

# OpenPCDet: An Open-source Toolbox for 3D Object Detection from Point Clouds

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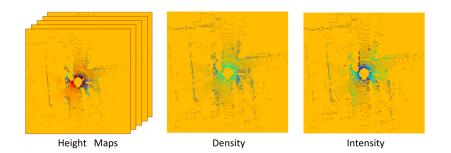
Nov. 30, 2020

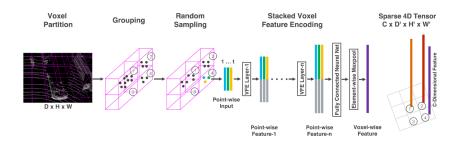
# Outline

- Introduction to 3D object detection from point clouds
  - Feature learning from point clouds (voxel-based vs. point-based)
  - 3D box generation (anchor-based vs. anchor-free)
  - Two-stage detector with 3D box refinement
- OpenPCDet architecture and experiments

### Feature learning from point clouds

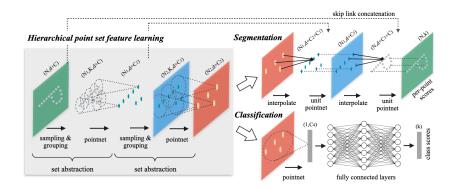
- 2D CNN with bird's view representation
- 3D Voxel CNN (or sparse conv) with Voxel Feature Encoding (VFE)
- PointNet++ series
- Combination of the above strategies





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#### Reference

1. Xiaozhi Chen, et al., Multi-View 3D Object Detection Network for Autonomous Driving

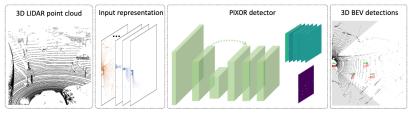
2. Yin Zhou, et al., VoxelNet: End-to-End Learning for Point Cloud Based 3D Object Detection

3. Charles Qi, et al., PointNet++: Deep Hierarchical Feature Learning onPoint Sets in a Metric Space

#### Anchor-based 3D box generation



• Generate 3D boxes by selecting and tuning pre-defined 3D anchors on the bird-view 2D feature maps

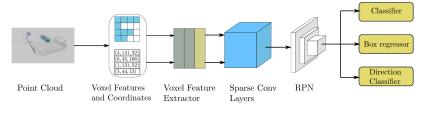


**Region Proposal Network** 

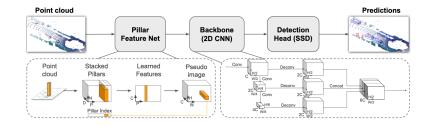
**Convolutional Middle Layers** 

Feature Learning Network

PIXOR



SECOND



PointPillar

#### Reference

1. Yan Yan, et al., SECOND: Sparsely EmbeddedConvolutional Detection

2. Bin Yang, et al., PIXOR: Real-time 3D Object Detection from Point Clouds

3. Alex H. Lang, et al., PointPillars: Fast Encoders for Object Detection from Point Clouds

VoxelNet

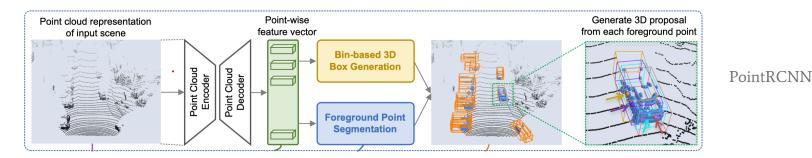
4. Yin Zhou, et al., VoxelNet: End-to-End Learning for Point Cloud Based 3D Object Detection

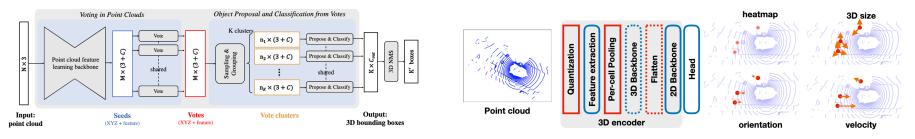
#### Anchor-free 3D box generation



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• Generating 3D box without predefined dense anchors





VoteNet

CenterPoint

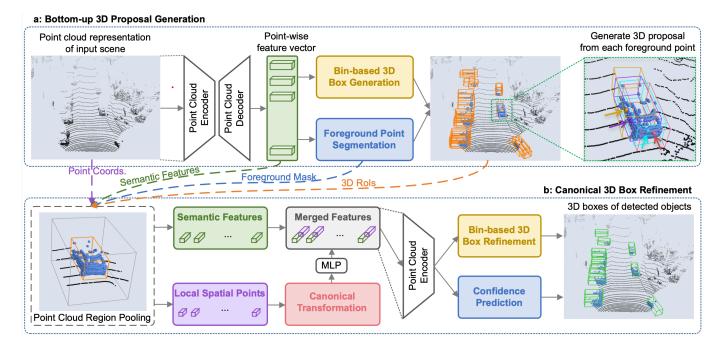
#### Reference

- 1. Shaoshuai Shi, et al., PointRCNN: 3D Object Proposal Generation and Detection from Point Cloud
- 2. Charles Qi, et al., Deep Hough Voting for 3D Object Detection in Point Clouds
- 3. Tianwei Yin, et al., Center-based 3D Object Detection and Tracking

### Two-stage detector with 3D box refinement



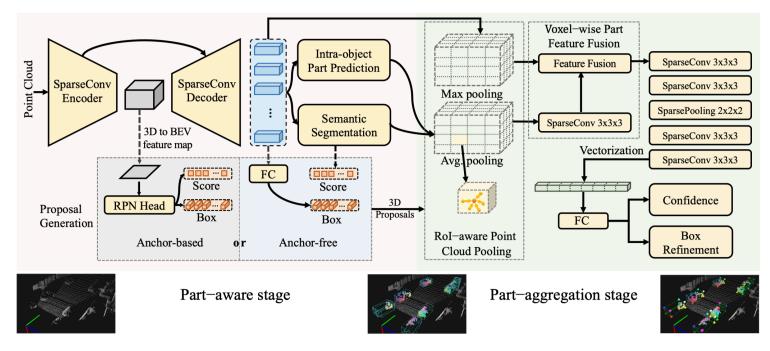
• PointRCNN: 3D box refinement with point cloud region pooling



## Two-stage detector with 3D box refinement



• Part-A2-Net: 3D box refinement with RoI-aware point cloud pooling



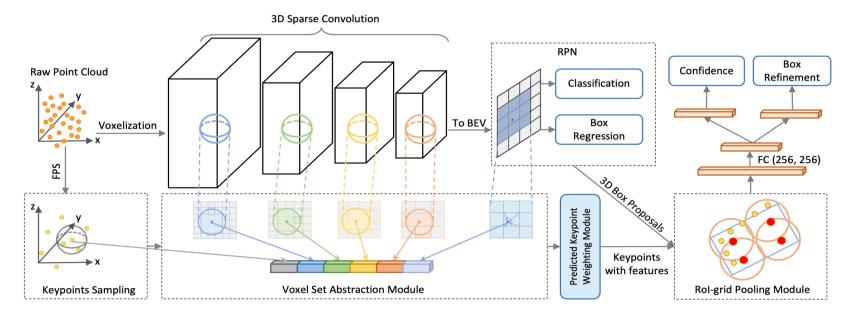
#### Reference

1. Shaoshuai Shi, et al., From Points to Parts: 3D Object Detection from Point Cloud with Part-aware and Part-aggregation Network

## Two-stage detector with 3D box refinement

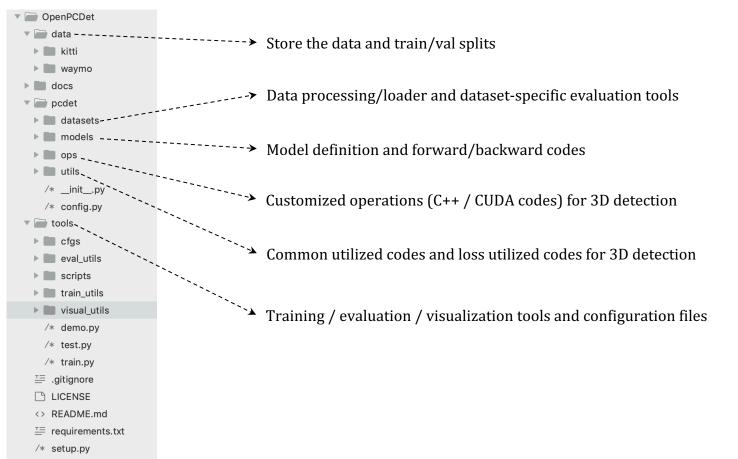


• PV-RCNN: 3D box refinement with RoI-grid pooling



#### **OpenPCDet: 3D Detection Toolbox**

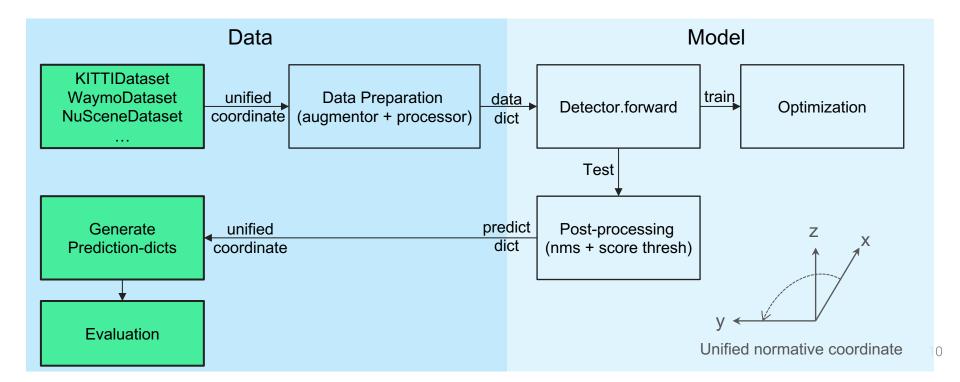




#### **OpenPCDet:** Architecture



• Data-Model separation with unified coordinate and box definition across different datasets

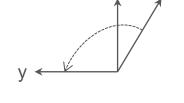


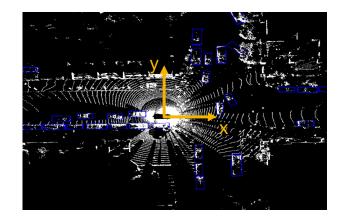
#### Unified 3D box definition

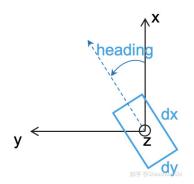
#### 11

## OpenPCDet: Unified coordinate and 3D box definition

- Unified coordinate:
  - x-axis points towards the front
  - y-axis points towards the left
  - z-axis points towards the up direction
- Unified 3D box definition:
  - (cx, cy, cz, dx, dy, dz, heading)
  - Heading angle: the angle from x-axis to yaxis (heading = 0 for the direction of x-axis)



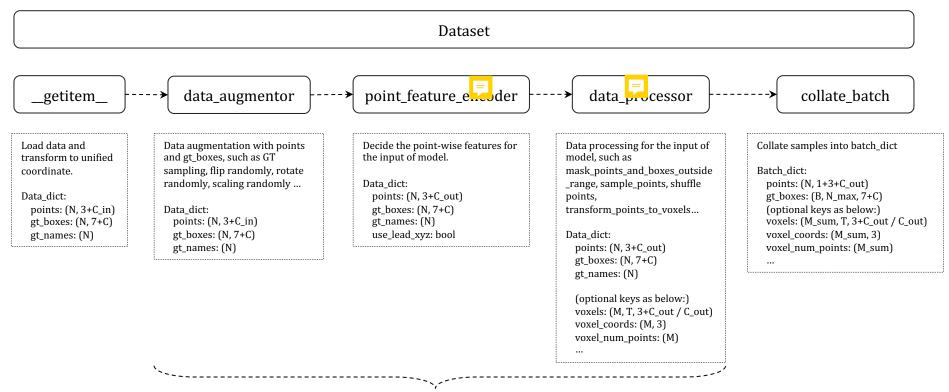






#### **OpenPCDet:** Data flow

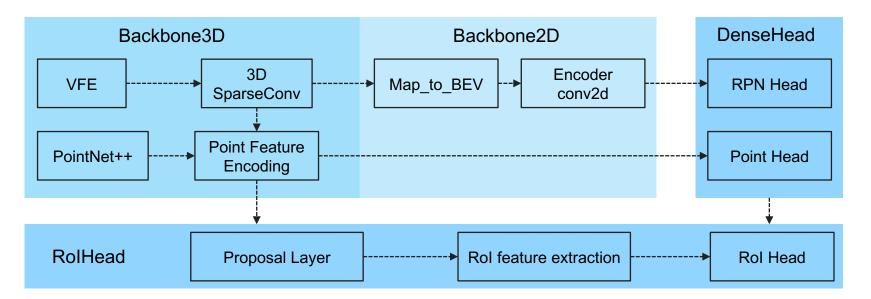




#### **OpenPCDet:** Modularization

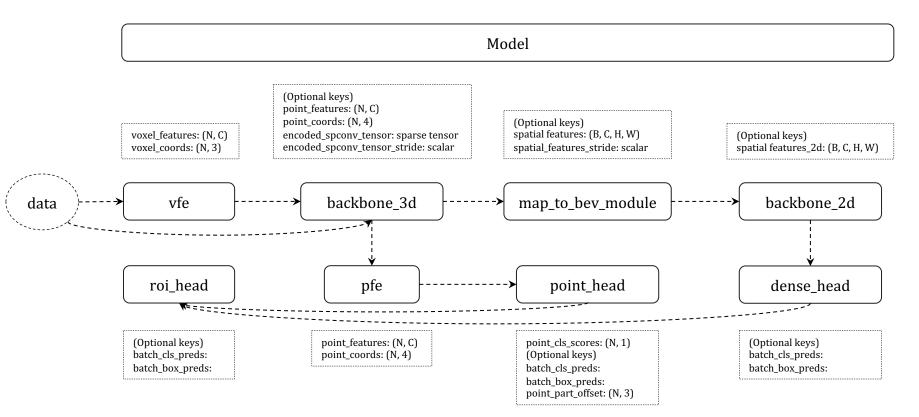


- Modularization to support various 3D detectors within one framework
  - Backbone: 3D Voxel CNN / Sparse Conv / PointNet++ / 2D CNN
  - DenseHead: Anchor-based / Anchor-free / Point-based
  - RoIHead: proposal generation / RoI Pooling / proposal target assigner



#### OpenPCDet: Model topology





#### OpenPCDet: Build detector with configuration dict



- Build your detector with customized configs
  - Detector3DTemplate.build\_networks()

```
self.module_topology = [
    'vfe', 'backbone_3d', 'map_to_bev_module', 'pfe',
    'backbone_2d', 'dense_head', 'point_head', 'roi_head'
```

### **OpenPCDet: Model forward & optimization**

- Call each module sequentially with their topology
- Optimization (Calculate the losses within each head):
  - DETECTOR.get\_training\_loss()
  - HEAD.get\_loss()



def forward(self, batch\_dict): batch\_dict = self.vfe(batch\_dict) batch\_dict = self.backbone\_3d(batch\_dict) batch\_dict = self.map\_to\_bev\_module(batch\_dict) batch\_dict = self.backbone\_2d(batch\_dict) batch\_dict = self.dense\_head(batch\_dict)

1 fi 2	<pre>m .detector3d_template import Detector3DTemplate</pre>
4 <b>c</b>	<pre>ss PointRCNN(Detector3DTemplate):</pre>
	<pre>definit(self, model_cfg, num_class, dataset):</pre>
	<pre>super()init(model_cfg=model_cfg, num_class=num_class, dataset=dataset)</pre>
	<pre>self.module_list = self.build_networks()</pre>
	<pre>def forward(self, batch_dict):</pre>
10	<pre>for cur_module in self.module_list:</pre>
11	<pre>batch_dict = cur_module(batch_dict)</pre>
12	
13	if self.training:
14	<pre>loss, tb_dict, disp_dict = self.get_training_loss()</pre>
15	
16 17	<pre>ret_dict = {     'loss': loss</pre>
18	}
10 19	return ret dict, tb dict, disp dict
20	else:
21	<pre>pred_dicts, recall_dicts = self.post_processing(batch_dict)</pre>
22	return pred_dicts, recall_dicts
23	
24	<pre>def get_training_loss(self):</pre>
25	disp_dict = {}
26	<pre>loss_point, tb_dict = self.point_head.get_loss()</pre>
27	<pre>loss_rcnn, tb_dict = self.roi_head.get_loss(tb_dict)</pre>
28	
29	loss = loss_point + loss_rcnn
30	return loss, tb_dict, disp_dict
31	

#### OpenPCDet: DetectorTemplate

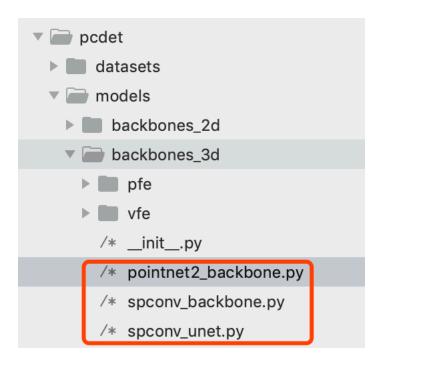
- Task: The top module for a detector
  - Build\_networks
  - Forward
  - Loss calculation
  - Post\_processing (NMS + score threshold)



▼ 📄 pcdet		
datasets		
<ul> <li>models</li> </ul>		
backbones_2d		
backbones_3d		
dense_heads		
detectors	ר	
∕∗initpy		
<pre>/* detector3d_template.py</pre>	,	
/∗ PartA2_net.py		
<pre>/* point_rcnn.py</pre>		
∕∗ pointpillar.py		
∕∗ pv_rcnn.py		
/* second_net.py		

#### OpenPCDet: 3D backbone networks

- Task: extract voxel-wise / point-wise features
- 3D encoder with sparse convolution (with VFE):
  - VoxelBackBone8x
  - VoxelResBackBone8x
- 3D UNet with sparse convolution (with VFE):
  - UNetV2
- Point-wise networks (PointNet++)
  - PointNet2MSG

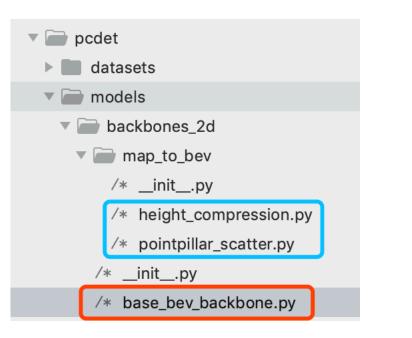


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#### OpenPCDet: 2D backbone networks

- Task: extract 2D feature maps
  - (with map\_to\_bev\_module)
- Map\_to\_bev\_module (map 3D features to 2D maps):
  - HeightCompression
  - PointPillarScatter
- 2D convolution encoder with FPN-like upsampling
  - BaseBEVBackbone

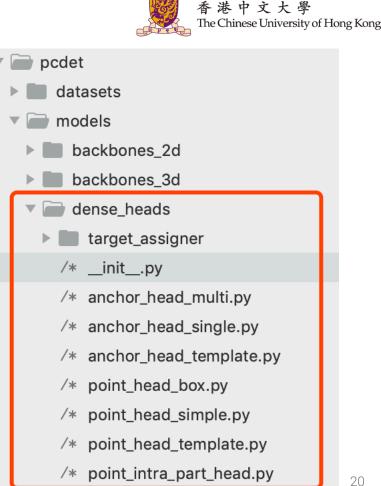


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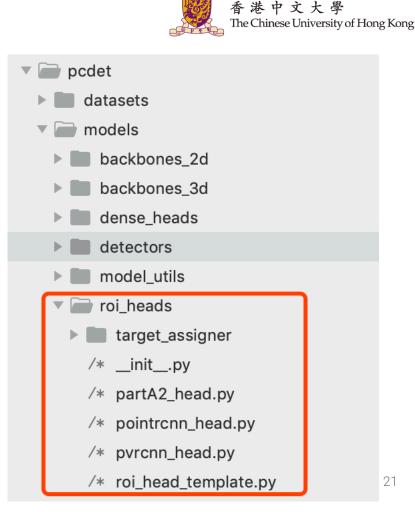
### **OpenPCDet:** Denseheads

- Task: Generate dense 3D boxes
  - Target assigning 0
  - Head loss calculation  $\bigcirc$
- Dense head with BEV features (AnchorHeadTemplate):
  - AnchorHeadSingle Ο
  - AnchorHeadMulti  $\bigcirc$
  - CenterHead (anchor-free) Ο
- Dense head with point-wise features (PointHeadTemplate):
  - PointHeadSimple Ο
  - PointHeadBox  $\bigcirc$
  - PointIntraPartOffsetHead  $\bigcirc$



### **OpenPCDet:** RoIHeads

- Task: Refine 3D proposals with RoI-aligned features
  - Extract RoI-aligned features
  - proposal\_layer
  - ProposalTargetLayer
  - Head loss calculation
- Stage-II RoI refinement (RoIHeadTemplate):
  - PointRCNNHead
  - PartA2FCHead
  - PVRCNNHead



#### **OpenPCDet: Configuration files**



- Hierarchical configuration with YAML file
- Each model takes its own config

1 CLASS_N/	<pre>MES: ['Car', 'Pedestrian', 'Cyclist']</pre>	<pre>class DatasetTemplate(torch_data.Dataset):</pre>	
2			None):
3 DATA_CO		<pre>super()init()</pre>	
4 _BA	<pre>SE_CONFIG_: cfgs/dataset_configs/kitti_dataset.yaml</pre>		
5			
6		<pre>class Detector3DTemplate(nn.Module):</pre>	
7 MODEL:		<pre>definit(self, model_cfg, num_class, dataset):</pre>	
	E: SECONDNet		
9			
10 VFE			
11	NAME: MeanVFE		
12		<pre>v class AnchorHeadTemplate(an_Module):</pre>	
	KBONE_3D:	<pre>v class Anchornead Comparison of the second se</pre>	
14	NAME: VoxelBackBone8x		
15 10			
	_T0_BEV: 🚥		
19			
20 BACI 28	KBONE_2D:		
	SE HEAD:		
29 DEN. 86	SE_READ:		
	PROCESSING:		
100 FUS			
100 101			
	ATION: 🚥		00
122			22

## OpenPCDet example: Build one-stage detector



• SECOND / PointPillar

	CLASS_NAMES: ['Car', 'Pedestrian', 'Cyclist']
2	
2 3	DATA_CONFIG:
	<pre>BASE_CONFIG_: cfgs/dataset_configs/kitti_dataset.ya</pre>
	MODEL:
8	NAME: SECONDNet
10	VFE:
11	NAME: MeanVFE
12	
13	BACKBONE 3D:
14	NAME: VoxelBackBone8x
15	
16	MAP_TO BEV: ==
19	
20	BACKBONE_2D:
28	
29	DENSE HEAD:
86	
87	POST PROCESSING: ==
100	
101	
102	OPTIMIZATION:
122	
122	

1	<pre>CLASS_NAMES: ['Car', 'Pedestrian', 'Cyclist']</pre>
2	
3	DATA_CONFIG:
49	
50	MODEL:
51	NAME: PointPillar
52	
53	VFE: 🚥
59	
60	MAP_TO_BEV: 🚥
63	
64	BACKBONE_2D: 🚥
71	
72	DENSE_HEAD: 🚥
129	
130	POST_PROCESSING: 🚥
143	
144	
145	OPTIMIZATION:
165	



### OpenPCDet example: Build two-stage detector



• PointRCNN / PV-RCNN

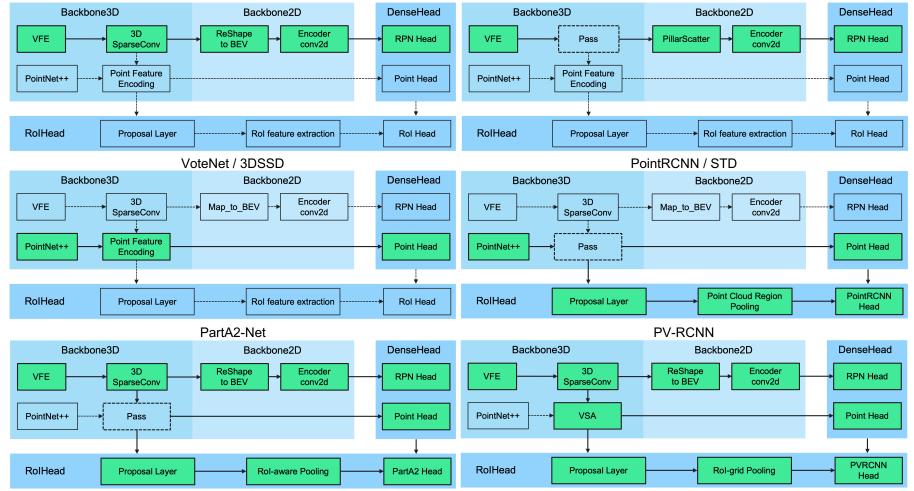
1	CLASS_NAMES: ['Car', 'Pedestrian', 'Cyclist']
2 3	
3	DATA_CONFIG:
21	
22	MODEL:
23	NAME: PointRCNN
24	
25	BACKBONE_3D: 🚥
36	·
37	POINT_HEAD: ==
62	
63	ROI_HEAD: 🚥
125	
126	POST_PROCESSING:
139	
140	
141	OPTIMIZATION: ••
161	

1	CLASS_NAMES: ['Car', 'Pedestrian', 'Cyclist']
2 3	DATA_CONFIG:
31	
32	MODEL:
33	NAME: PVRCNN
34	
35	VFE:
36	NAME: MeanVFE
37	
38 39	BACKBONE_3D: NAME: VoxelBackBone8x
- 59 - 40	NAME: VOXetbackboneox
40	MAP_TO_BEV:
44	
45	BACKBONE_2D: ==
53	
54	DENSE_HEAD: ==
111	
112	PFE:
145	
146	POINT_HEAD: 🚥
158	
159 214	ROI_HEAD: ••
214	POST_PROCESSING: -
228	
229	
230	OPTIMIZATION: ==
250	

PointRCNN

VoxelNet / SECOND / VoxelFPN

PointPillar / PIXOR / CenterPoint



### OpenPCDet: 3D Detection Toolbox



- How to add customized dataset ?
  - 1. Write your own DatasetModule that inherited from DatasetTemplate
  - 2. Overload the self.\_\_getitem\_\_() function to load point clouds / gt\_boxes and transform them to the unified coordinate and box definition of OpenPCDet
  - 3. Call self.prepare\_data() to process the data
  - 4. Overload the self.generate\_prediction\_dicts() function to transform the predicted results to the format what you like.
  - 5. Overload the self.evaluation() function to evaluate the results with your own metric.

#### OpenPCDet: 3D Detection Toolbox



- How to support more models ?
  - 1. Write your own detector that inherited from DetectorTemplate
  - 2. Write your own configuration files
  - 3. Write your own modules to specific directories if we do not provide it
  - 4. Overload the forward() functionn
  - 5. Overload the get\_training\_loss() function

# OpenPCDet: Tips with OpenPCDet



- Group the configuration files into different directories
- Multi-gpu training and multi-gpu testing
- View the training process in the tensorboard
- Start a separate evaluation program for fast training
- The following command line parameters are useful:
  - --extra\_tag
  - --pretrained\_model
  - --set
  - --eval\_tag
  - --eval\_all

Training script:

bash scripts/dist\_train.sh 8 --cfg\_file cfgs/kitti\_models/pv\_rcnn.yaml --extra\_tag ex1\_tag --set OPTIMIZATION. LR 0.03

Testing script:

sh scripts/dist\_test.sh 8 --cfg\_file cfgs/kitti\_models/pv\_rcnn.yaml --extra\_tag ex1\_tag --eval\_all --eval\_tag nms02 --set MODEL.POST\_PROCESSING.NMS\_CONFIG.NMS\_THRESH 0.2

# Q & A

#### Welcome to Star / Fork / PR to OpenPCDet (https://github.com/open-mmlab/OpenPCDet)

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