Fast R-CNN

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- R-CNN and SPP-nets
- Past R-CNN
 - General architecture
 - Rol pooling layer
 - Pre-trained networks
 - Fine-tuning
- 3 Experiments
- 4 Bonus

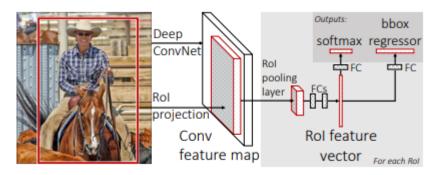
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R-CNN

- R-CNN architecture
 - DeepConv
 - SVM
 - Bbox regression
- Spatial pyramid pooling networks
 - Pooling for matching
 - Increased speed
- Disadvantages
 - Memory
 - Speed

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General architecture



Rol pooling layer

- Rol is described by four numbers (r, c, h, w)
- Pooling grid cell size $H \times W$
- Applied independently to each feature map channel

Pre-trained networks

- 3 networks (5 pooling layers, 5-13 convolutional layers)
- Network preparation
 - Last pooling layer is replaced by Rol pooling layer
 - Last fully connected layer and softmax are replaced with the 2 sibling layers
 - Two data inputs: images and a list of Rols

Fine-tuning

- 128 Images with one Rol VS 2 images, 128 Rols each
- One optimization for bounding box regression and softmax output (multitask loss)
- Two outputs: $p = \{p_0, ..., p_K\}$ and $t^k = \{t_x^k, t_y^k, t_w^k, t_h^k\}$

Multitask loss

- Loss function: $L(p, u, t^u, v) = L_{cls}(p, u) + \lambda[u \ge 1]L_{loc}(t^u, v)$
- Classification loss: $L_{cls}(p, u) = -\log p_u$
- Bbox regressoin loss: $L_{loc}(t^u, v) = \sum_{i \in \{x, y, w, h\}} smooth_{L_1}(t^u_i v_i)$

$$smooth_{L_1}(x) = \begin{cases} 0.5x^2 & \textit{if } |x| < 1; \\ |x| - 0.5 & \textit{otherwise}; \end{cases}$$

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Experiments 1

method	train set	aero	bike bir	rd boat	bottle	bus	car	cat	chair	cow	1
SPPnet BB [11] [†]	07 \ diff	73.9	72.3 62	.5 51.5	44.4	74.4	73.0	74.4	42.3	73.6	-
R-CNN BB [10]	07	73.4	77.0 63	.4 45.4	44.6	75.1	78.1	79.8	40.5	73.7	
FRCN [ours]	07	74.5	78.3 69	.2 53.2	36.6	77.3	78.2	82.0	40.7	72.7	7
FRCN [ours]	07 \ diff	74.6	79.0 68	.6 57.0	39.3	79.5	78.6	81.9	48.0	74.0	
FRCN [ours]	07+12	77.0	78.1 69	.3 59.4	38.3	81.6	78.6	86.7	42.8	78.8	
table dog horse	mbike per	sn plan	it sheep	sofa t	rain tv	m	AP				
57.7 70.3 74.6	74.3 54.	2 34.0	56.4	56.4 (57.9 73.	.5 63	3.1				
62.2 79.4 78.1	73.1 64.	2 35.0	6 66.8	67.2	70.4 71 .	.1 60	5.0				
67.9 79.6 79.2	73.0 69.	0 30.1	1 65.4	70.2	75.8 65.	.8 60	5.9				
67.4 80.5 80.7	74.1 69.	6 31.8	8 67.1	68.4	75.3 65.	.5 68	3.1				
68.9 84.7 82.0	76.6 69.	9 31.8	8 70.1	74.8 8	80.4 70.	4 70	0.0				

Experiments 2

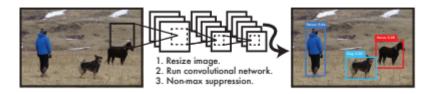
method	train set	aero	bike bire	l boat	bottle	bus (ar cat	chair	cow 1
BabyLearning	Prop.	78.0	74.2 61.3	45.7	42.7	68.2 6	6.8 80.2	2 40.6	70.0
NUS_NIN_c20	00 Unk.	80.2	73.8 61.9	43.7	43.0	70.3 6	7.6 80.3	7 41.9	69.7
R-CNN BB [1] 12	79.6	72.7 61.9	41.2	41.9	65.9 6	6.4 84.6	38.5	67.2
FRCN [ours]	12	80.3	74.7 66.9	46.9	37.7	73.9 6	8.6 87.	7 41.7	71.1
FRCN [ours]	07++12	82.3	78.4 70.3	52.3	38.7	77.8 7	1.6 89.3	3 44.2	73.0
table dog hor	se mbike pe	rsn plai	nt sheep	sofa t	train tv	mA	P		
49.8 79.0 74	5 77.9 6	.0 35.	3 67.9	55.7	68.7 62.	.6 63.2	2		
51.7 78.2 75	2 76.9 65	5.1 38.	6 68.3	58.0	68.7 63.	.3 63.8	3		
46.7 82.0 74	8 76.0 65	5.2 35.	6 65.4	54.2	67.4 60.	.3 62.4	1		
51.1 86.0 77	8 79.8 69	0.8 32.	1 65.5	63.8	76.4 61.	.7 65.7	7		
55.0 87.5 80	5 80.8 72	2.0 35.	1 68.3	65.7	80.4 64.	.2 68.4	ı		

Summary

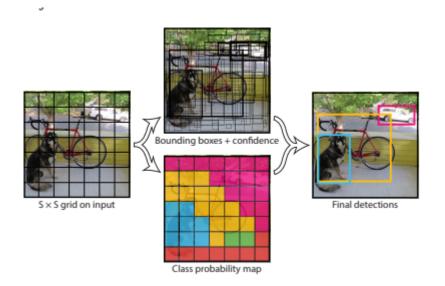
- The first main message of your talk in one or two lines.
- The second main message of your talk in one or two lines.
- Perhaps a third message, but not more than that.
- Outlook
 - Something you haven't solved.
 - Something else you haven't solved.

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YOLO



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