

Al未来之光

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北京理工大学自动化学院研究生会 北京邮电大学自动化学院研究生会 中国科学院计算技术研究所研究生会



Deep Fitting Degree Scoring Network for Monocular 3D Object Detection

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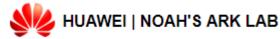














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01

Problem

Monocular 3D Object Detection



"Perception is our best guess as to what is in the world given our current sensory input and our prior experience"

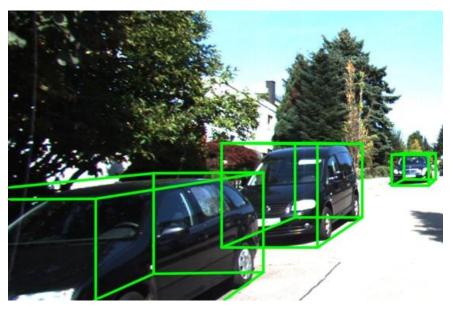


- Helmholtz (1866)

Problem Setting: Monocular 3D Object Detection

- ➤ Input: A single RGB image & camera intrinsic
- > Solve: 9 DoF, including orientation, dimension, location
- Cue: Appearance (sensory input) & Projection Law (prior experience)

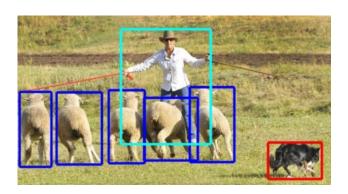


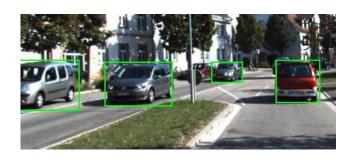




Why 3D perception

- > 3D perception is the key to human intelligence
- > Autonomous driving & Robotic grasping



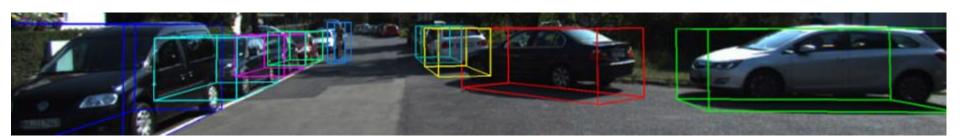






Challenges in Location Estimation

- Dimension and orientation estimation are easier than location estimation
- > Ambiguities arising from 2D-3D mapping
- > Real 3D information unavailable
- Occlusions, Truncation, Scale variation......





02

Motivation

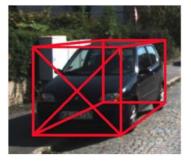
How we come up with our idea



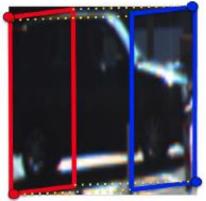
Previous Methods

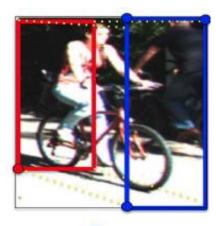
- ➤ Tight Constraints (CVPR17)
- Solves the location by placing the 3D proposal in the 2D bounding box compactly.













Drawbacks:

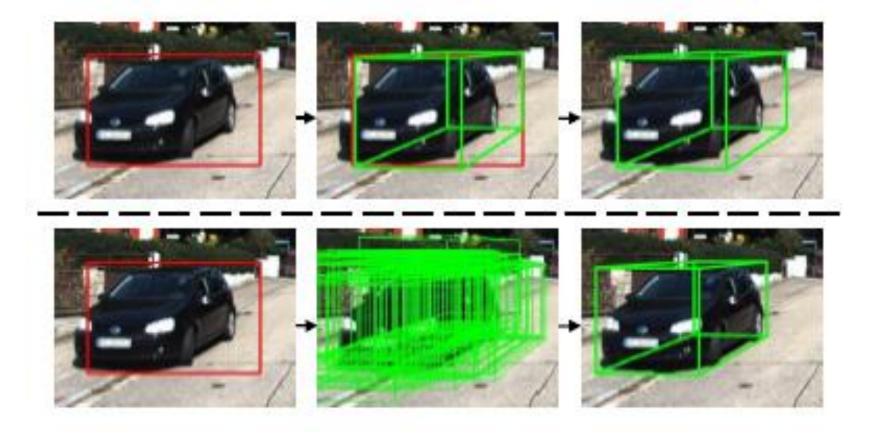
- 1) Image appearance clue is not used
- 2) Performance highly depends on the 2D detection accuracy





How do human do?

➤ Hypothesize-and-Verify





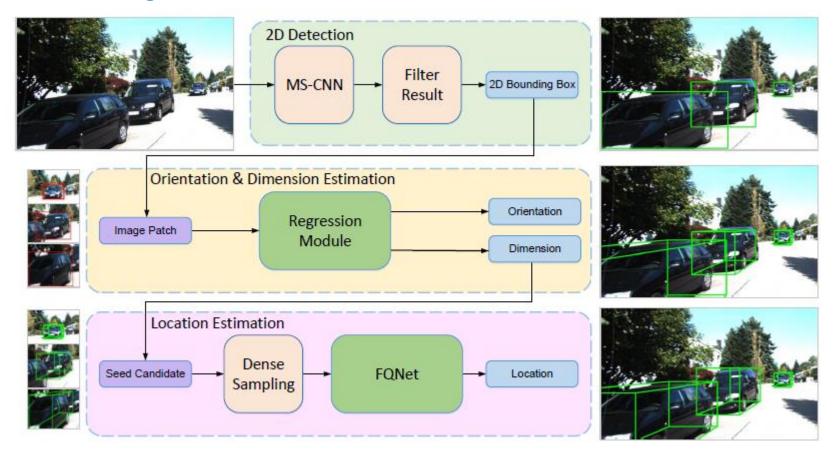
03

Approach

Some details

Overall framework

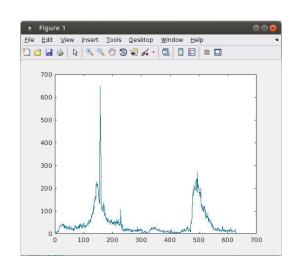
> Three-stage

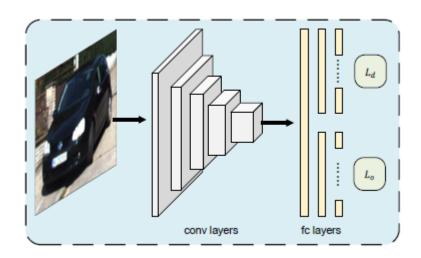




Regression Module

> Anchor cuboid & Anchor angle





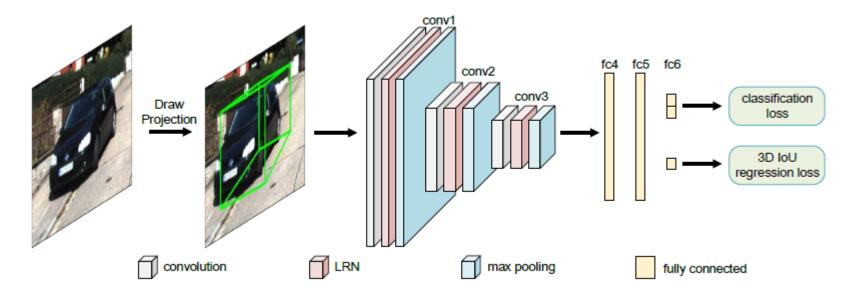
$$L_d = -\log \sigma(c_{i^*}) + [1 - IoU(\boldsymbol{A}_{i^*} + [\Delta w_{i^*}, \Delta h_{i^*}, \Delta h_{i^*}, \Delta l_{i^*}], \boldsymbol{G})]$$

$$L_o = -\log \sigma(c_{i^*}) + [1 - \cos(\Theta_{i^*} + \Delta\theta_{i^*} - \theta_G)]$$



Location Estimation

- Dense sampling
- > FQNet



$$\Theta^{\star} = \arg\min_{\Theta} ||\mathcal{F}(\boldsymbol{I}, \boldsymbol{S}_{i}|\Theta) - IoU(\boldsymbol{I}, \boldsymbol{S}_{i})||$$



04

Experiments

Some demonstration



Experimental Results on KITTI dataset

> Orientation & Dimension

Table 1. Comparisons of the Average Orientation Similarity (AOS) with the state-of-the-art methods on the KITTI dataset.

Method		Easy]	Moderate		Hard			
Wiethod	train/val 1	train/val 2	test	train/val 1	train/val 2	test	train/val 1	train/val 2	test	
3DOP [9]	91.58	-	91.44	85.80	-	86.10	76.80	-	76.52	
Mono3D [8]	91.90	-	91.01	86.28	-	86.62	77.09	-	76.84	
3DVP [42]	-	78.99	86.92	-	65.73	74.59	-	54.67	64.11	
SubCNN [43]	-	94.55	90.67	-	85.03	88.62	-	72.21	78.68	
Deep3DBox [31]	-	97.50	92.90	-	96.30	88.75	-	80.40	76.76	
3D-RCNN [23]	90.70	97.70	89.98	89.10	96.50	89.25	79.50	80.70	80.07	
Our Method	97.28	97.57	92.58	93.70	96.70	88.72	79.25	80.45	76.85	

Method	train/val 1	train/val 2
3DOP [9]	0.3527	-
Mono3D [8]	0.4251	-
Deep3DBox [31]	-	0.1934
Our Method	0.1698	0.1465

Experimental Results on KITTI dataset

Location

Table 2. Comparisons of the 2D AP with the state-of-the-art methods on the KITTI Birds Eyed View validation dataset.

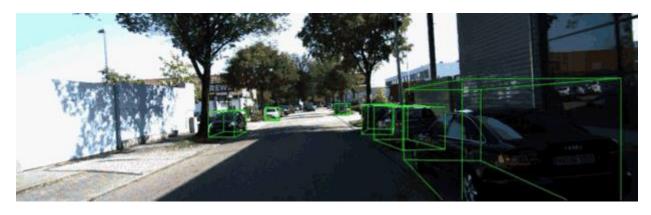
	IoU = 0.5							IoU = 0.7						
Method	Method Easy		Moderate		Hard		Easy		Moderate		Hard			
	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2		
3DOP [9]	55.04	-	41.25	-	34.55	-	12.63	-	9.49	-	7.59	-		
Mono3D [8]	30.50	-	22.39	-	19.16	-	5.22	-	5.19	-	4.13	-		
Deep3DBox [31]	-	30.02	-	23.77	-	18.83	-	9.99	-	7.71	-	5.30		
Our Method	32.57	33.37	24.60	26.29	21.25	21.57	9.50	10.45	8.02	8.59	7.71	7.43		

Table 4. Comparisons of the 3D AP with the state-of-the-art methods on the KITTI 3D Object validation dataset.

	IoU = 0.5							IoU = 0.7					
Method	Easy		Moderate		Hard		Easy		Moderate		Hard		
	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2	t/v 1	t/v 2	
3DOP [9]	46.04	-	34.63	-	30.09	-	6.55	-	5.07	-	4.10	-	
Mono3D [8]	25.19	-	18.20	-	15.52	-	2.53	-	2.31	-	2.31	-	
Deep3DBox [31]	-	27.04	-	20.55	-	15.88	-	5.85	-	4.10	-	3.84	
Our Method	28.16	28.98	21.02	20.71	19.91	18.59	5.98	5.45	5.50	5.11	4.75	4.45	



Qualitative Results

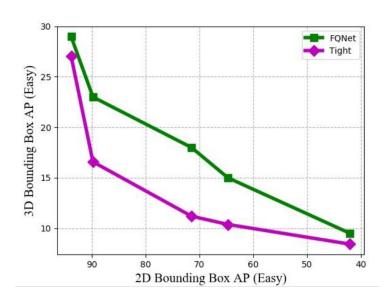


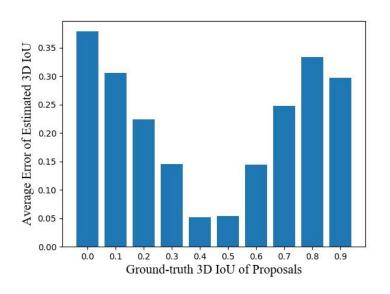




Effectiveness

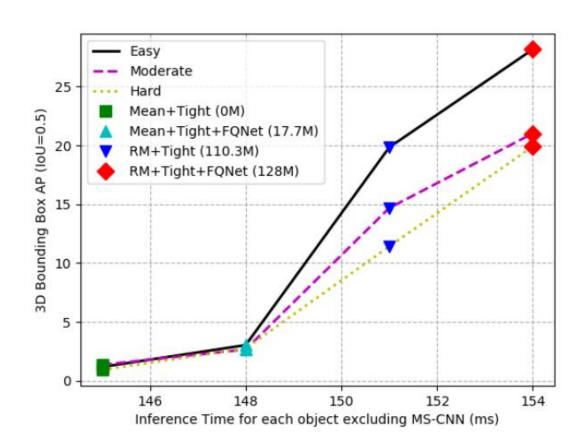
- ➤ Not sensitive to 2D detection precision
- ➤ 3D IoU regression





Accuracy vs Speed

- > Ablation study
- Efficiency



THANK YOU

