Paper Discussion

Joint 3D Proposal Generation and Object Detection from View Aggregation (AVOD)

2018-03-29



Input Image

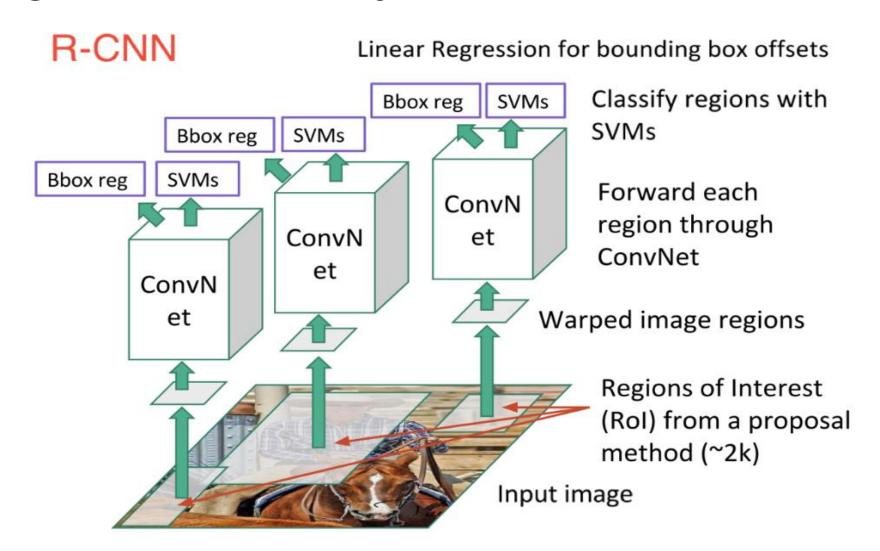
Question: Where are the cars in the image?

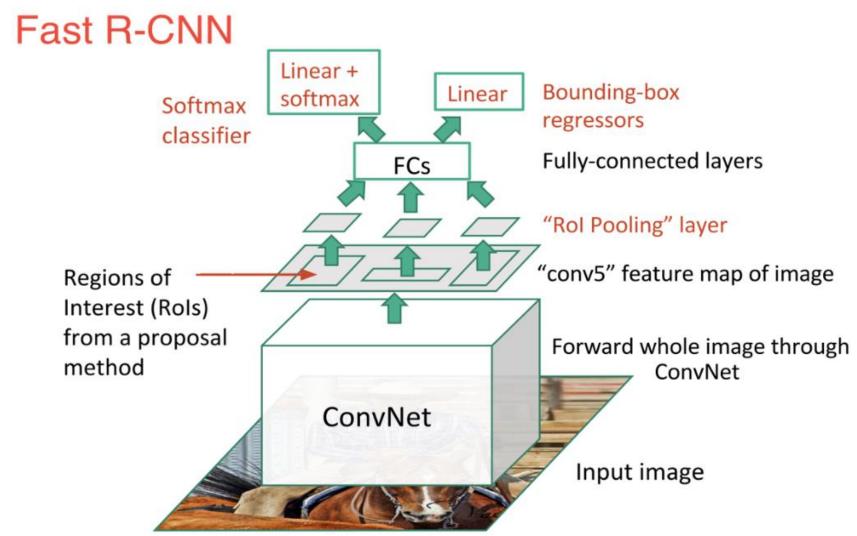
Answer:



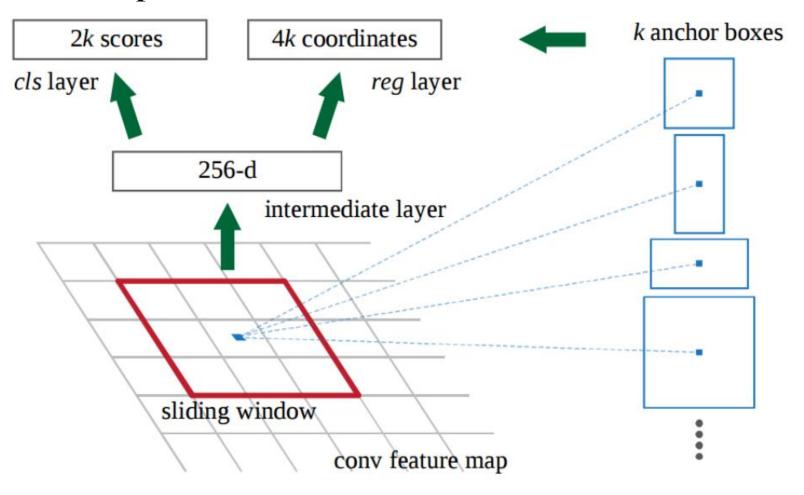
Object Detection Approach: Recognition + Localization

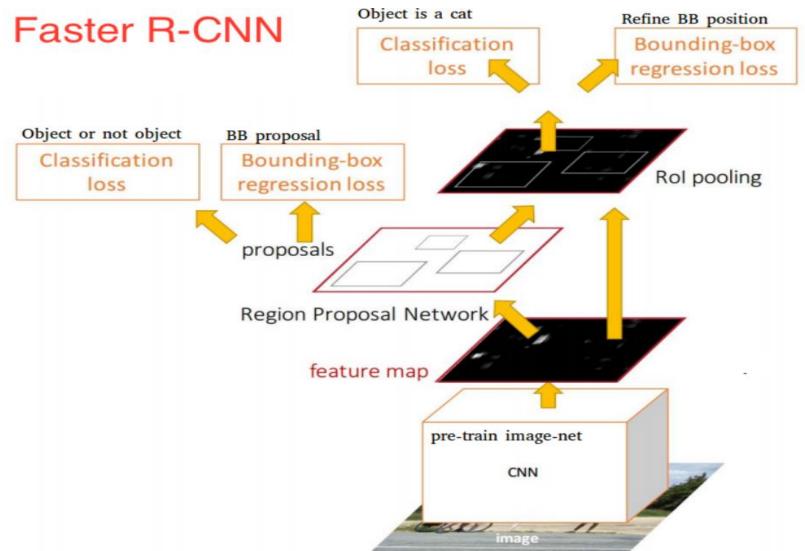
- Candidate Box Selection
- Feature Extraction
- Classification+ Bounding Box Regression

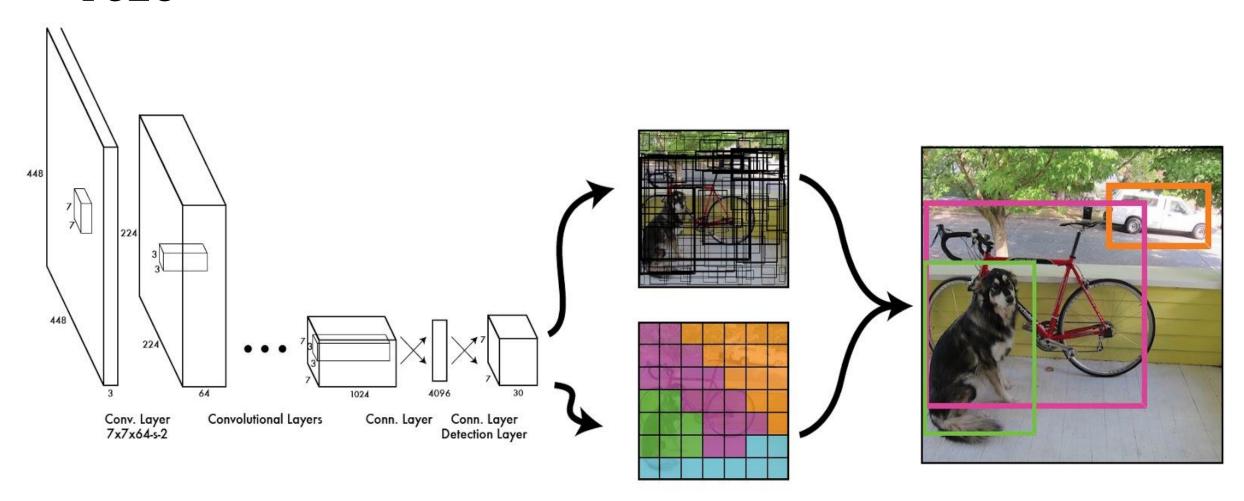




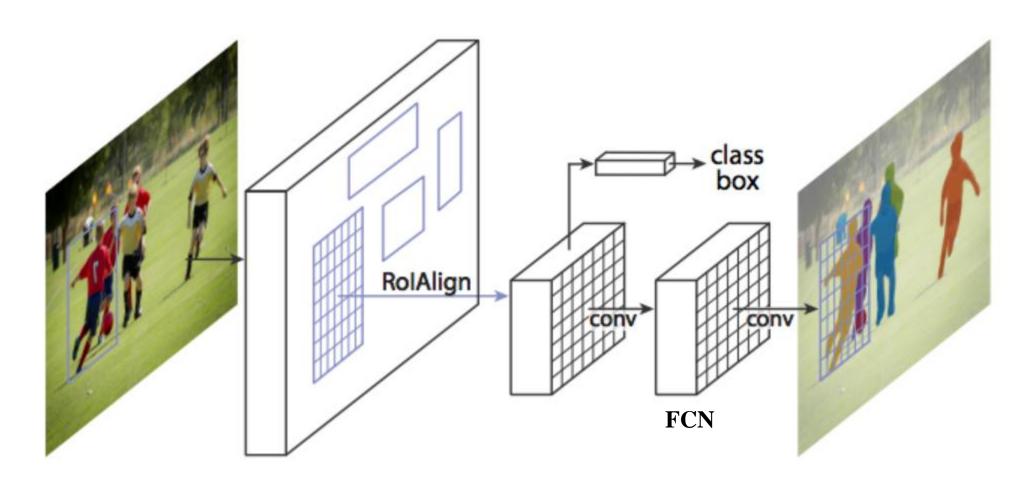
RPN: Regional Proposal Network







Mask R-CNN

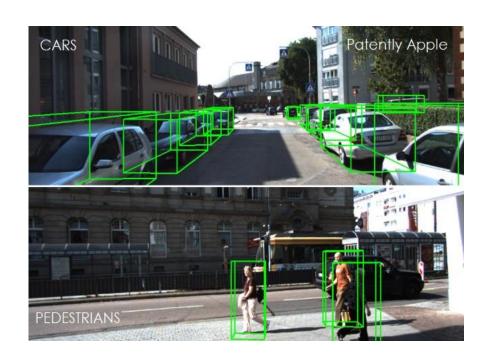


Problem Domain

• 3D object detection from images and point cloud for autonomous driving

• Deep neural networks

• KITTI 3D object detection benchmark



Challenges

 $2D \longrightarrow 3D$:

• Low resolution of the input data

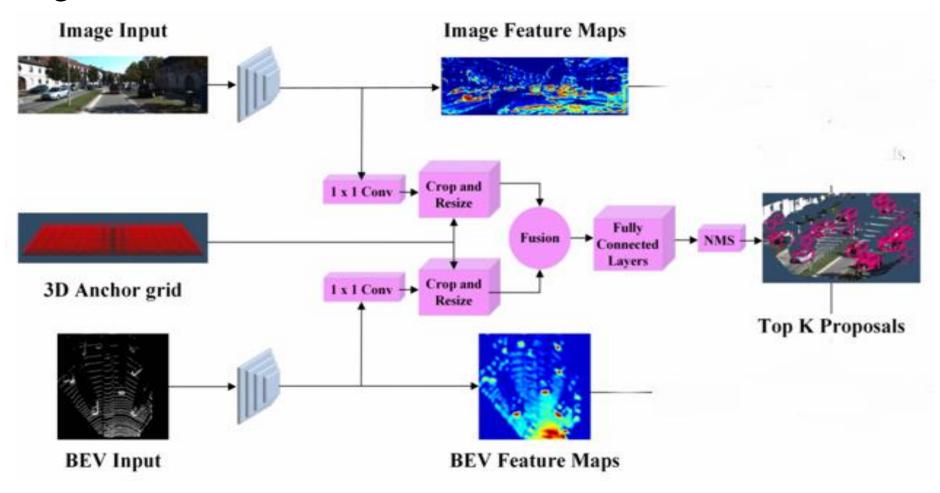
 Missed instances during region proposal generation cannot be recovered

Oriented bounding box estimation

Related Work

- Hand Crafted Feature For Proposal Generation
- Proposal Free Single Shot Detectors
- Monocular-Based Proposal Generation
- Monocular-Based 3D Object Detections
- 3D Region Proposal Networks (RPN)

First Stage Detection

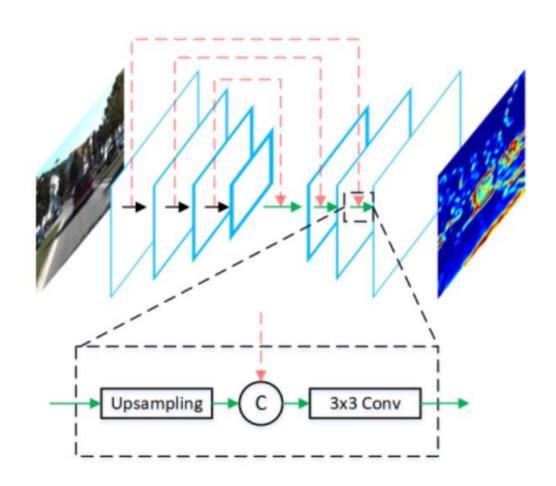


Feature Map Generating

• BEV images and RGB images

• Feature Pyramid Network (FPN)

Deconder (Upsampling)



3D Anchor Generation (t_x, t_y, t_z) , (d_x, d_y, d_z)

• 3D anchor box grid (80-100K anchors)

• Crop and resize from the feature maps (projection on BEV and RGB feature maps)

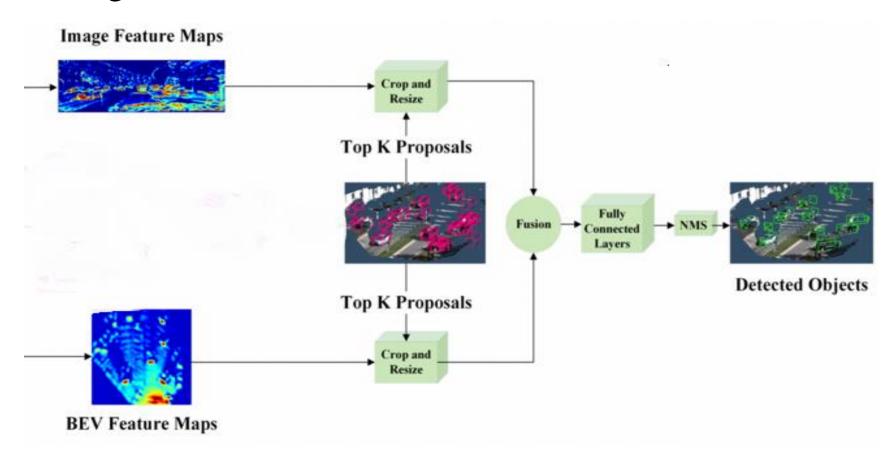
• Dimensionality Reduction Via 1*1 Convolutional Layers

3D Proposal Generation

• Two sets of feature crops are fused (element-wise mean)

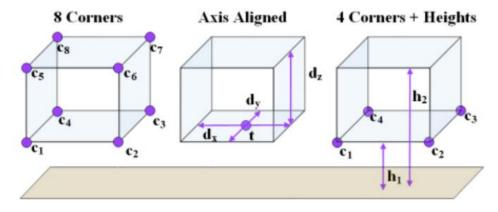
• Fed into fully collected layers to calculate object score and 3D box regression

Second Stage Detection



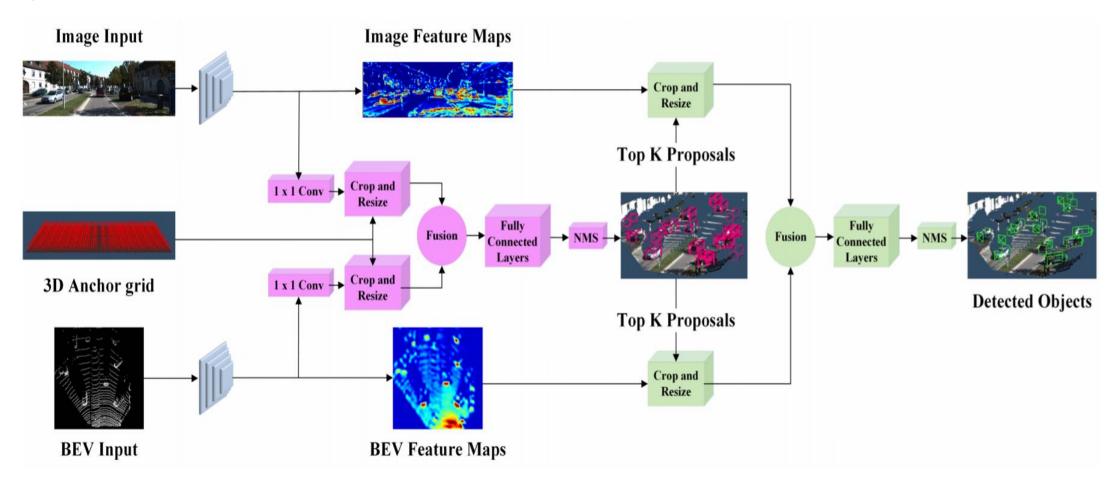
Second Stage Detection

• 3D Bounding Box Encoding (8 corners to 4 corners+ heights)



Orientation Vector Regression

Overview



Result

<u>Car</u>

	Method	Setting	Code	<u>Moderate</u>	Easy	Hard	Runtime
1	AVOD-FPN	*::	<u>code</u>	71.88 %	81.94 %	66.38 %	0.1 s
J. Ku,	, M. Mozifian, J. Lee, A	. Harakeh and	S. Wasland	ler: <u>Joint 3D Prop</u>	osal Generation	and Object Det	ection from Viev
2	F-PointNet	:::		70.39 %	81.20 %	62.19 %	0.17 s
3	DF-PC CNN	**		66.22 %	80.28 %	58.94 %	0.5 s
4	AVOD	**	<u>code</u>	65.78 %	73.59 %	58.38 %	0.08 s
J. Ku	, M. Mozifian, J. Lee, A	. Harakeh and	S. Wasland	ler: <u>Joint 3D Prop</u>	osal Generation	and Object Det	ection from Viev
5	<u>VxNet(LiDAR)</u>	:::		65.11 %	77.47 %	57.73 %	0.03 s
6	MV3D	***		62.35 %	71.09 %	55.12 %	0.36 s

<u>Cyclist</u>

	Method	Setting	Code	<u>Moderate</u>	Easy	Hard	Runtime
1	F-PointNet	:::		56.77 %	71.96 %	50.39 %	0.17 s
2	VxNet(LiDAR)	*::		48.36 %	61.22 %	44.37 %	0.03 s
3	AVOD-FPN	:::	<u>code</u>	46.12 %	59.97 %	42.36 %	0.1 s
Ku	, M. Mozifian, J. Lee,	, A. Harakeh an	d S. Wasland	ler: <u>Joint 3D Prop</u>	osal Generation	and Object Dete	ection from Vie
1	AVOD	*::	<u>code</u>	44.90 %	60.11 %	38.80 %	0.08 s
Ku	, M. Mozifian, J. Lee,	, A. Harakeh an	d S. Wasland	ler: <u>Joint 3D Prop</u>	osal Generation	and Object Dete	ection from Vie
5	LMNetV2	*::		3.23 %	2.84 %	3.28 %	0.02 s

<u>Pedestrian</u>

	Method	Setting	Code	<u>Moderate</u>	Easy	Hard	Runtime
1	F-PointNet	***		44.89 %	51.21 %	40.23 %	0.17 s
2	AVOD-FPN	::	<u>code</u>	39.00 %	46.35 %	36.58 %	0.1 s
J. Kı	ı, M. Mozifian, J. Lee,	A. Harakeh and	S. Wasland	ler: <u>Joint 3D Prop</u> o	sal Generation	and Object Dete	ection from View
3	<u>VxNet(LiDAR)</u>	:::		33.69 %	39.48 %	31.51 %	0.03 s
4	<u>AVOD</u>	::	<u>code</u>	31.51 %	38.28 %	26.98 %	0.08 s
J. Kı	ı, M. Mozifian, J. Lee,	, A. Harakeh and	I S. Wasland	ler: <u>Joint 3D Prop</u> o	sal Generation	and Object Dete	ction from View
5	<u>3dSSD</u>			17.35 %	20.22 %	17.20 %	0.03 s
6	LMNetV2	* **		11.46 %	13.64 %	11.57 %	0.02 s