

2016年9月

# 智能驾驶与视觉感知

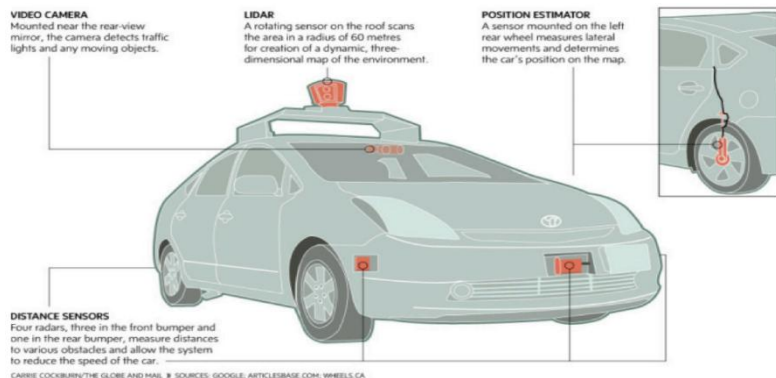
清华大学电子系 马惠敏



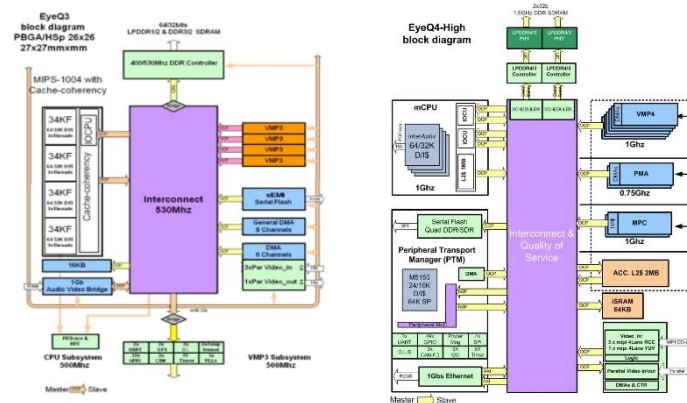
# 智能驾驶与视觉感知

## Vision Conception for Semi-Autonomous Driving

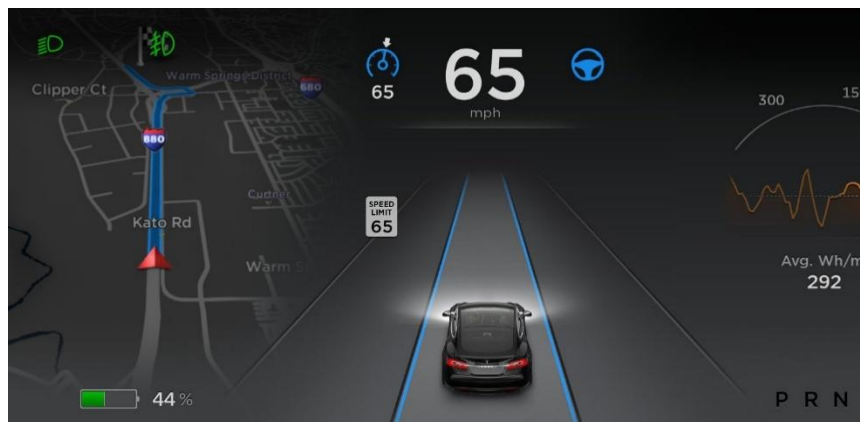
### Google



### Mobileye



### Tesla



### NVIDIA



# 智能驾驶与视觉感知

## Vision Conception for Semi-Autonomous Driving

### Google

- LIDAR
- Radar/ Sonar
- Stereo Camera
- Maps

### Mobileye

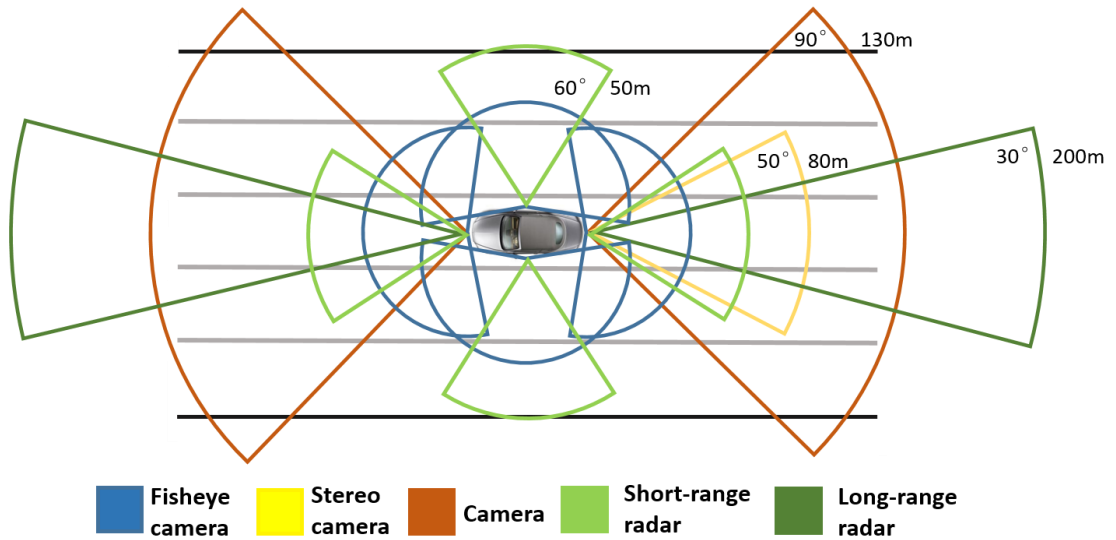
- EyeQ3 (Now)
- EyeQ4 (2018)  
8 cameras

### NVIDIA

- Drive PX: 2 Jetson TX1,  
12 cameras
- Drive PX2: 12 CPU cores,  
2 Pascal GPUs  
DriveNet

### Tesla

- Mobileye EyeQ3
- Forward Radar
- Forward Camera
- 360° Ultrasonic Sonar
- GPS



# 智能驾驶与视觉感知

## Vision Conception for Semi-Autonomous Driving

### □ 功能/Functions

- 前向碰撞预警  
Forward Collision Warning
- 自适应巡航控制  
Adaptive Cruise Control
- 自动变道  
Automatic Lane Changing
- 交通拥堵辅助  
Traffic Jam Assistance
- 自动泊车  
Autonomous parking

### □ 技术/Techniques

- 车道检测  
Line Detection
- 交通标识/灯检测  
Traffic Sign/Light Detection
- 驾驶道路标识  
Drivable Space Labeling
- 通用物体检测  
General Object Detection
- 整体路径规划  
Holistic Path Planning





# 智能驾驶与视觉感知

Vision Conception for Semi-Autonomous Driving

核心问题：小目标、强遮挡、高动态

Small objects, Heavy Occlusion, Rapid moving





# 智能

## 物体检测

### □ ImageNet:

- 1000 classes, ~2M images



### □ PASCAL VOC:

- 20 classes, ~20K images



Professional photographs, simple scenes, large objects recall > **95%** (1K Prop.)

### □ KITTI:

- 3 classes, ~20K images



recall < **75%**

### □ COCO:

- 80 classes, ~200K images



recall < **40%**

微软152层**50%**

smaller objects, heavy occlusion, high resolution

# 智能 +

任务关联实验范式

1. 汽车检测
2. 行人检测

.....



提取人的眼动注意规律

建立注意关联显著性模型

挖掘任务关联图像检测识别规律

建立图像认知网络模型

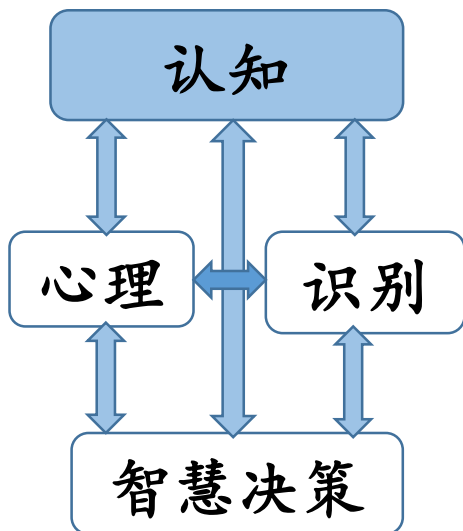


Overview

Challenges

测试验证





## ➤ 让机器学习人的思考模式

- 显著性物体检测：语义注意认知模型
- 部件与结构认知模型：抵抗遮挡能力
- 3D 场景物体识别：适应复杂环境
- 图像认知心理学：心理特征量化提取

## ➤ 代表性文章与专利

**PRL** : Geodesic Weighted Bayesian Model for Saliency Optimization, 2016

**PR** : Manifold topological multi-resolution analysis method, 2011

**CVPR**: Monocular 3D Object Detection for Autonomous Driving, 2016

**CVPR**: Improving Object Proposals with Multi- Thresholding Straddling Expansion, 2015

**NIPS** : 3D Object Proposals for Accurate Object Class Detection. 2015

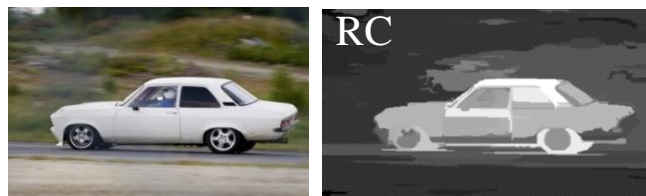
专利：201310221563.9 一种图像认知心理分析系统

专利：201310222323.0 基于MMPI心理量表的图像库及其构建方法

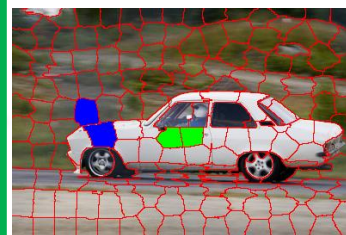


# (1) 显著性物体检测

- 提出自顶向下的认知模型
  - 利用贝叶斯框架抑制噪声
  - 利用区域空间关系统一标出整个显著目标



图像的Saliency map:

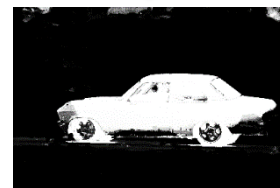


Geodesic Similarity

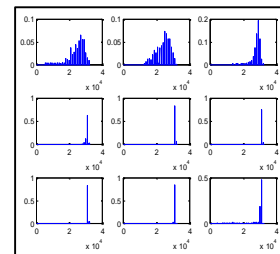


original saliency map

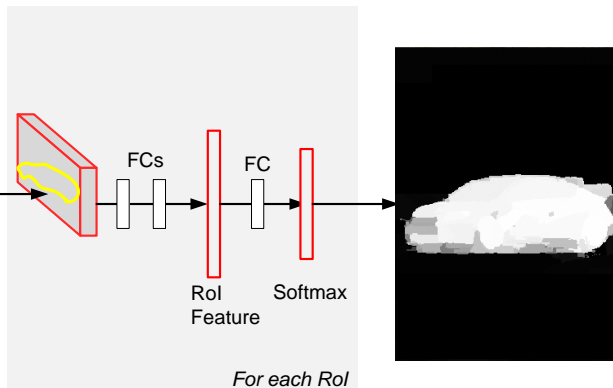
Bayesian inference



Initial salient regions



v: color, texture



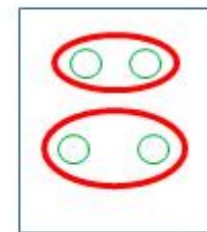
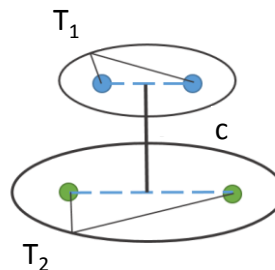
# (1) 显著性物体检测



## (2) 部件与结构认知模型

### ● 提出通用对称对结构模型

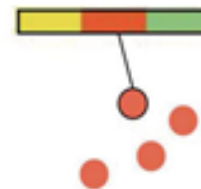
- 建立了动作图象中的对称对结构模型
- 降低了结构对部件准确检测的要求



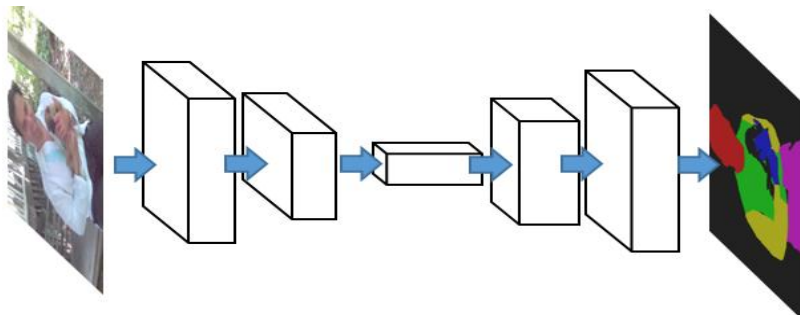
对称对结构模型

### ● 提出部件的Top-down空间结构模型

- 确定部件的空间关系提高动作检测准确性
- 建立空间结构神经网络和分级分辨神经网络



动作部件空间关系



用全卷积网络确定并定位部件



Human-level  
concept learning



### (3) 3D 场景物体识别

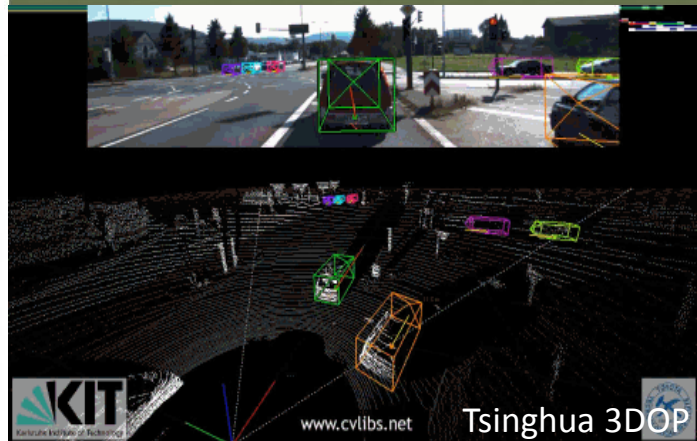
- 提出三维环境认知学习模型

- 建立人在物体检测中的认知模型
- 将三维环境先验与深度学习结合
- 提出3D物体检测、定位方法

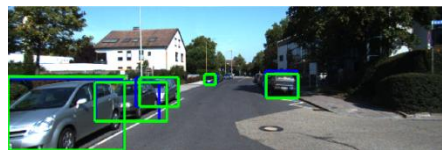
The KITTI Vision Benchmark Suite  
A project of Karlsruhe Institute of Technology and Toyota Technological Institute at Chicago



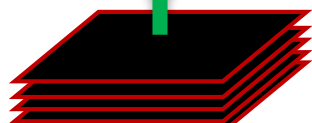
home setup stereo flow scene flow odometry object tracking road semantics raw data submit results jobs



小目标、强遮挡、姿态变化



Joint box regression and orientation estimation



CNN (VGG)

Project to 2D



3D object proposals

三维环境模型:

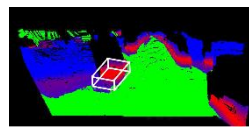
- 高度先验
- 高度对比度
- 点云密度
- 自由空间



3D Proposal

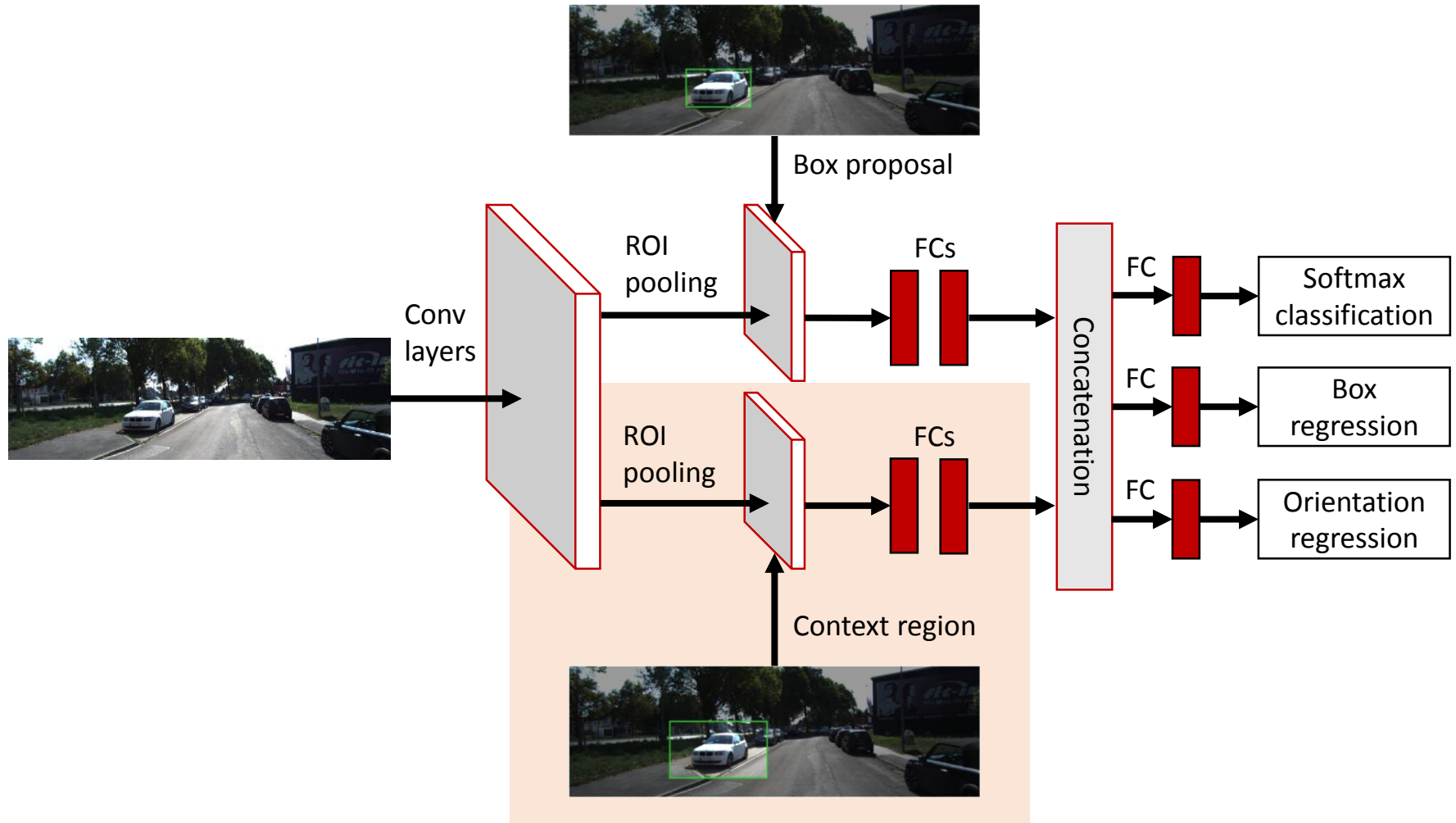


Free Space



Height prior

# 3D Object Detection Network

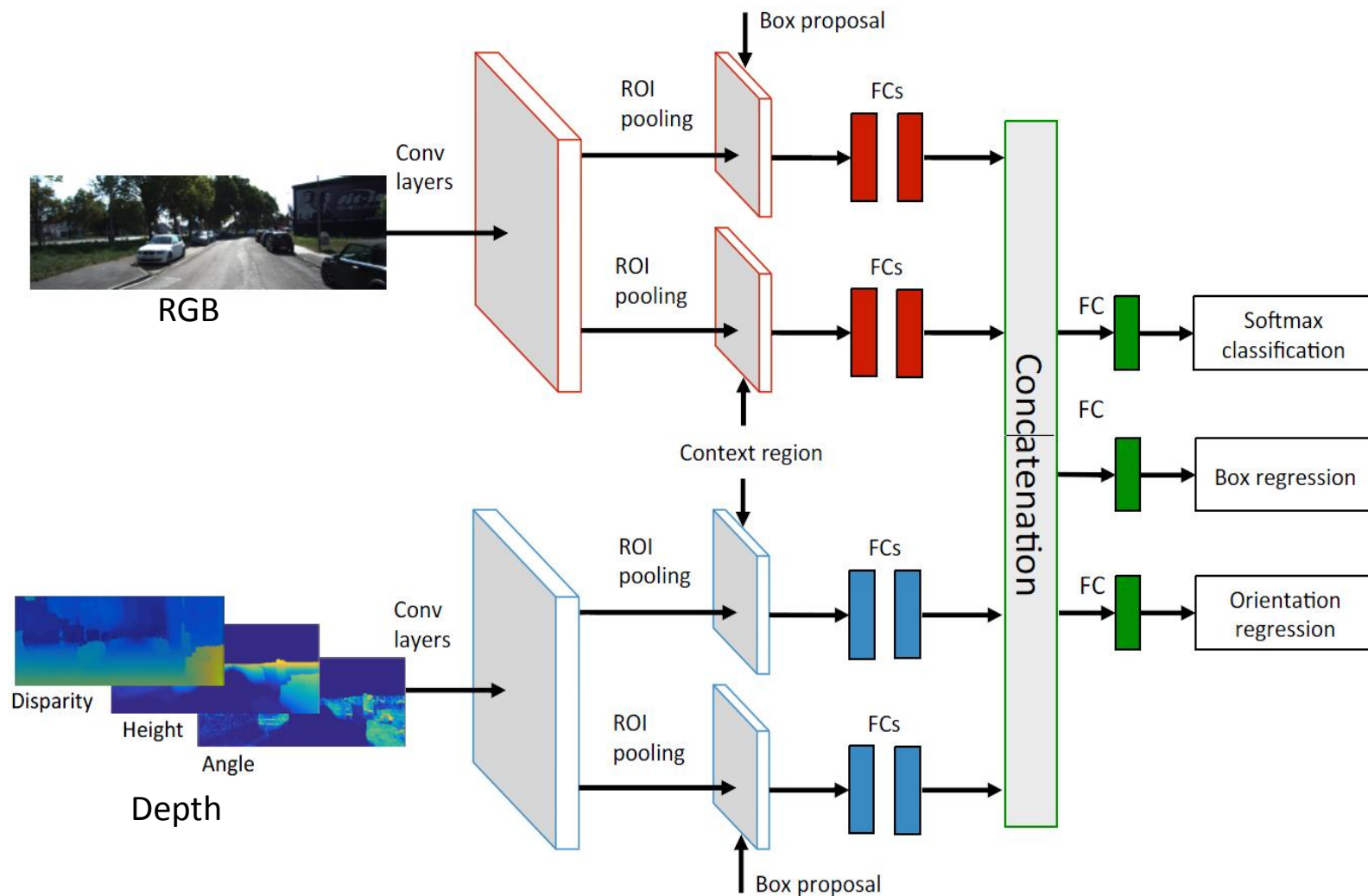


[Chen & Ma, et al. NIPS'15 & CVPR'16]

# 3D Object Detection Network

[Chen & Ma, et al. NIPS'15 & CVPR'16]

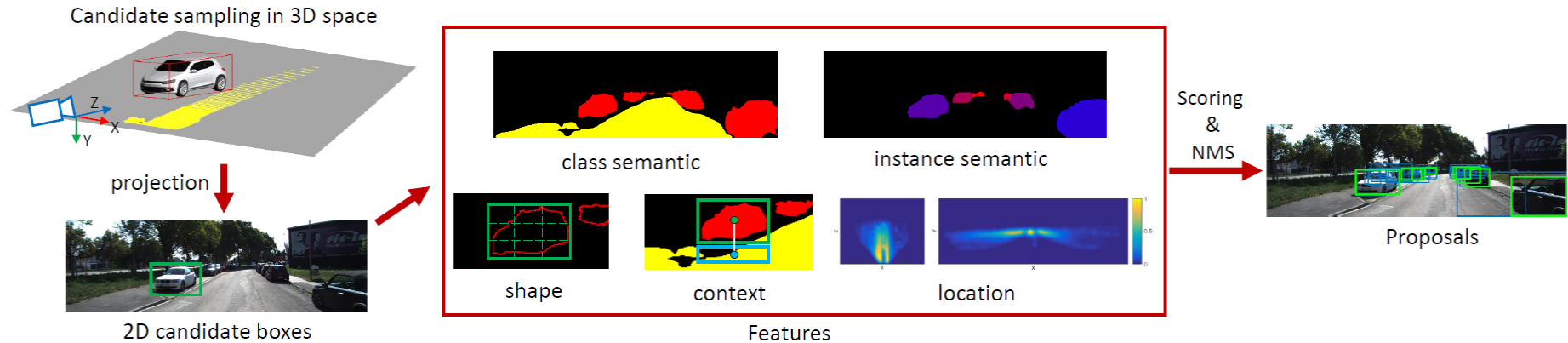
- 包含上下文信息: Incorporating context information
- 多任务监督: Multi-task supervision
- 多通道特征学习: Multi-stream feature learning





# 单目三维类目标检测

## Monocular 3D Object Proposals - Overview



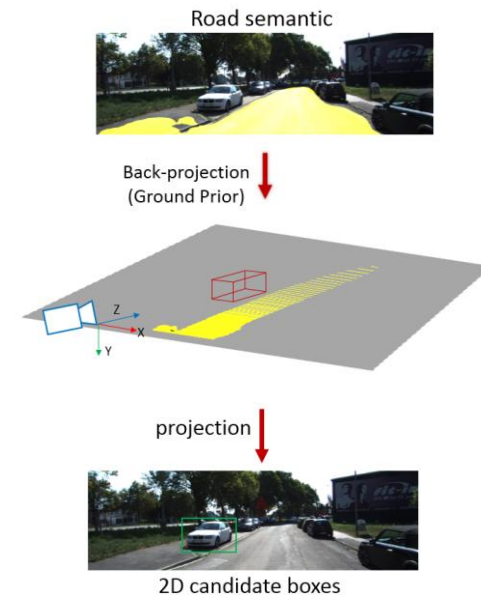
### ❑ 立体/Stereo

- 3D Sampling
- Road Estimation from 3D
- Point Cloud Features
- Exhaustive Search
- Structured SVM



### ❑ 单目/Monocular


- 3D Sampling
- Road Estimation from 2D
- Semantic Features
- Exhaustive Search
- Structured SVM




# Results: Object Detection and Orientation Estimation

## ➤ Object detection (AP)


### Car

	Method	Setting	Code	Moderate	Easy	Hard
1	<a href="#">DenseBox2</a>			89.32 %	93.94 %	79.81 %
2	<a href="#">DJML</a>			88.79 %	91.31 %	77.73 %
3	<a href="#">3DOP</a>			88.64 %	93.04 %	79.10 %
X. Chen, K. Kundu, Y. Zhu, A. Berneshawi, H. Ma, S. Fidler and R. Urtasun: <a href="#">3D Object Proposals for</a>						
4	<a href="#">SubCNN</a>			88.55 %	90.74 %	77.95 %
Anonymous submission						
5	<a href="#">CVPR #1408</a>			88.09 %	90.11 %	78.32 %

### Pedestrian


	Method	Setting	Code	Moderate	Easy	Hard
1	<a href="#">3DOP</a>			67.47 %	81.78 %	64.70 %
X. Chen, K. Kundu, Y. Zhu, A. Berneshawi, H. Ma, S. Fidler and R. Urtasun: <a href="#">3D Object Proposals for</a>						
2	<a href="#">CVPR #1408</a>			66.34 %	80.25 %	63.41 %
Anonymous submission						
3	<a href="#">SubCNN</a>			66.13 %	79.13 %	61.27 %

### Cyclist


	Method	Setting	Code	Moderate	Easy	Hard
1	<a href="#">3DOP</a>			68.94 %	78.39 %	61.37 %
X. Chen, K. Kundu, Y. Zhu, A. Berneshawi, H. Ma, S. Fidler and R. Urtasun: <a href="#">3D Object Proposals for</a>						
2	<a href="#">CVPR #1408</a>			67.03 %	77.06 %	59.87 %
Anonymous submission						
3	<a href="#">SubCNN</a>			61.98 %	74.40 %	54.75 %

## ➤ Object detection and orientation estimation (AOS)


### Car

	Method	Setting	Code	Moderate	Easy	Hard
1	<a href="#">SubCNN</a>			87.88 %	90.49 %	77.10 %
Anonymous submission						
2	<a href="#">DJML</a>			87.51 %	90.67 %	76.33 %
3	<a href="#">3DOP</a>			86.10 %	91.44 %	76.52 %
X. Chen, K. Kundu, Y. Zhu, A. Berneshawi, H. Ma, S. Fidler and R. Urtasun: <a href="#">3D Object F</a>						
4	<a href="#">CVPR #1408</a>			85.66 %	88.31 %	75.89 %

### Pedestrian

	Method	Setting	Code	Moderate	Easy	Hard
1	<a href="#">3DOP</a>			59.80 %	72.94 %	57.03 %
X. Chen, K. Kundu, Y. Zhu, A. Berneshawi, H. Ma, S. Fidler and R. Urtasun: <a href="#">3D Objec</a>						
2	<a href="#">SubCNN</a>			59.40 %	72.61 %	54.78 %
Anonymous submission						
3	<a href="#">CVPR #1408</a>			57.86 %	71.13 %	54.94 %

### Cyclist

	Method	Setting	Code	Moderate	Easy	Hard
1	<a href="#">3DOP</a>			58.68 %	70.13 %	52.35 %
X. Chen, K. Kundu, Y. Zhu, A. Berneshawi, H. Ma, S. Fidler and R. Urtasun: <a href="#">3D Objec</a>						
2	<a href="#">CVPR #1408</a>			55.73 %	66.64 %	49.75 %
Anonymous submission						
3	<a href="#">SubCNN</a>			52.06 %	63.74 %	45.93 %

Car	Object detection (AP)			Object detection and orientation estimation (AOS)		
Method	Easy	Moderate	Hard	Easy	Moderate	Hard
SubCat [1]	84.14	75.46	59.71	74.42	83.41	58.83
3DVP [2]	87.46	75.77	65.38	86.92	74.59	64.11
AOG [3]	84.80	75.94	60.70	-	-	-
Regionlets [4]	84.75	76.45	59.70	-	-	-
spLBP [5]	87.19	77.40	60.60	-	-	-
Faster R-CNN [6]	86.71	81.84	71.12	-	-	-
<b>3DOP (Ours)</b>	<b>93.04</b>	88.64	<b>79.10</b>	<b>91.44</b>	86.10	76.52
<b>Mono3D (Ours)</b>	92.33	<b>88.66</b>	78.96	91.01	<b>86.62</b>	<b>76.84</b>

Pedestrian	Object detection (AP)			Object detection and orientation estimation (AOS)		
Method	Easy	Moderate	Hard	Easy	Moderate	Hard
DPM-VOC+VP [1]	59.48	44.86	40.37	53.55	39.83	35.73
FilteredICF [2]	67.65	56.75	51.12	-	-	-
DeepParts [3]	70.49	58.67	52.78	-	-	-
CompACT-Deep [4]	70.69	58.74	52.71	-	-	-
Regionlets [5]	73.14	61.15	55.21	-	-	-
Faster R-CNN [6]	78.86	65.90	61.18	-	-	-
<b>Mono3D (Ours)</b>	80.35	66.68	63.44	71.15	58.15	54.94
<b>3DOP (Ours)</b>	<b>81.78</b>	<b>67.47</b>	<b>64.70</b>	<b>72.94</b>	<b>59.80</b>	<b>57.03</b>

Cyclist	Object detection (AP)			Object detection and orientation estimation (AOS)		
Method	Easy	Moderate	Hard	Easy	Moderate	Hard
DPM-VOC+VP [1]	42.43	31.08	28.23	30.52	23.17	21.58
<b>Mono3D (Ours)</b>	76.04	66.36	58.87	65.56	54.97	48.77
<b>3DOP (Ours)</b>	<b>78.39</b>	<b>68.94</b>	<b>61.37</b>	<b>70.13</b>	<b>58.68</b>	<b>52.35</b>



# 复杂干扰环境中的三维目标识别

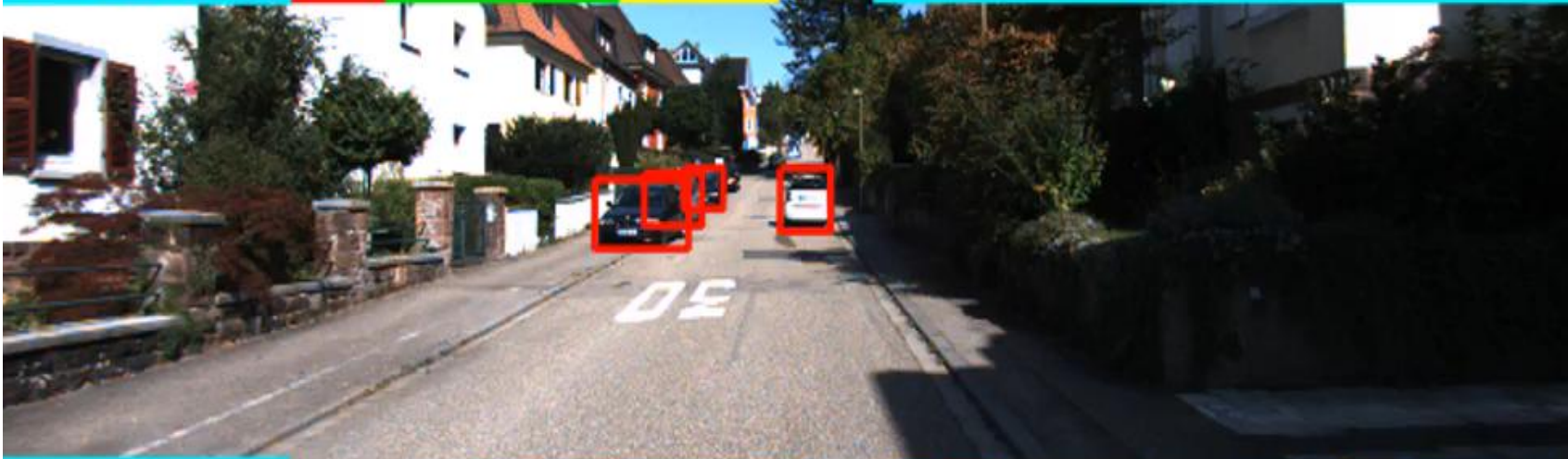
2D Detections

Car

Pedestrian

Cyclist

Note: no temporal information used



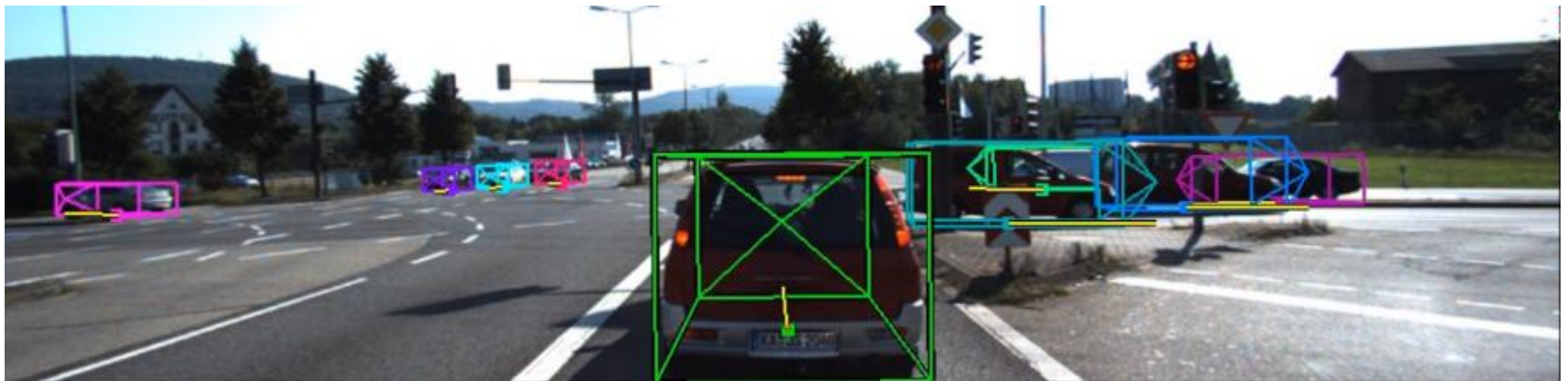
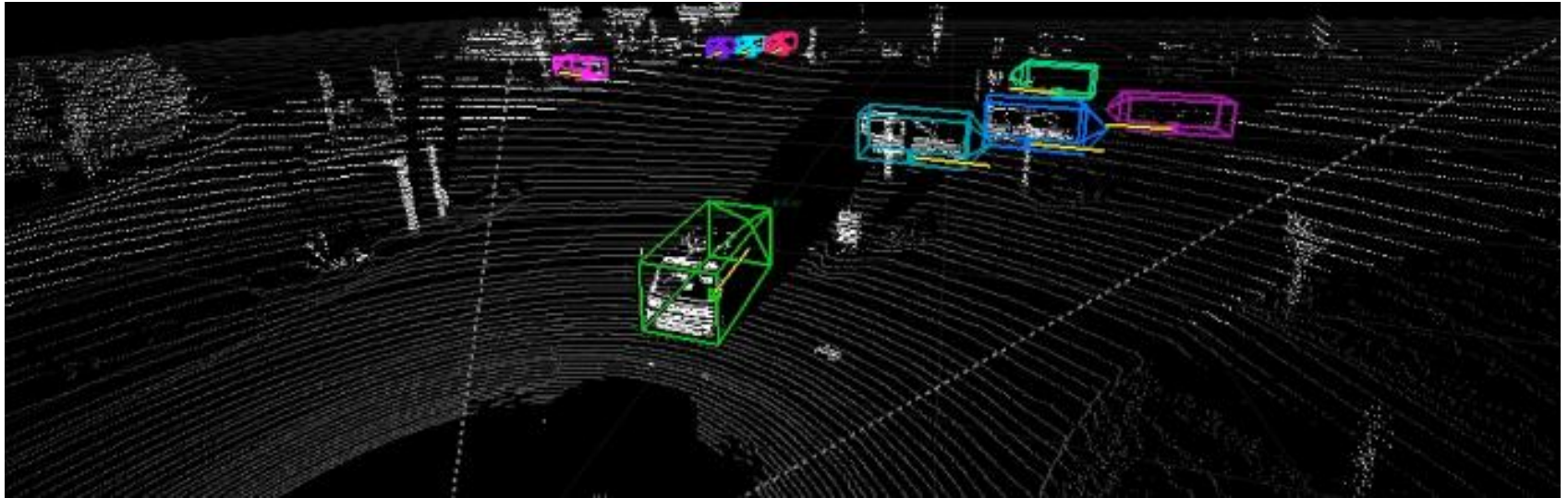
3D Detections



# 智能 ++ 高级认知：多模型&多任务&多视角

Multi-Modal & Multi-Task & Multi-View

- Multi-Modal





# 高级认知：多模型&多任务&多视角

Multi-Modal & Multi-Task & Multi-View

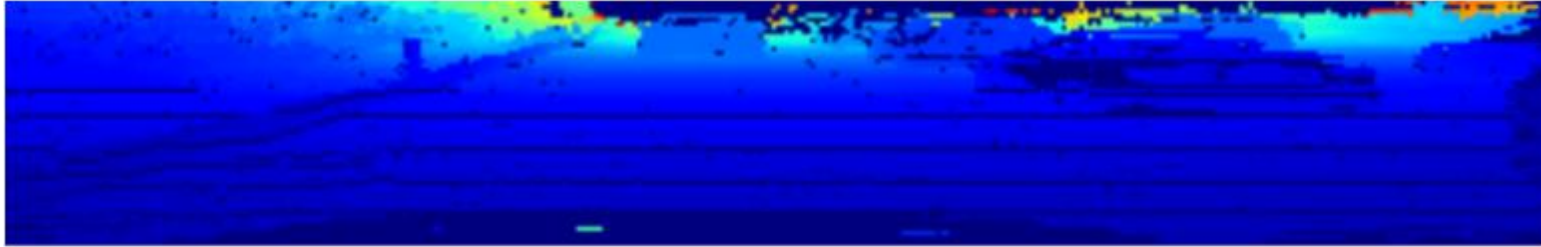
- **Multi-Task**



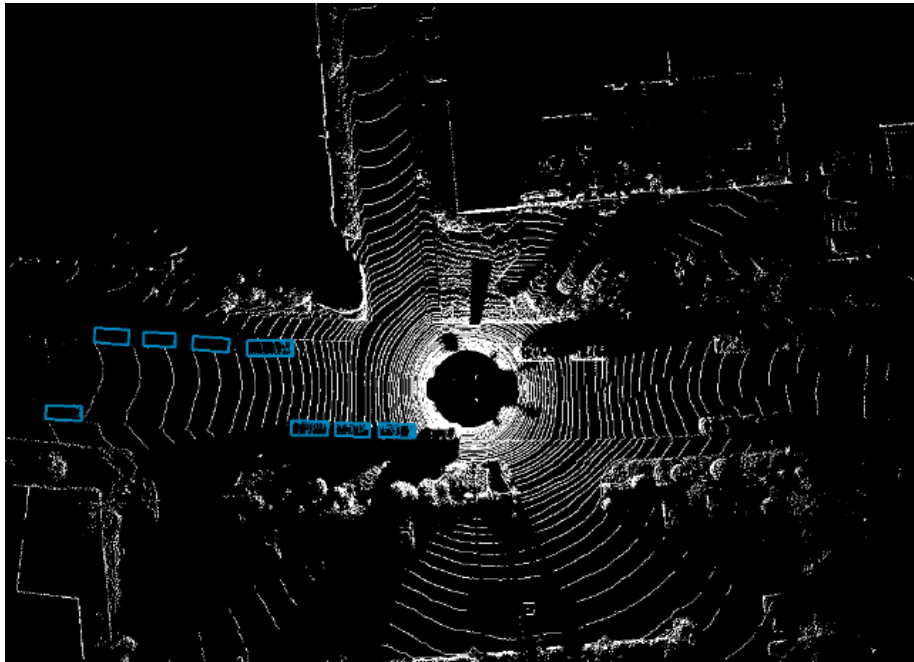


# 高级认知：多模型&多任务&多视角

- Front view
- **LIDAR**: Input: depth, height, distance
  - **RGB**



Bird's eye view

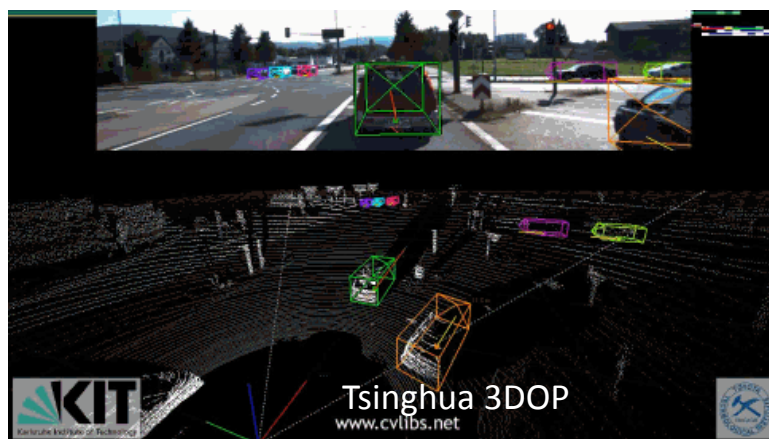


- Input representation
- Sparse points
- Extra small objects
- Oriented bounding boxes

# 应用：智能驾驶与视觉感知

➤ 小目标 ➤ 强遮挡 ➤ 任意姿态 ➤ 人

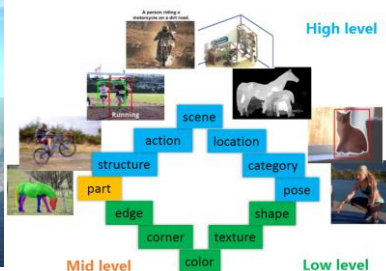
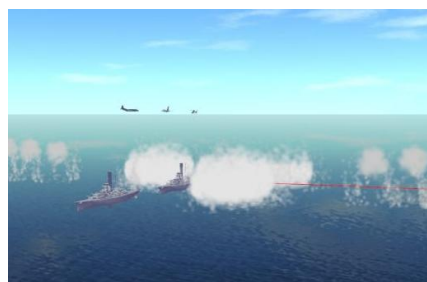
- 建立了国际上唯一的图像认知心理评测系统
- 2015自动驾驶KITTI国际评测，共六项获四项世界第一
- 复杂环境目标检测识别技术服务于航空、航天、交通



**The KITTI Vision Benchmark Suite**  
A project of Karlsruhe Institute of Technology and Toyota Technological Institute at Chicago

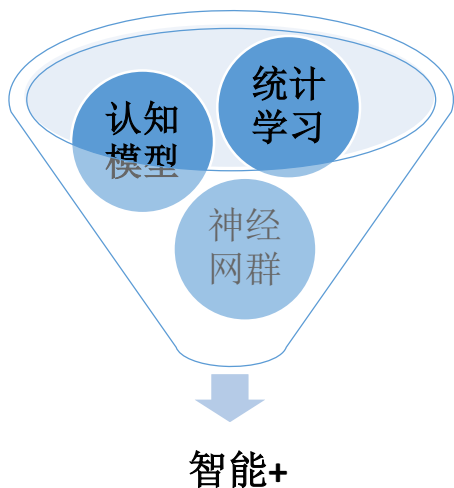


- 国家自然科学基金：基于图像认知的心理特征提取与分析
- 专项重点基金：适应复杂环境的成像导航
- 国家重点研发计划：智能电动汽车感知、决策与控制





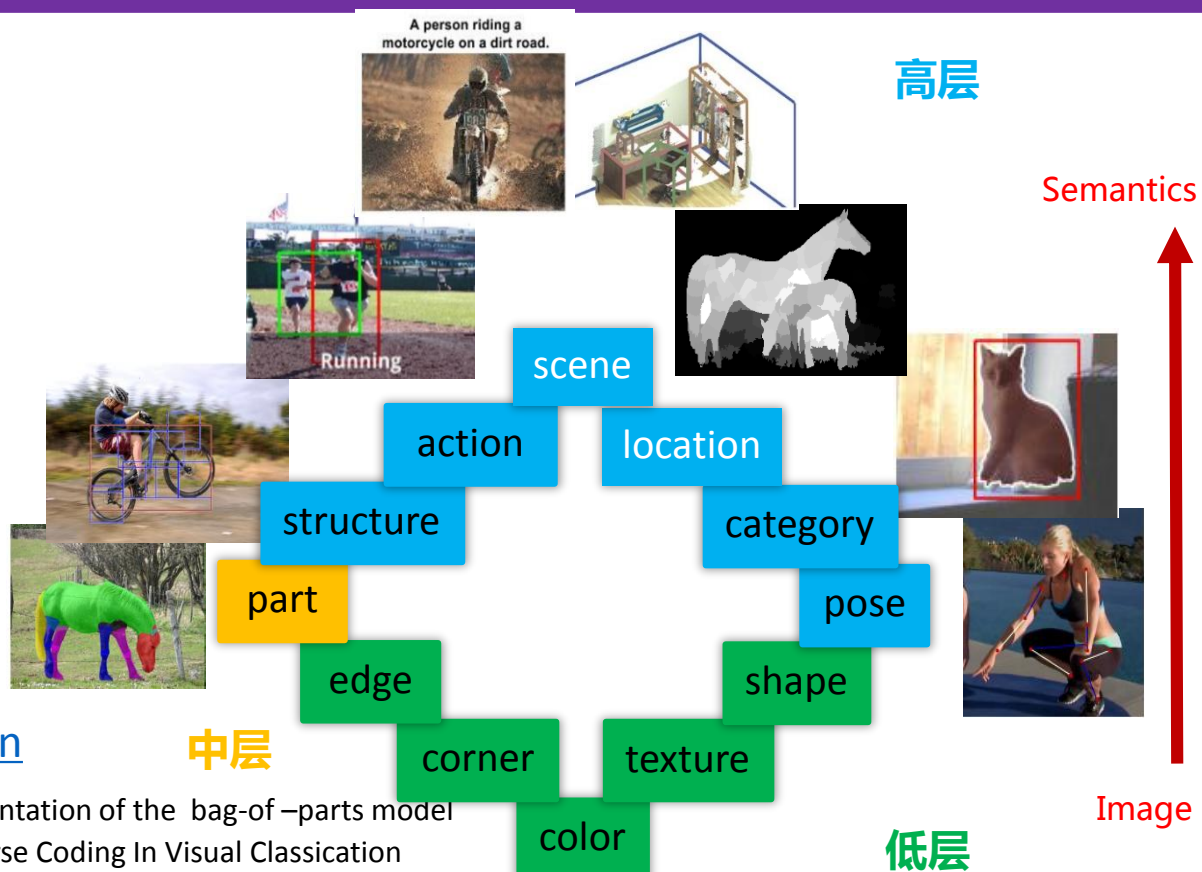
# 智能+



发表的相关文章:

<http://3dimage.ee.Tsinghua.edu.cn>

- ICIP 2014: Learning a compact latent representation of the bag-of-parts model
- ICCVG 2014: Protected Pooling Method Of Sparse Coding In Visual Classification
- ICIP 2015: Geodesic weighted Bayesian model for salient Object Detection.
- CVPR 2015: Improving Object Proposals with Multi-Thresholding Straddling Expansion
- NIPS 2015: 3D Object Proposals for Accurate Object Class Detection
- CVPR 2016: Monocular 3D Object Detection for Autonomous Driving
- ICIP 2016: Object Detection Via Fast R-CNN And Low-level Cues
- ICIP 2016: Region Candidate Combination for Action Recognition.
- PRL 2016: Geodesic Weighted Bayesian Model for Saliency Optimization
- PRL 2016: Semantic parts based top-down pyramid for action recognition
- PR 2016: Generalized Symmetric Pair Model for Image Recognition (修改中)



# Thank You

<http://3dimage.ee.Tsinghua.edu.cn>



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Electronic Engineering



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