleneck-block × 2 × 1
	$N \times 32 \times 32 \times 128$	FAX-CA, dim 128, head 4, bev win. sz. {32 × 32} feat win. sz. {16 × 16} MLP, dim 256 Res-Bottleneck-block × 2 × 1
FuseBEVT Backbone	$N \times 32 \times 32 \times 128$	FAX-SA, dim 128, head 4, win. sz. {8 × 8} MLP, dim 256 × 3
Decoder	$64 \times 64 \times 128$	Bilinear-upsample, Conv3x3, BN
	$128 \times 128 \times 64$	Bilinear-upsample, Conv3x3, BN
	$256 \times 256 \times 32$	Bilinear-upsample, Conv3x3, BN
	$256 \times 256 \times k$	Dyna. Obj. head: [Conv1x1, 2, stride 1] Stat. Obj. head: [Conv1x1, 3, stride 1]

CoBEVT has 4 modules, we will calculate the FLOPs for each and add them up

ResNet34 Encoder:

Base on my research, ResNet34 has 3.6 GFLOPs for input image of 224x224

CoBEVT input is 512x512 => ResNet34 GFLOPs are 4 times of 3.6

With 4 camera, and 3 ResNet34 layers => Total FLOPs = $3.6 \times 4 \times 4 \times 3$

SinBEVT Backbone:

- FAX-CA:
 - $\text{FLOPs} = N \times H \times W \times \text{Number of heads} \times \text{Output channels} \times (\text{bev win} + \text{fear win})$
- MLP:
 - $\text{FLOPs} = N \times H \times W \times \text{dimension} \times \text{Output channels} \times 2$
- Res-Bottleneck:
 - $\text{FLOPs} = N \times H \times W \times \text{Kernel size} \times \text{Input channels} \times \text{Output channels}$

FuseBEVT Backbone:

Same as SinBEVT backbone

Decoder:

Follow the general formula for FLOPs calculation

[My hand-written FLOPs calculation for CoBEVT:](#)

Assume we have 2 agents, each agent has 2 cameras

