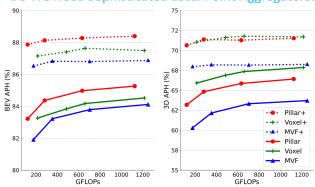
QCRAFT

PillarNeXt: Rethinking Network Designs for 3D Object Detection in LiDAR Point Clouds



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Do We Need Sophisticated Local Point Aggregators?

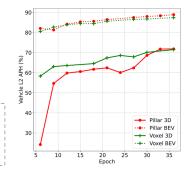


Overview of pillar, voxel and multi-view fusion (MVF) based 3D object detection networks under different GFLOPs. The dash lines denote the enhanced versions of corresponding models (+). We report the L2 BEV and 3D APH of vehicle on the validation set of Waymo Open Dataset (WOD).

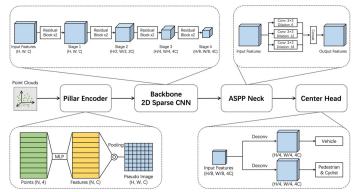
Training Matters

Learning behaviors of the pillar based and the voxel based models. We report the L2 3D and BEV APH of vehicle on the validation set on WOD.





Network Architecture



A schematic overview of the network architecture of the proposed PillarNeXt.

Study of Neck Modules

Method	Vehic	cle L1	Vehic	cle L2	Pedest	rian L1	Pedestrian L2	
Method	AP	APH	AP	APH	AP	APH	AP	APH
Neck of PillarNet [31] 91.39	90.58	84.54	83.72	87.90	83.02	81.93	77.20
FPN [17]	92.17	91.35	85.96	85.13	87.88	82.91	82.05	77.23
BiFPN [39]	92.71	91.90	86.92	86.09	87.86	82.88	82.05	77.23
Plain	91.01	90.19	83.86	83.04	87.59	82.61	81.52	76.71
Dilated Block [7]	92.70	91.90	86.61	85.79	87.84	82.91	82.09	77.29
ASPP [5]	92.77	91.94	86.99	86.14	87.74	82.85	82.00	77.26

Comparison of different neck modules integrated in our networks. Groups 1 and 2 correspond to the multi-scale and sing-scale necks under the BEV metrics.

Study of Resolutions

Comparison of different	In Size	Backbone ↓	Head ↑	Out Size	Veh	Ped	Latency
resolutions by pillar size	0.3	1	1	0.3	65.0	67.2	255
(m) of input grids and	0.075	8	1	0.6	62.8	66.6	131
output features to head.	0.075	8	2	0.3	64.8	69.0	173

Boosting Roadmap



■Pedestrian L2 3D APH ■Vehicle L2 3D APH ■Pedestrian L2 BEV APH ■Vehi

Experimental Results

Method	Frames	Vehicle L1		Vehicle L2		Pedestrian L1		Pedestrian L2		Cyclist L1		Cyclist L2	
Method	rrames	AP	APH	AP	APH	AP	APH	AP	APH	AP	APH	AP	APH
PillarNet-18 [31]	2	79.59	79.06	71.56	71.08	82.11	78.82	74.49	71.35	70.41	69.57	68.27	67.46
PillarNet-34 [31]	2	79.98	79.47	72.00	71.53	82.52	79.33	75.00	71.95	70.51	69.69	68.38	67.58
PV-RCNN++* [32]	2	80.17	79.70	72.14	71.70	83.48	80.42	75.54	72.61	74.63	73.75	72.35	71.50
RSN* [37]	3	78.4	78.1	69.5	69.1	79.4	76.2	69.9	67.0	-	-	-	-
SST-TS* [11]	3	78.66	78.21	69.98	69.57	83.81	80.14	75.94	72.37				
SWFormer [36]	3	79.4	78.9	71.1	70.6	82.9	79.0	74.8	71.1	-	-	-	-
PillarNeXt-B	3	80.58	80.08	72.89	72.42	85.04	82.11	78.04	75.19	78.92	77.94	76.71	75.74
CenterFormer [49]	8	78.8	78.3	74.3	73.8	82.1	79.3	77.8	75.0	75.2	74.4	73.2	72.3
MPPNet [8]	16	82.74	82.28	75.41	74.96	84.69	82.25	77.43	75.06	77.28	76.66	75.13	74.52
3DAL [†] [29]	ALL	84.50	-	-	-	82.88	-	-	-	-	-	-	-
	-	Vehic	ele L1	Vehic	le L2	Pedest	rian L1	Pedest	rian L2	Cycl	ist L1	Cycli	ist L2
Method	Frames	AP	APH	AP	APH	AP	APH	AP	APH	AP	APH	AP	APH
PV-RCNN++* [32]	1	91.57	-	-	-	85.43	-	-	-	75.94	-	-	-
PillarNeXt-B	1	93.30	92.60	87.26	86.53	88.19	82.13	81.77	75.82	75.67	74.61	72.97	71.95
SWFormer [36]	3	92.60	-	-	-	87.50	-	-		-	-	-	-
PillarNeXt-B	3	94.41	93.73	89.36	88.66	90.20	86.94	84.66	81.36	81.35	80.32	79.23	78.22
3DAL [†] [29]	ALL	93.30	-	-	-	86.32	-	-	-	-	-	-	-

Comparisons under the 3D (top) and BEV (bottom) metrics on the validation set of WOD.

Method	Encoder	Grid Size	NDS	mAP	mATE↓	mASE↓	mAOE↓	mAVE↓	mAAE↓
CenterPoint [45]	V	0.075	66.8	59.6	0.292	0.255	0.302	0.259	0.193
OHS [6]	V	0.1	66.0	59.5	-	-	-	-	-
PillarNet-18 [31]	P	0.075	67.4	59.9	-	-		-	-
Transfusion-L [1]	V	0.075	66.8	60.0	-	-	-	-	-
UVTR-L [15]	V	0.075	67.7	60.9	0.334	0.257	0.300	0.204	0.182
VISTA [9]	V+R	0.1	68.1	60.8	-	-	-	-	-
PillarNeXt-B	P	0.075	68.8	62.5	0.278	0.251	0.269	0.248	0.201
Our Voxel-B	V	0.075	68.2	62.4	0.278	0.250	0.308	0.263	0.198

Comparisons on the validation set of nuScenes.