

# Point-Voxel CNN for Efficient 3D Deep Learning

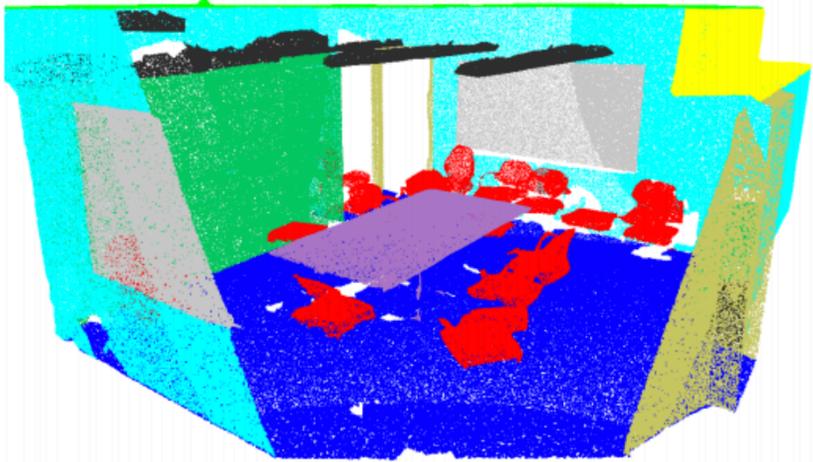
**Zhijian Liu\***, **Haotian Tang\***, **Yujun Lin**, and **Song Han**

Project Page: <http://pvcnn.mit.edu/>

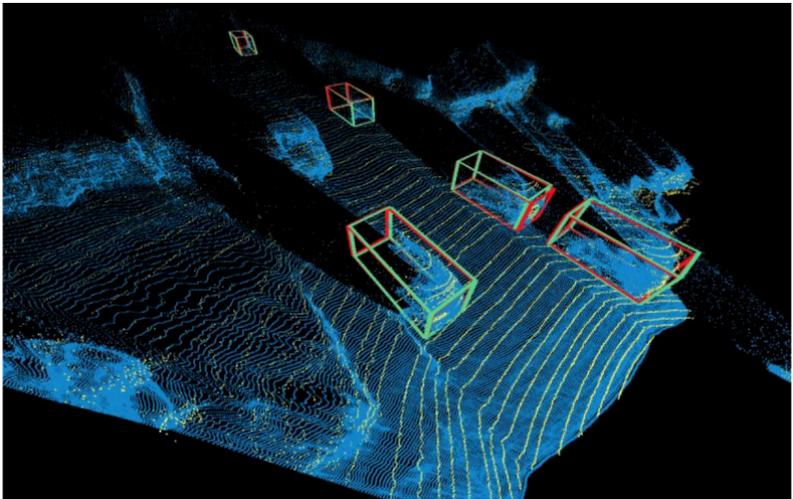
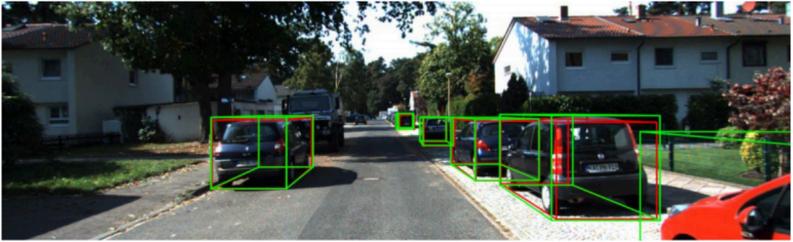
# 3D Deep Learning



**3D Part Segmentation**  
(for Robotic Systems)

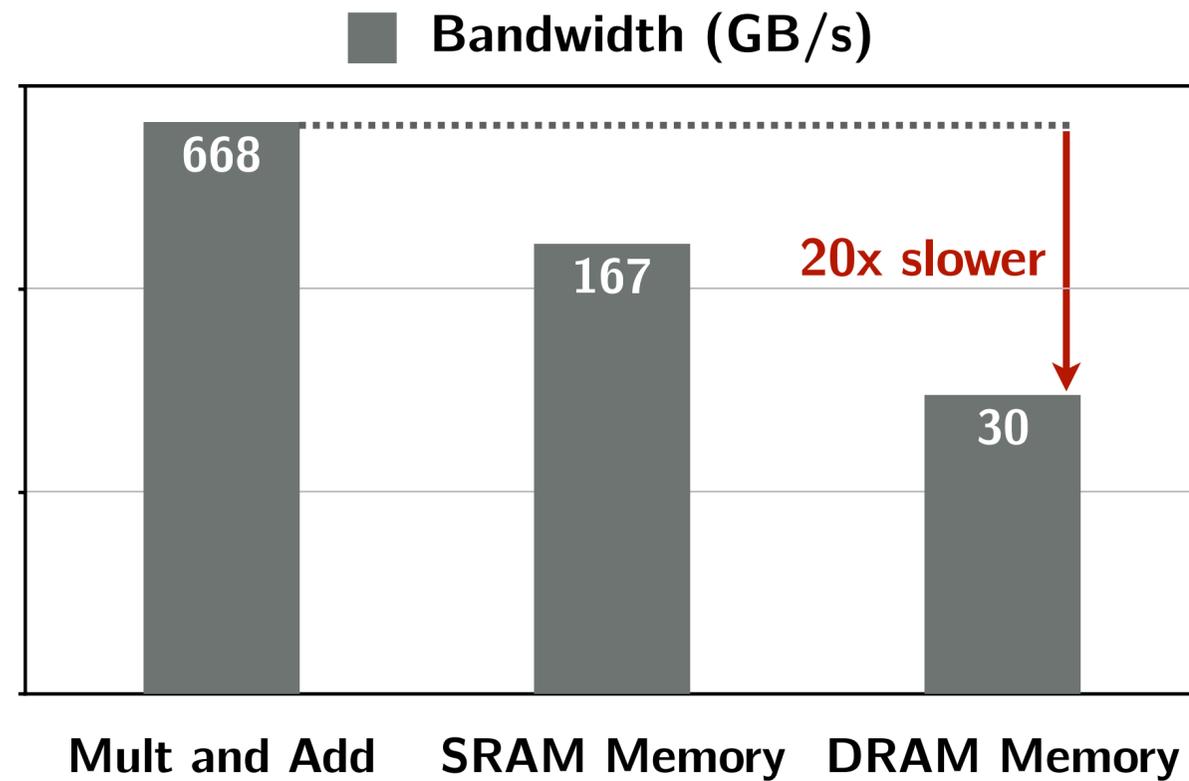


**3D Semantic Segmentation**  
(for VR/AR Headsets)



**3D Object Detection**  
(for Self-Driving Cars)

# Efficient 3D Deep Learning



**Off-chip DRAM access** is much more expensive than arithmetic operation!



Sequential Memory Access

1	2	3	4	5	6	7	8
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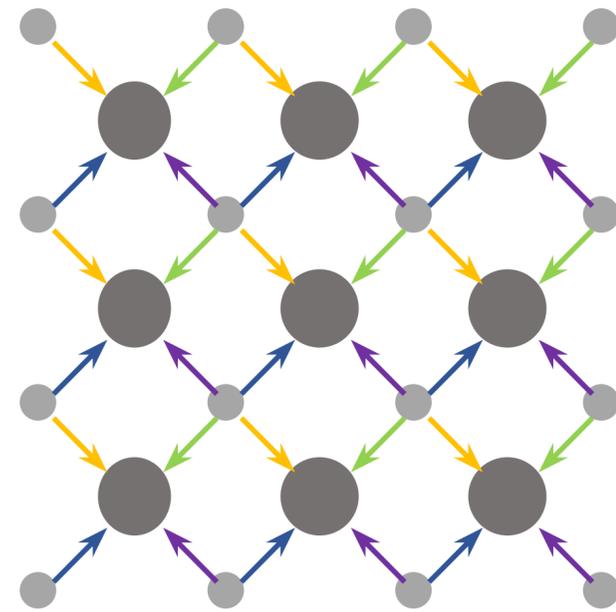


Random Memory Access

7	5	2	4	6	1	8	3
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**Random memory access** is inefficient due to the potential bank conflicts!

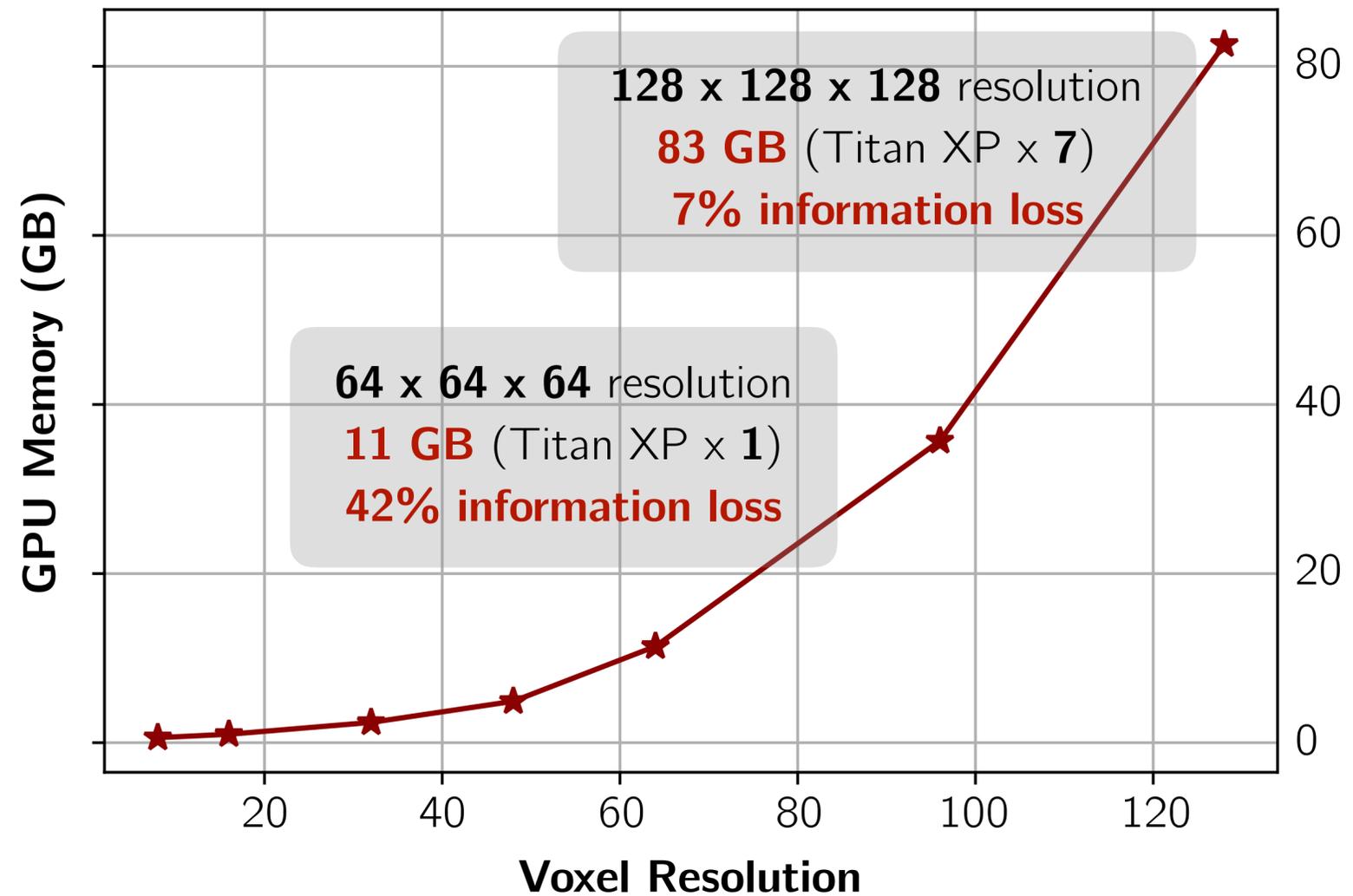
# Voxel-Based Models: Cubically-Growing Memory



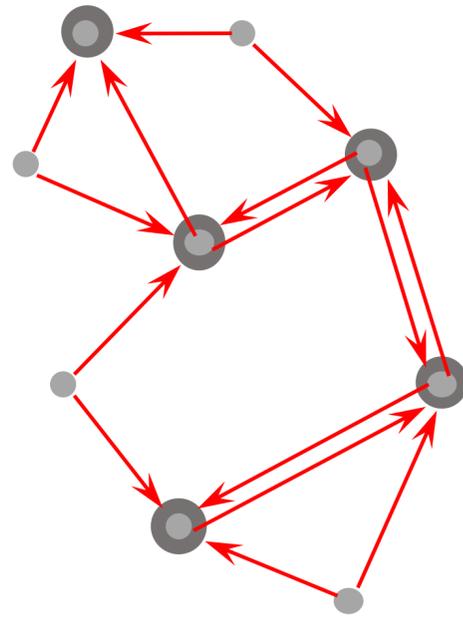
3D ShapeNets [CVPR'15]

VoxNet [IROS'15]

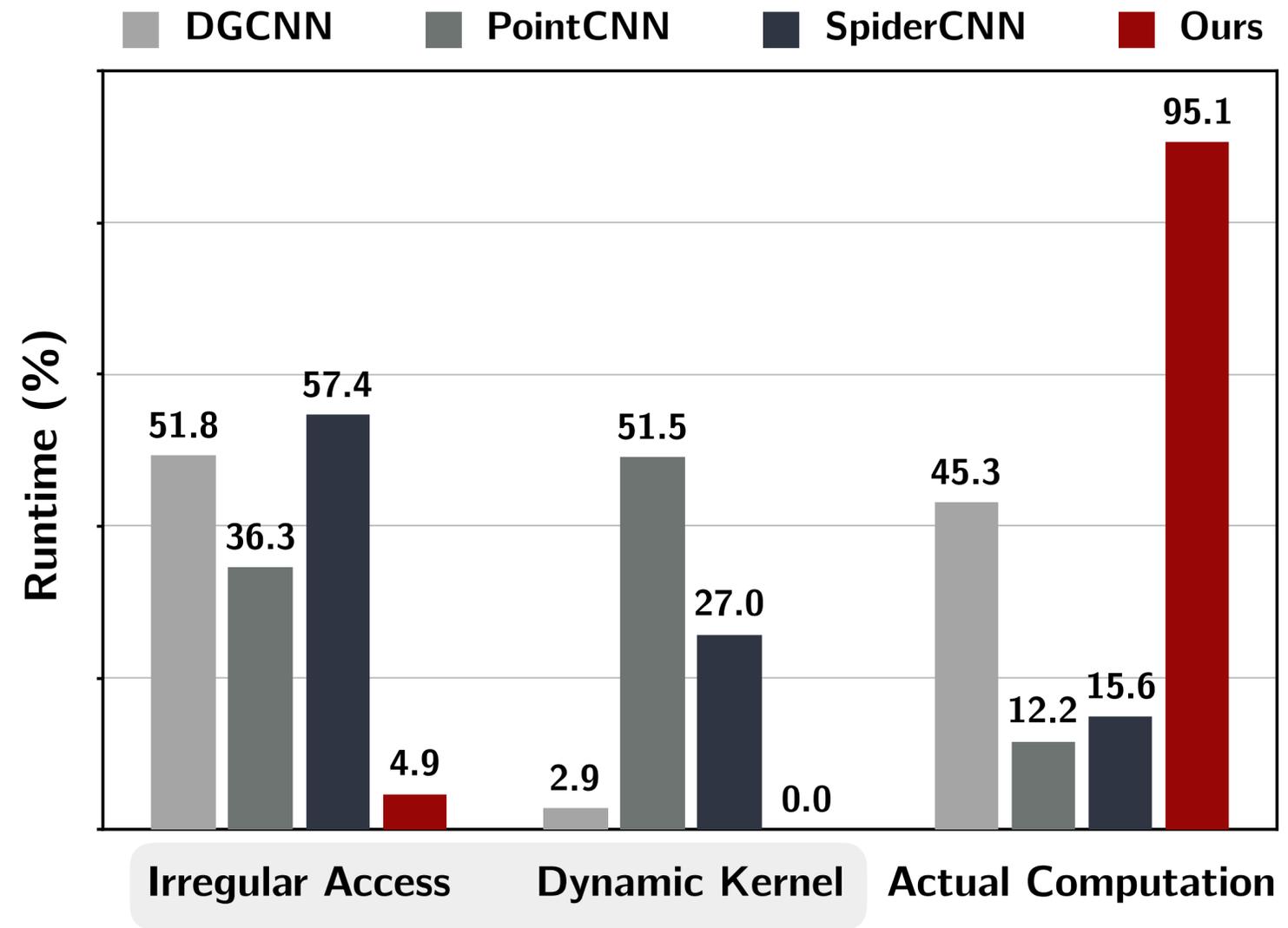
3D U-Net [MICCAI'16]



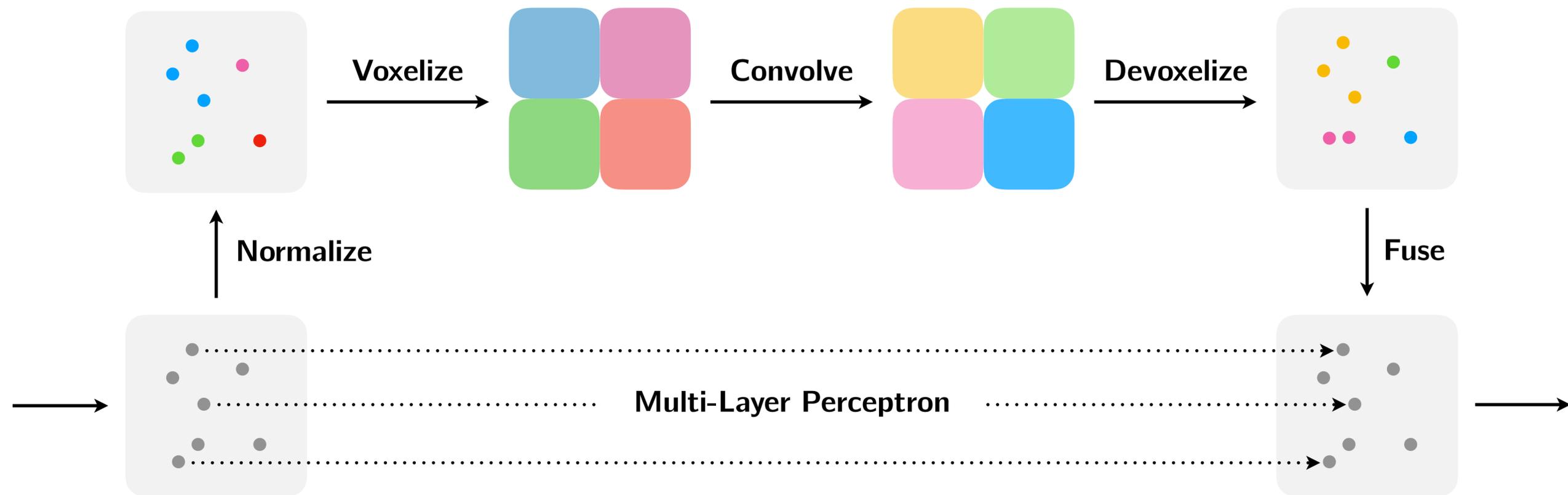
# Point-Based Models: Sparsity Overheads



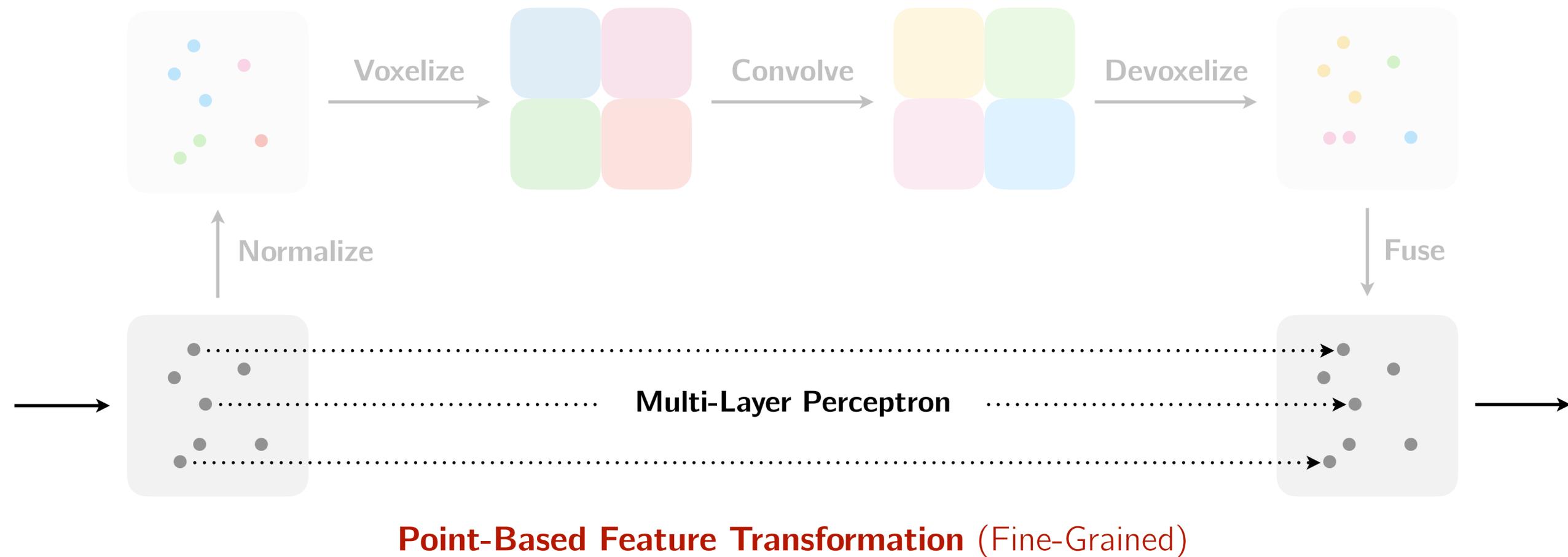
PointNet [CVPR'17]  
PointCNN [NeurIPS'18]  
DGCNN [SIGGRAPH'19]



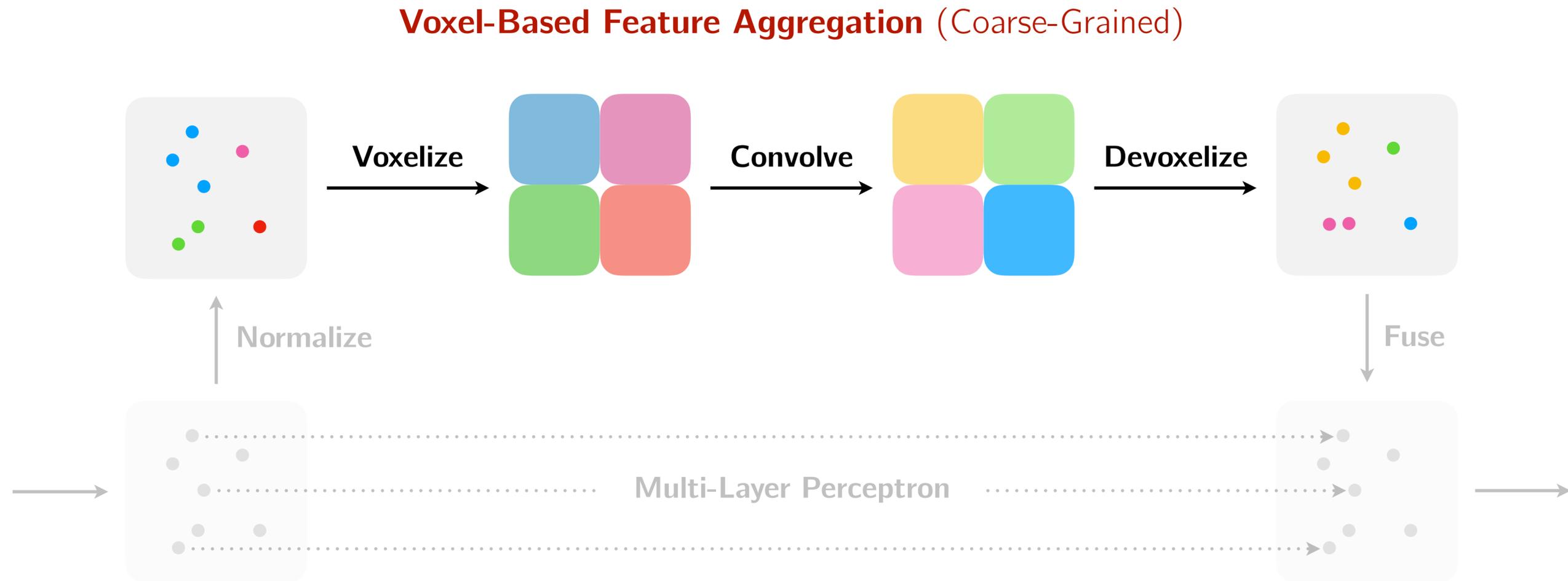
# Point-Voxel Convolution (PVConv)



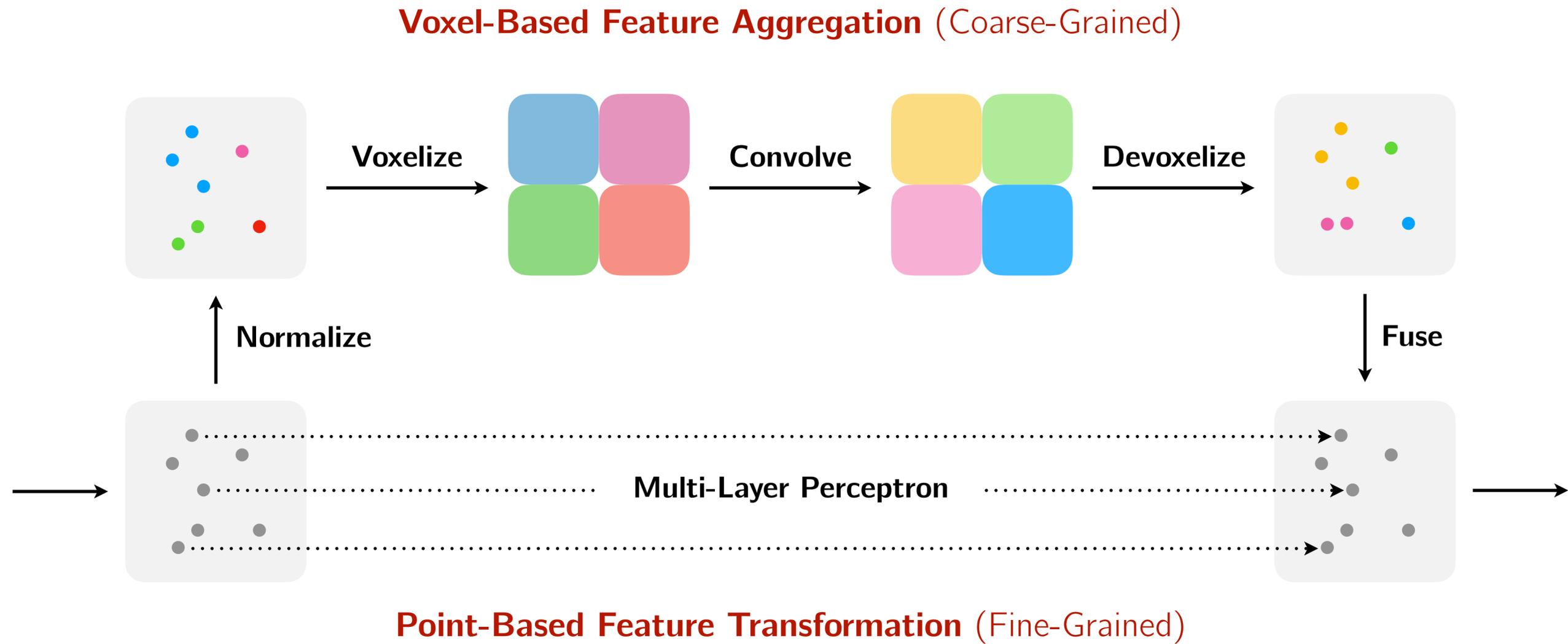
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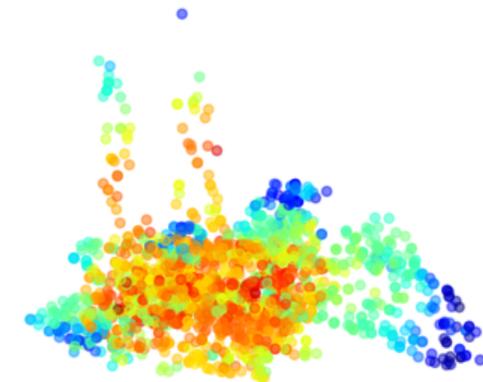
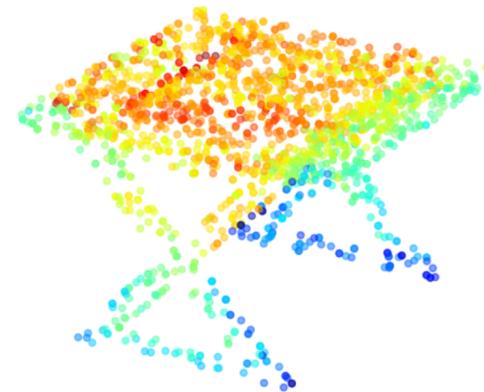
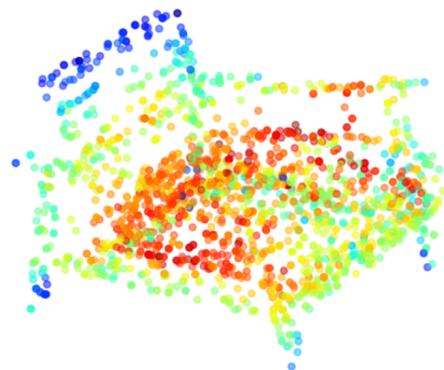
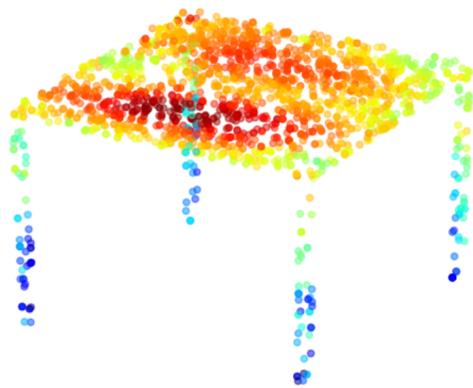


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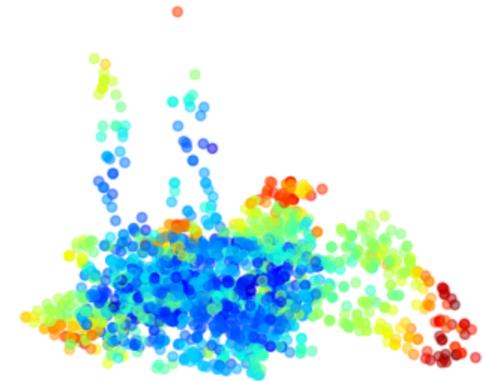
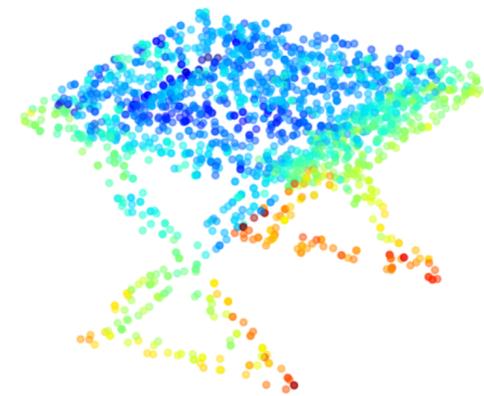
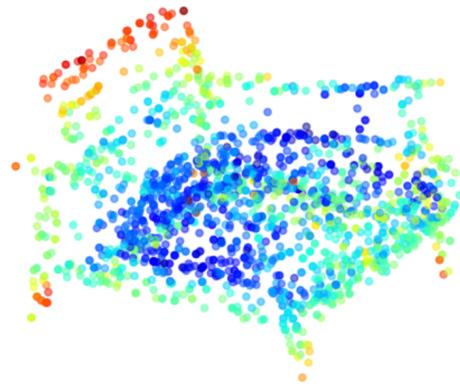
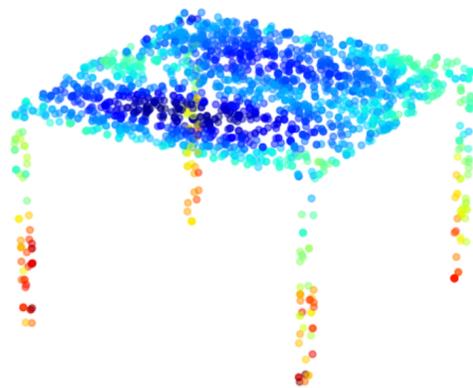


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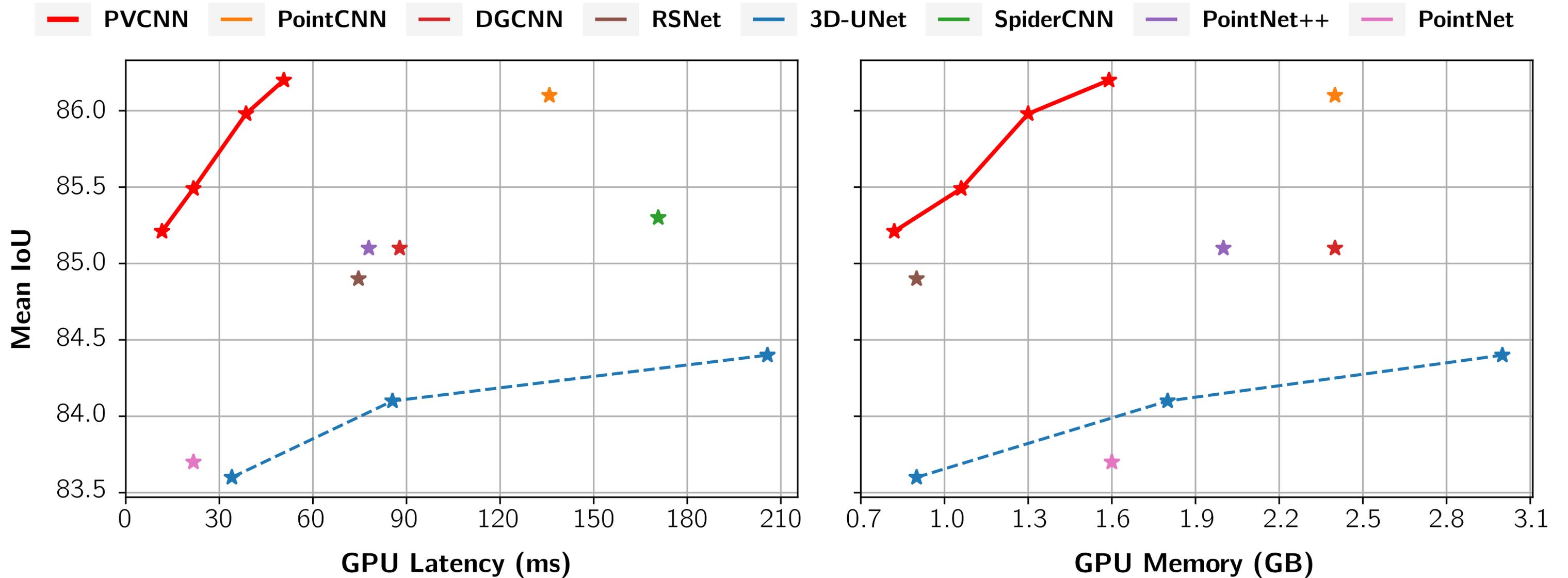
Features from **Voxel-Based Branch**:



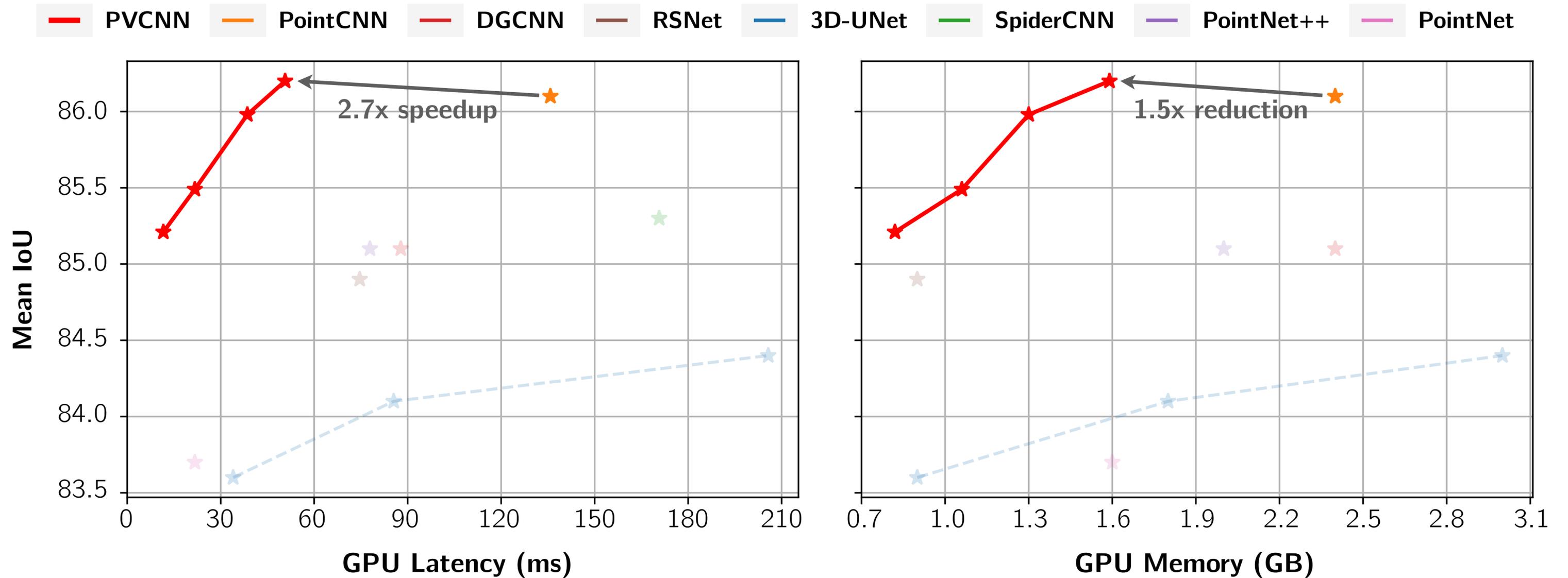
Features from **Point-Based Branch**:



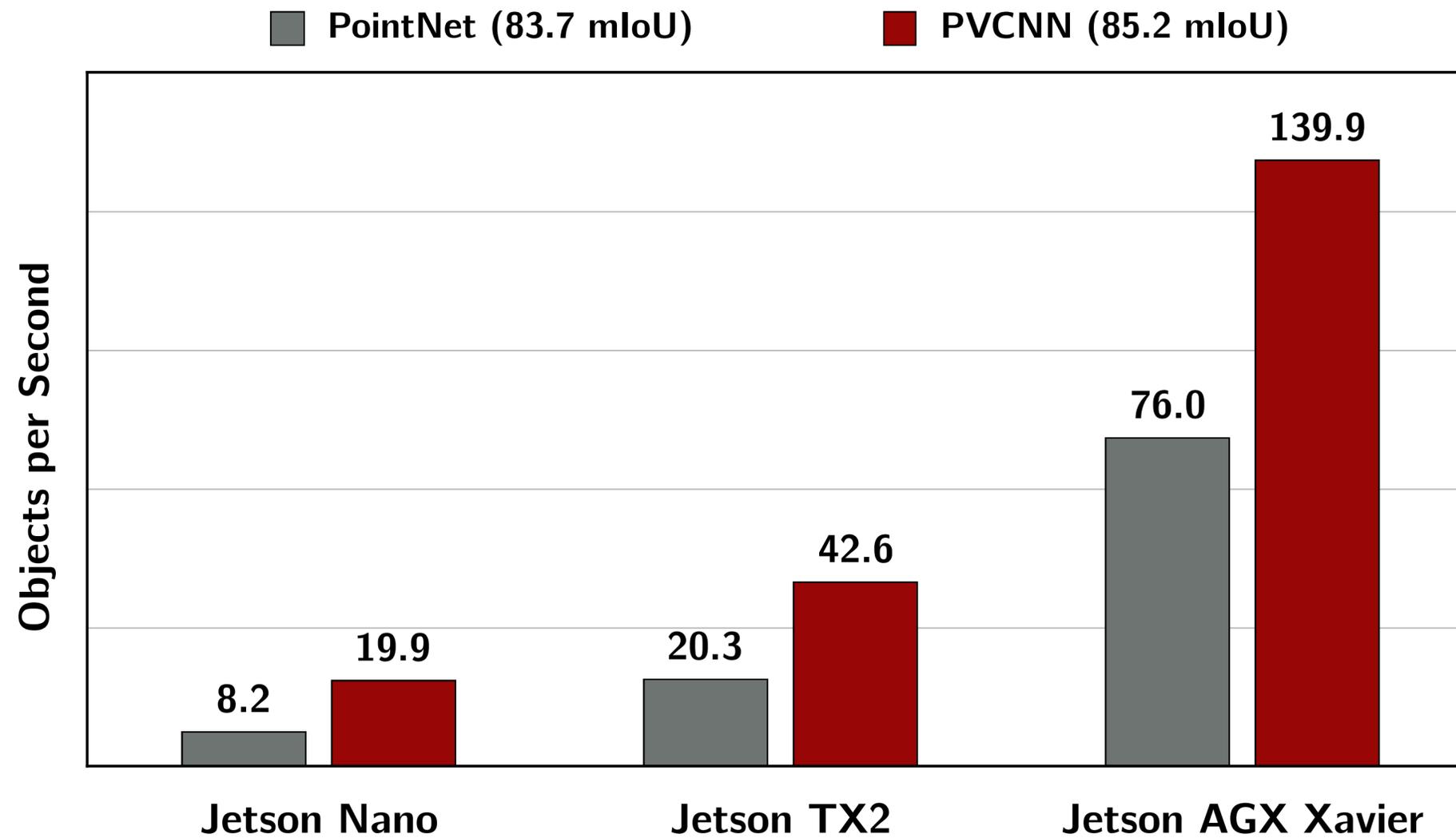
# Results: 3D Part Segmentation (ShapeNet)



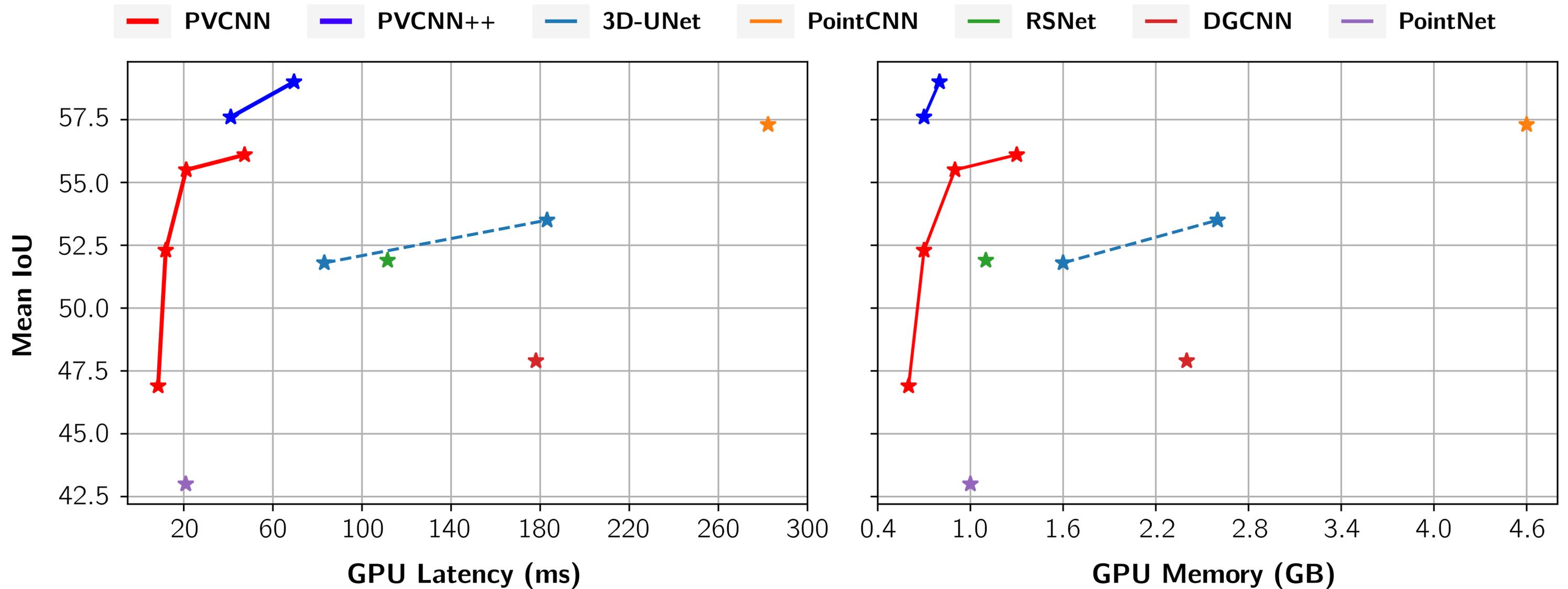
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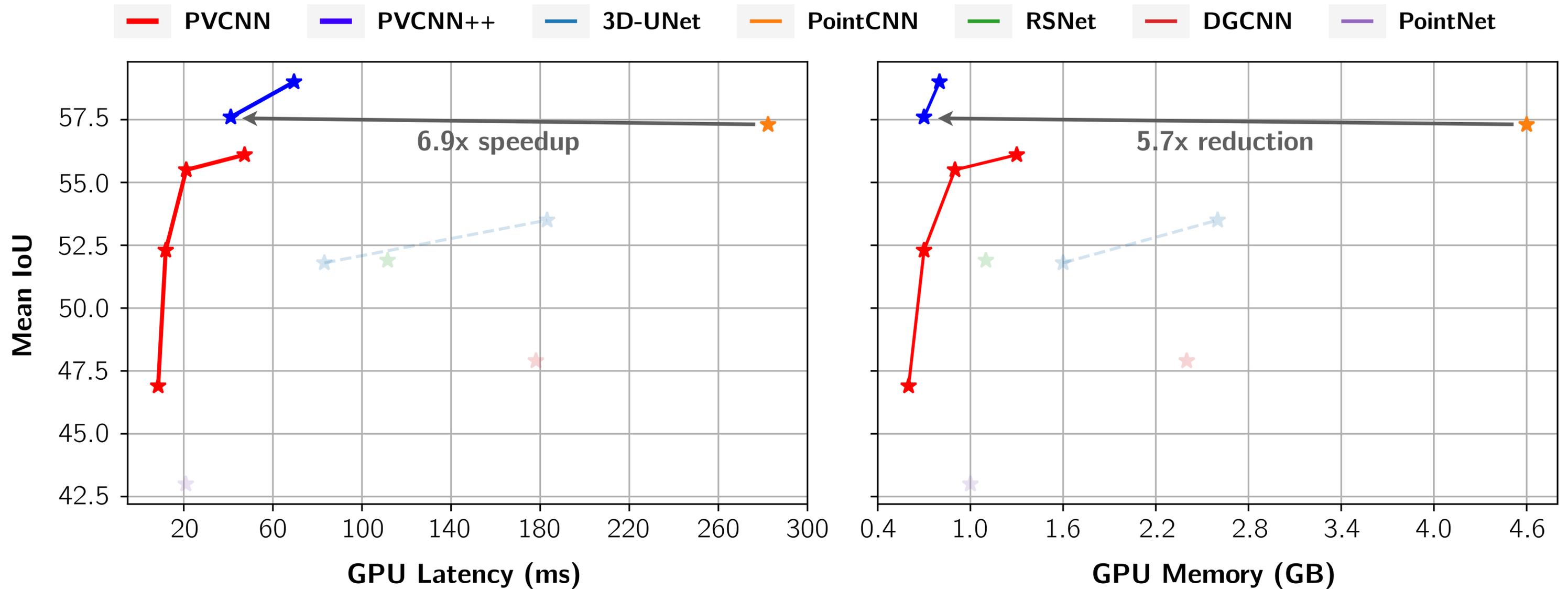
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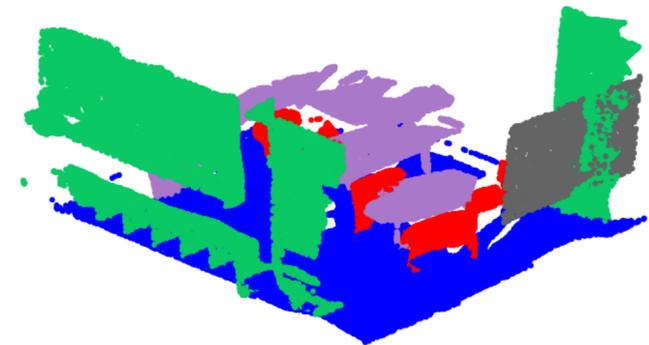
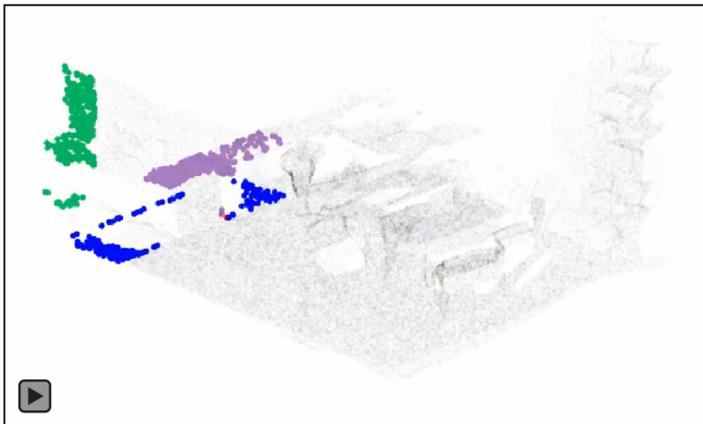
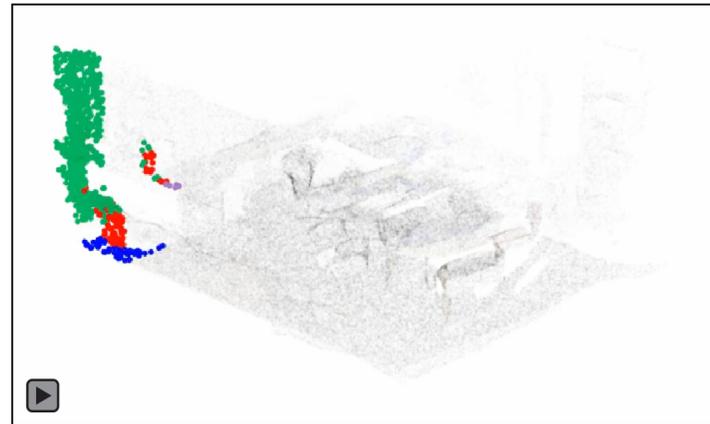
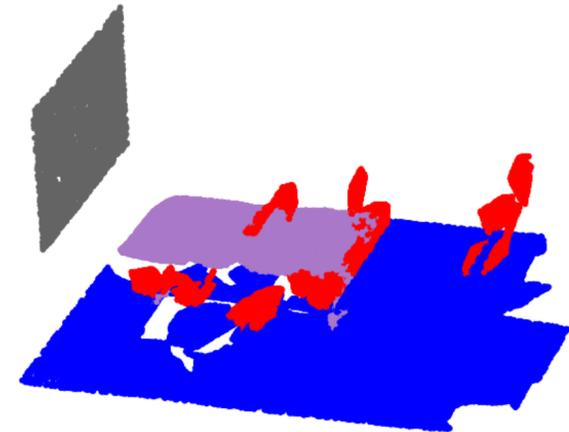
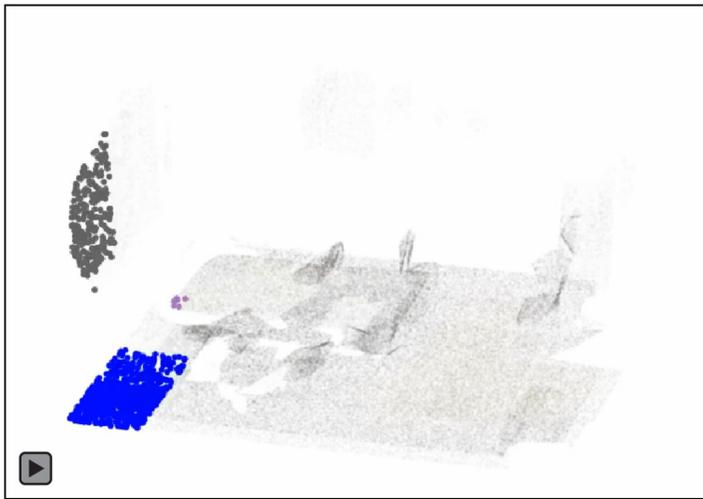
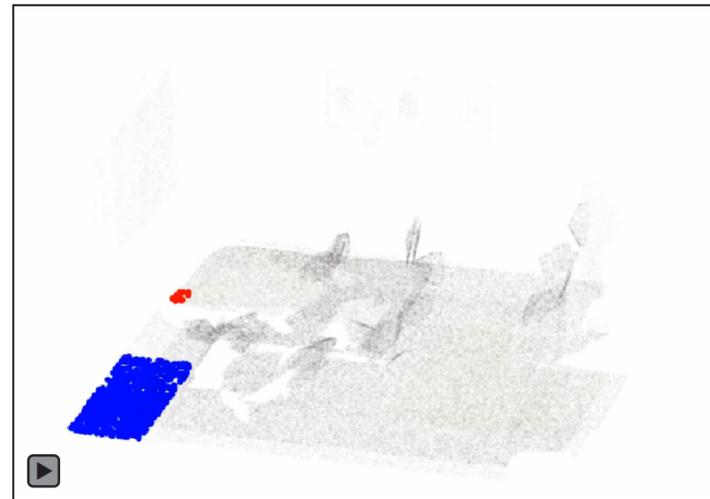
# Results: 3D Semantic Segmentation (S3DIS)



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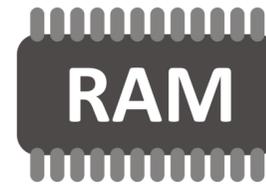
Input Scene

PointNet

PVCNN  
(1.8x faster)

Ground Truth

# Results: 3D Object Detection (KITTI)



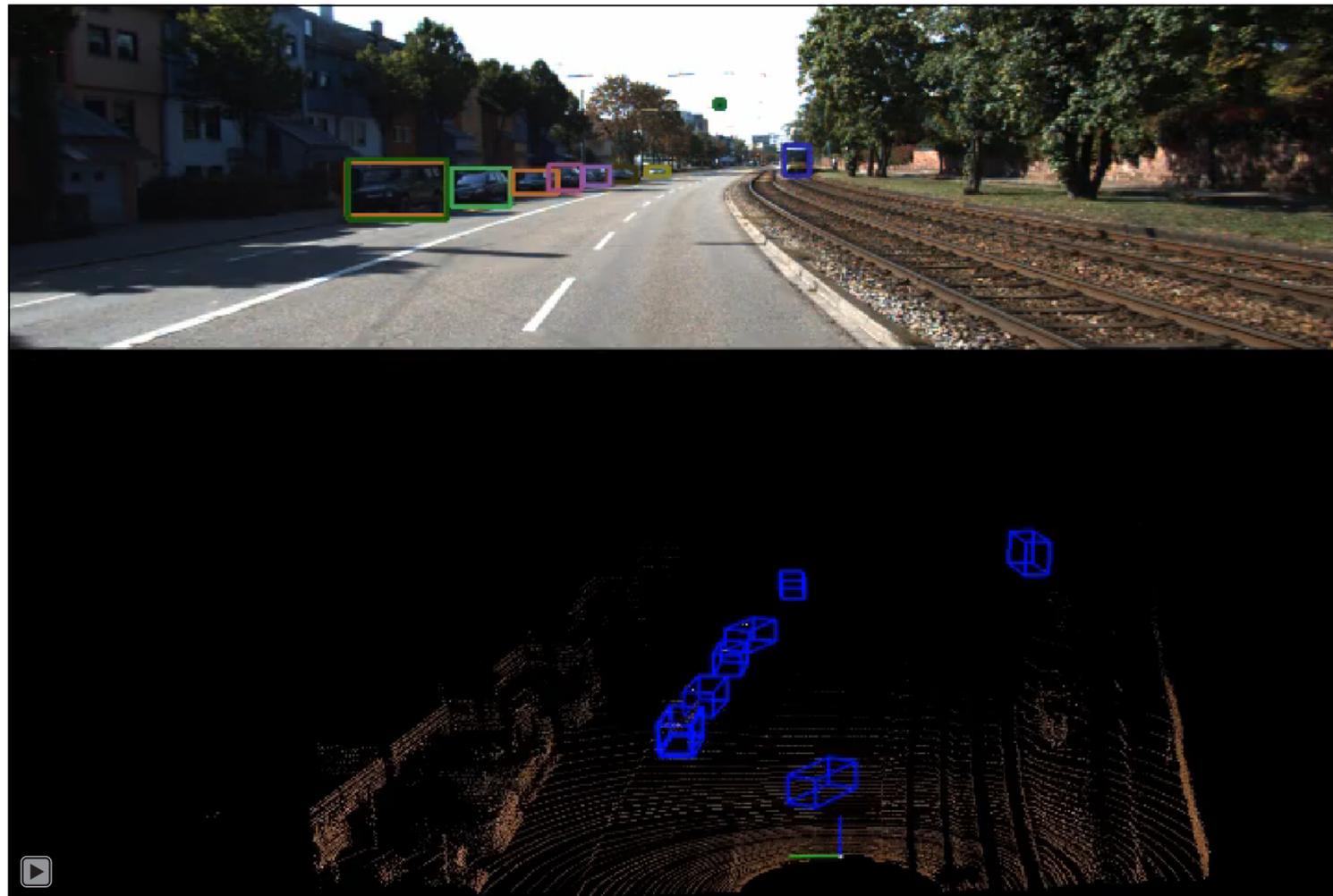
	GPU Latency	GPU Memory	Pedestrian	Cyclist	Car
<b>F-PointNet++</b>	105.2 ms	2.0 GB	61.6	62.4	72.8
<b>PVCNN</b> (efficient)	<b>58.9 ms</b> <b>(1.8x)</b>	<b>1.4 GB</b> <b>(1.4x)</b>	60.7 (-0.9)	63.6 (+1.2)	73.0 (+0.2)
<b>PVCNN</b> (complete)	<b>69.6 ms</b> <b>(1.5x)</b>	<b>1.4 GB</b> <b>(1.4x)</b>	<b>64.9</b> <b>(+3.3)</b>	<b>65.9</b> <b>(+3.5)</b>	<b>73.1</b> <b>(+0.3)</b>

**Faster**

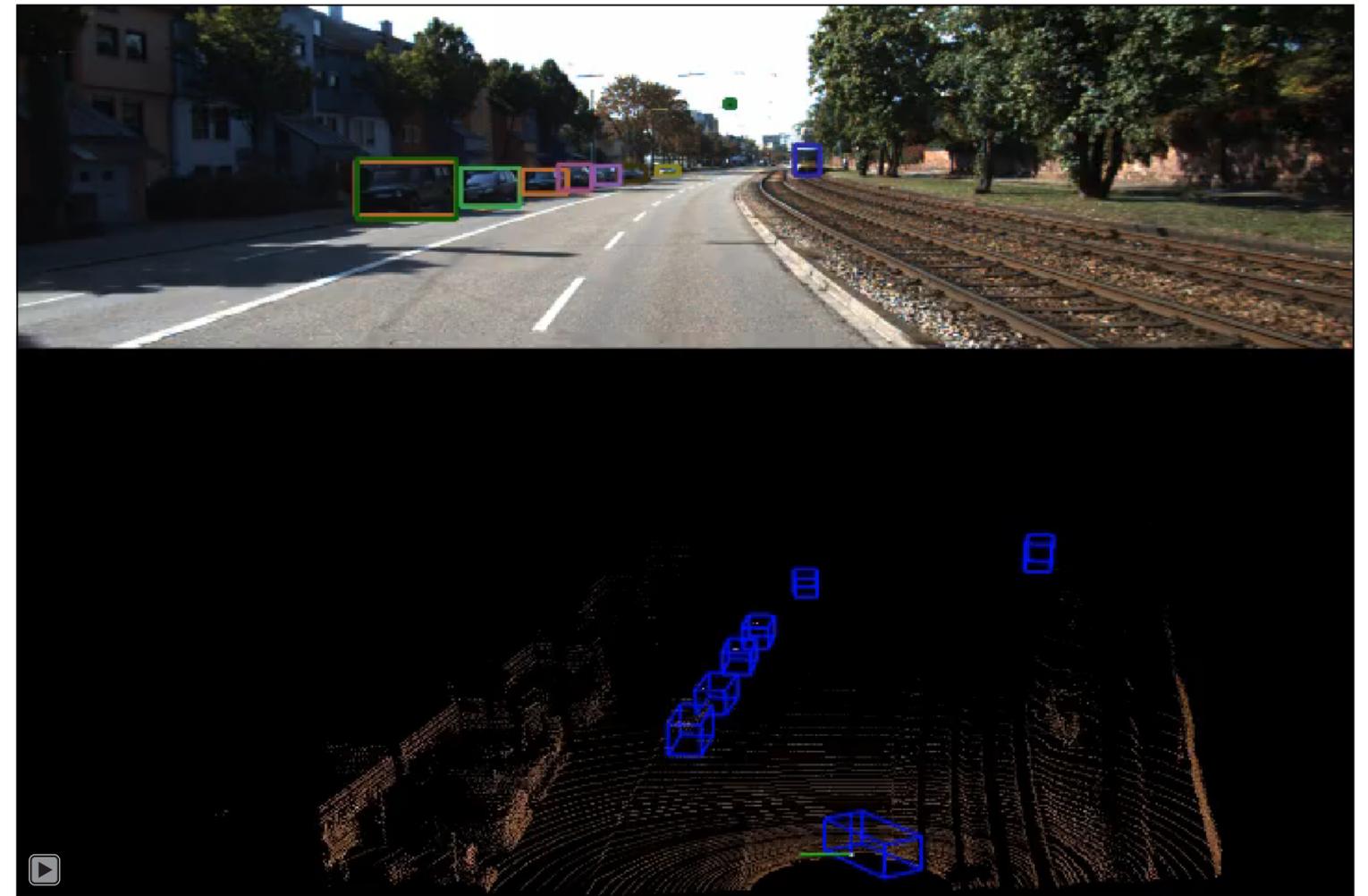
**Lower**

**More Accurate**

# Results: 3D Object Detection (KITTI)



**F-PointNet++**  
(10 FPS)

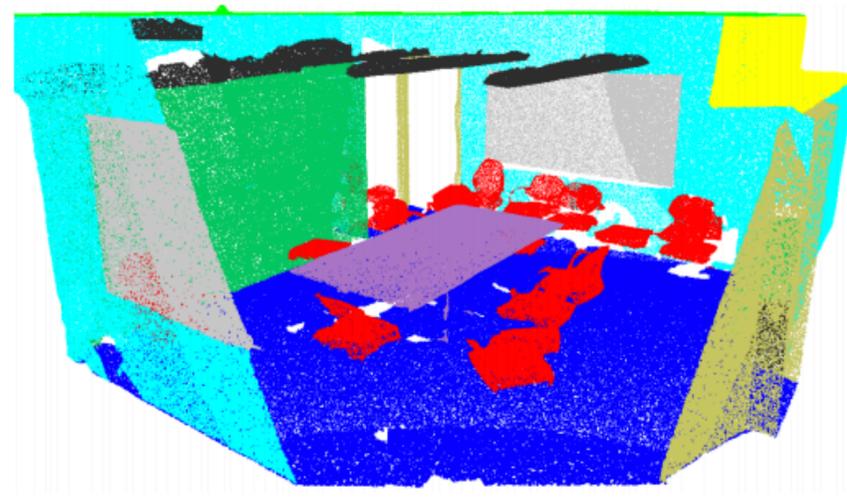


**PVCNN**  
(17 FPS, 1.8x faster)

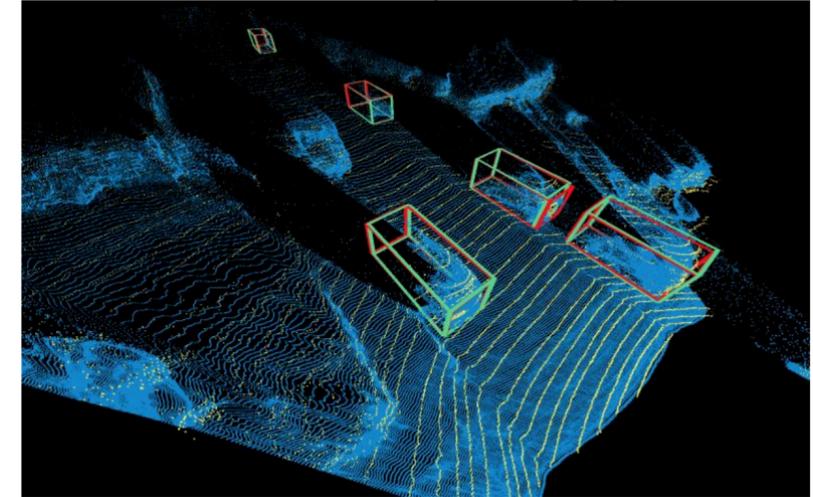
# Point-Voxel CNN for Efficient 3D Deep Learning



**2.7x** measured speedup  
**1.5x** memory reduction



**6.9x** measured speedup  
**5.7x** memory reduction



**1.8x** measured speedup  
**1.4x** memory reduction

**Gold Medal** in Lyft Challenge on 3D Object Detection for Autonomous Vehicles

Poster: 10:45-12:45 PM @ East Exhibition Hall B + C #112

GitHub: <https://github.com/mit-han-lab/pvcnn>

Project Page: <http://pvcnn.mit.edu>