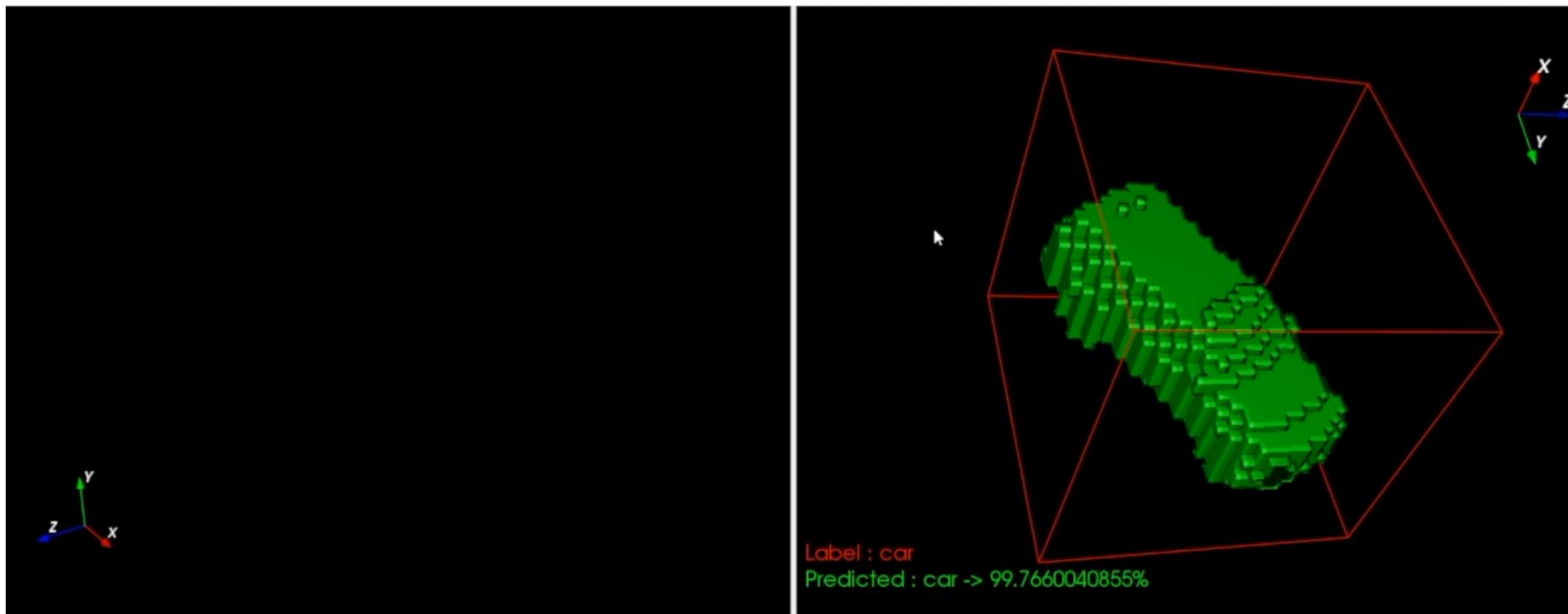
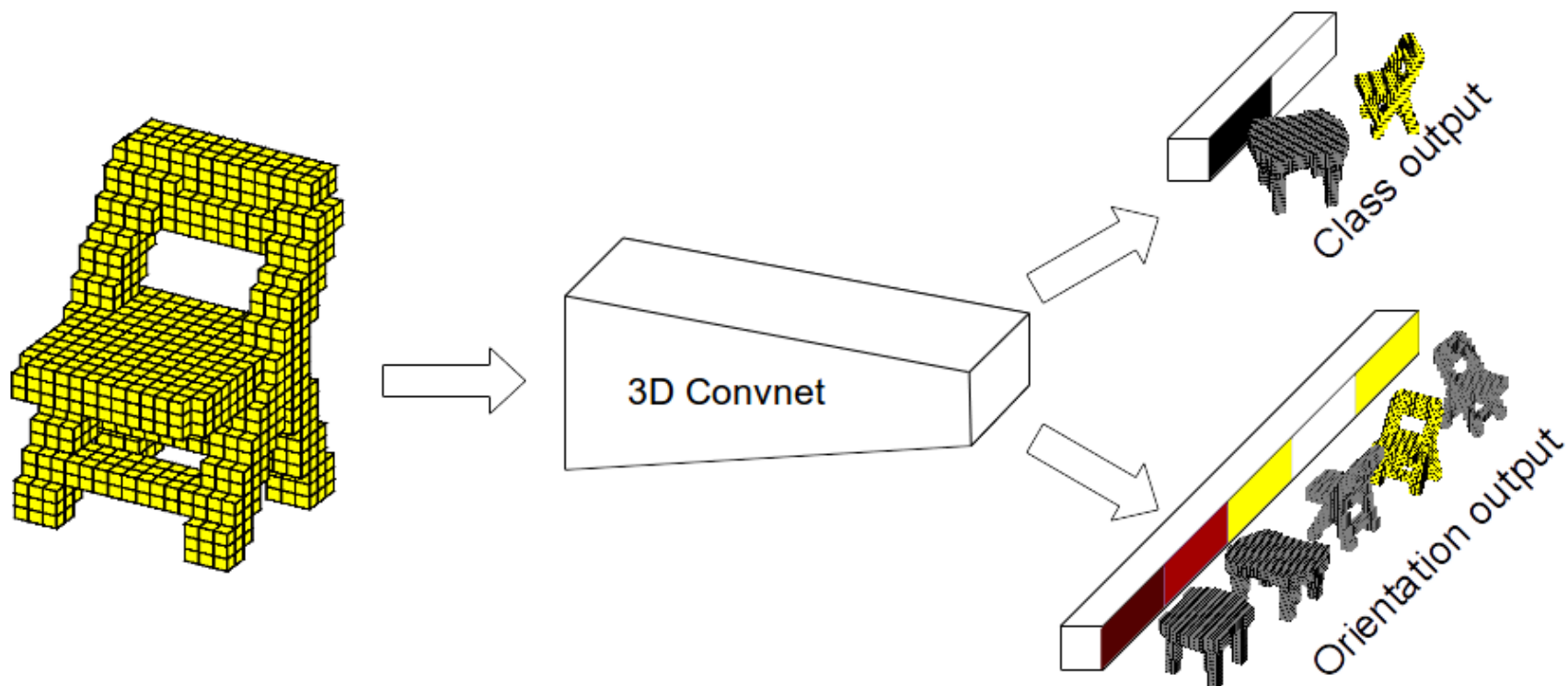


# Orientation-boosted Voxel Nets for 3D Object Recognition

# Object identification and 3D CNN



category-level classification task as a multi-task problem



# Method

- ORientation-boosted vOxel Net- ORION
  - Z-axis
- Sensors:
  - LiDAR (3D Laser scanning)
  - RGBD cameras (Depth as a fourth input)
  - Synthetic/CAD
- Experiments:
  - Detection
  - Classification

# ORION

- <https://github.com/lmb-freiburg/orion>
- Language: Matlab
- Built on the Voxnet architecture

# Datasett

- Sydney Urban Objects - LiDAR/Pointcloud
- NYUv2 - Kinect/RGBD
- ModelNet - Synthetic/CAD
- KITTI - LiDAR/Pointcloud

# Sydney Urban Objects - LiDAR/Pointcloud

- LiDAR scans of 631 objects in 26 categories
- Incomplete point-clouds
  - Single viewpoint
- Converted to voxel-grids of size 32x32x32

# NYUv2 - Kinect/RGBD

- 2808 RGBD images, 10 object classes
- Converted to voxel-grids of size 32x32x32
- Orientation annotation from SUN-RGBD benchmark



# ModelNet - Synthetic/CAD

- synthetic CAD models
- uniformly aligned objects of the same classes
- voxel grids of size 28x28x28
- Manual annotation and automated annotations

# KITTI - LiDAR/Pointcloud

- Object detection: 7481 training images and 7518 test images
- Only used for the detection experiment

# Classification results

Method↓		Dataset					
		# Conv	# param	Sydney	NYUv2	ModelNet10	
Hand-crafted feat.	Recursive D [31]	-	-	-	37.6	-	
	Recursive D+C [31]	-	-	-	44.8	-	
	Triangle+SVM [0]	-	-	67.1	-	-	
	GFH+SVM [0]	-	-	71.0	-	-	
Deep Network	FusionNet [33]		118M	-	-	93.1	
	VRN <sup>†</sup> [0]	43	18M	-	-	93.6	
Shallow Network	ShapeNet [38]	3	-	-	57.9	83.5	
	DeepPano [29]	4	-	-	-	85.5	
	VoxNet [21] (baseline)	2	890K	72	71	92	
	ORION (Ours)		2	910K	<b>77.8</b>	<b>75.4</b>	<b>93.8</b>
			4	4M	<b>77.5</b>	<b>75.5</b>	<b>93.9</b>

# Classification results

- Results are just over the state of the art alternatives
  - Far shallower architecture than previous state of the art
    - 2-conv layers vs 43
- Slightly deeper networks => overfitting (Too small dataset)

# Results: Sydney Urban Objects - LiDAR/Pointcloud

- Average F1 score
  - Dataset unbalanced but taken into account
  - 4 folds/subset used for cross validation (3 folds training, 1 testing)
    - Small dataset => ran three times with different seeds

# Results: ModelNet - Synthetic/CAD

Method	Conv. Layers	Batch Norm.	Accuracy (%)		
			No Alignment	Rough, Automatic Alignment	Perfect, Manual Alignment
VoxNet [21] (baseline)	2	×	83	-	-
ORION (Ours)	2	×	-	88.1	87.5
	2	✓	-	88.6	88.2
	4	✓	-	89.4	<b>89.7</b>

# Classification examples











					
Ground Truth	4wd	building	bus	desk	monitor
Baseline	car	bus	car	sofa	chair
Ours	4wd	building	bus	desk	monitor
					
Ground Truth	table	chair	ute	toilet	bathtub
Baseline	nite-stnd	table	ute	toilet	table
Ours	table	chair	truck	chair	bed

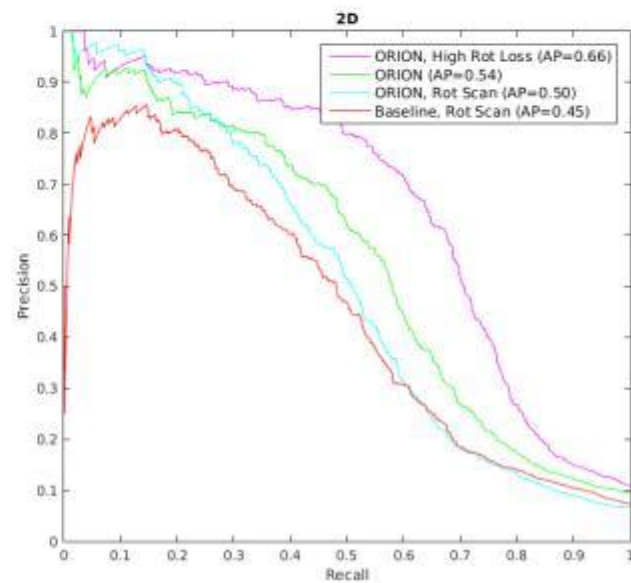
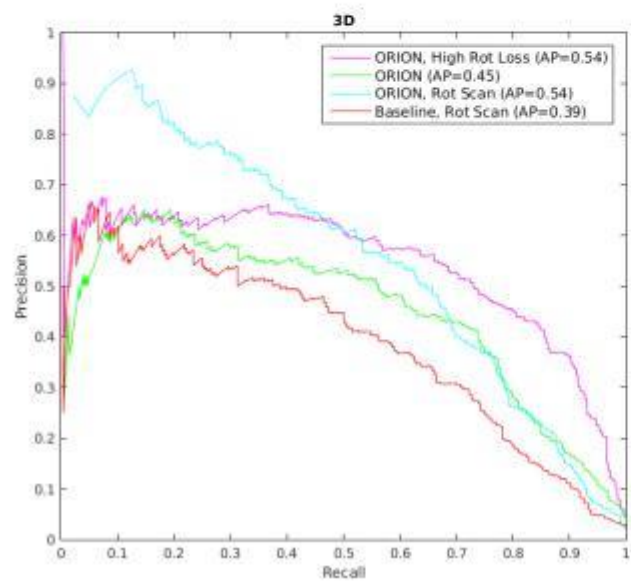
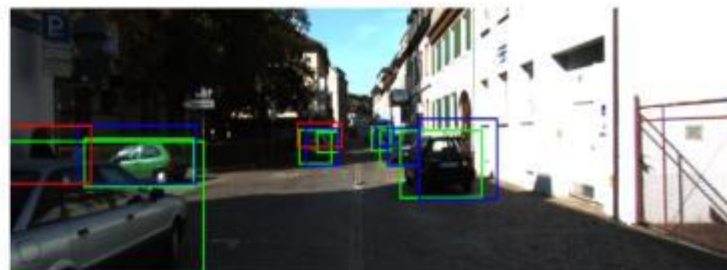
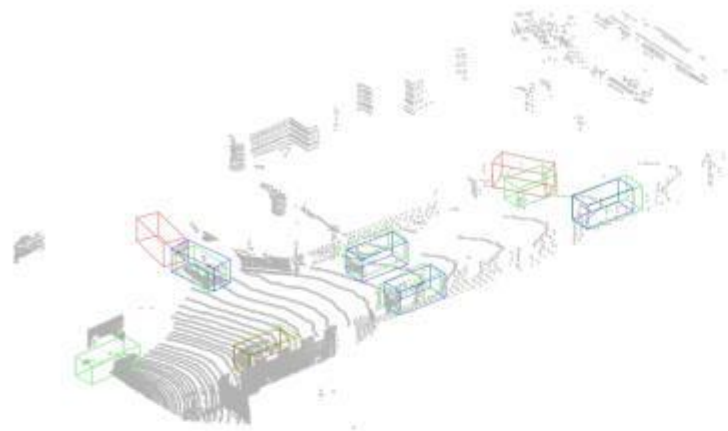
Figure 4: Some exemplar classification results. We show examples on which the outputs of the two networks differ.

# Results: Detection (KITTI)

- 3D detector to detect cars
- Orientation sensitive network => binary object classifier
- 3D point cloud (No RGB)



# Results: Detection (KITTI)



# Conclusion

Object orientation during training => better classification results at test time.

# Source

- <https://www.youtube.com/watch?v=a246GAffWZk> modelnet40 trained example