# Scene Graph Generation from Objects, Phrases and Region Captions

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발표자 : 한연지

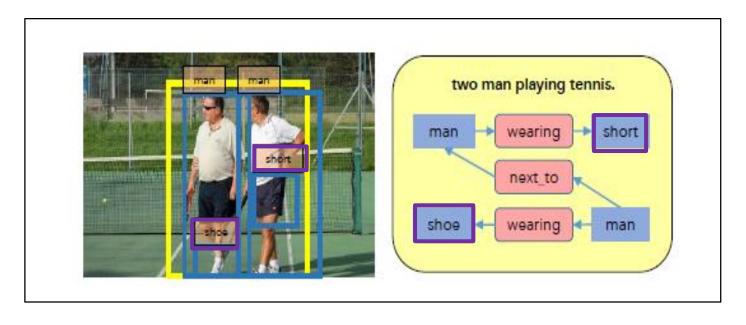


- ICCV 2017 paper (ECCV 2018에도 기본 구조로도 제출됨)
- LI Yikang's Interests
- Computer Vision
- Robotics

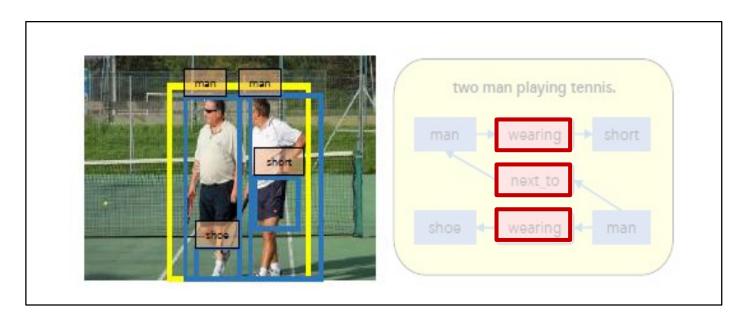
<한 줄 요약 : 이미지 설명할 수 있는(객체, 동사 등을 detection) 여러 모델을 하나의 모델로 사용하여 지식 그래프를 완성>



LI Yikang



**Object Detection** 



**Predicate Detection** 

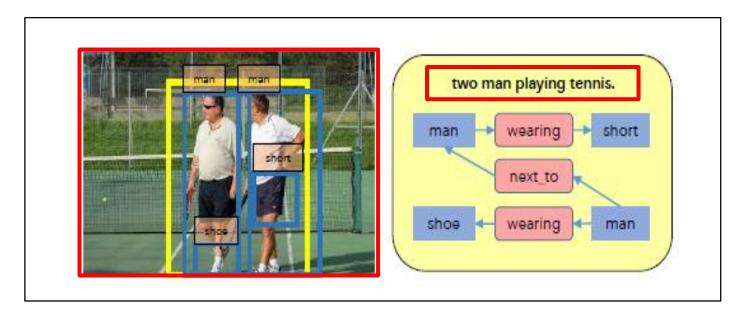
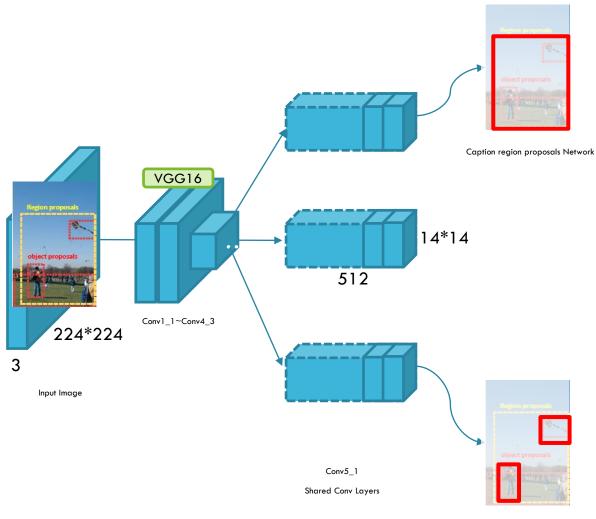
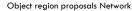


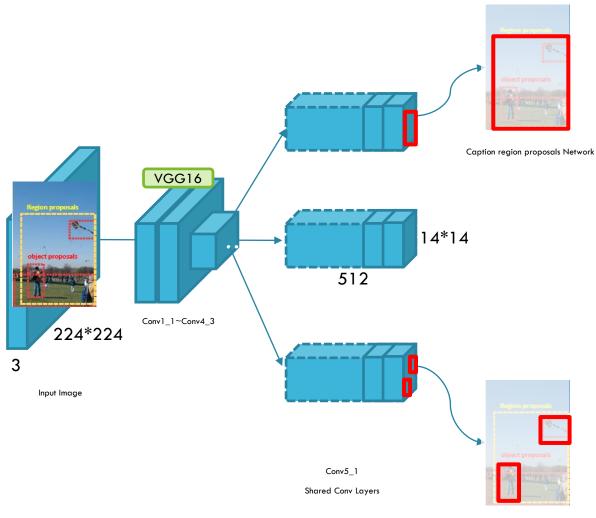
Image Caption

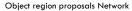




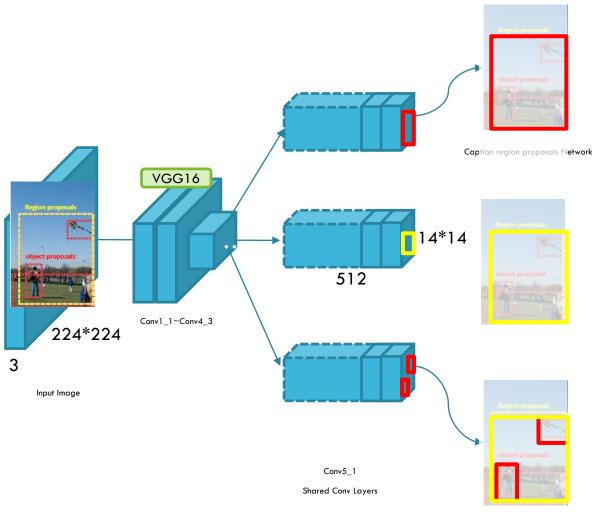


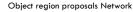


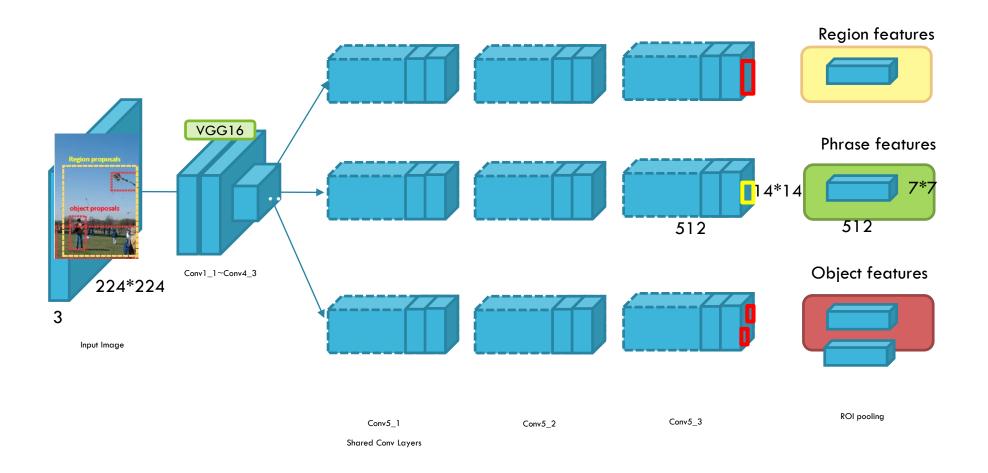


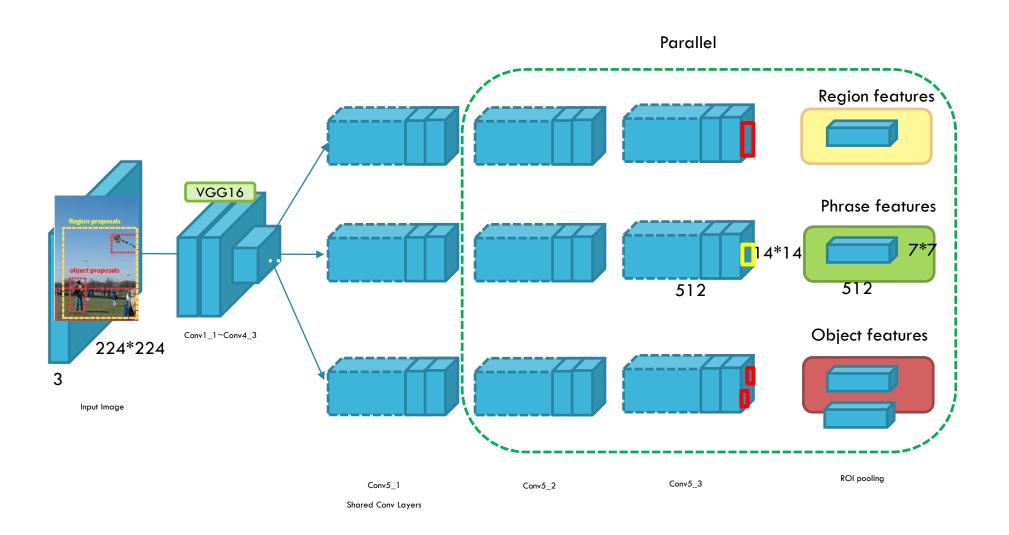




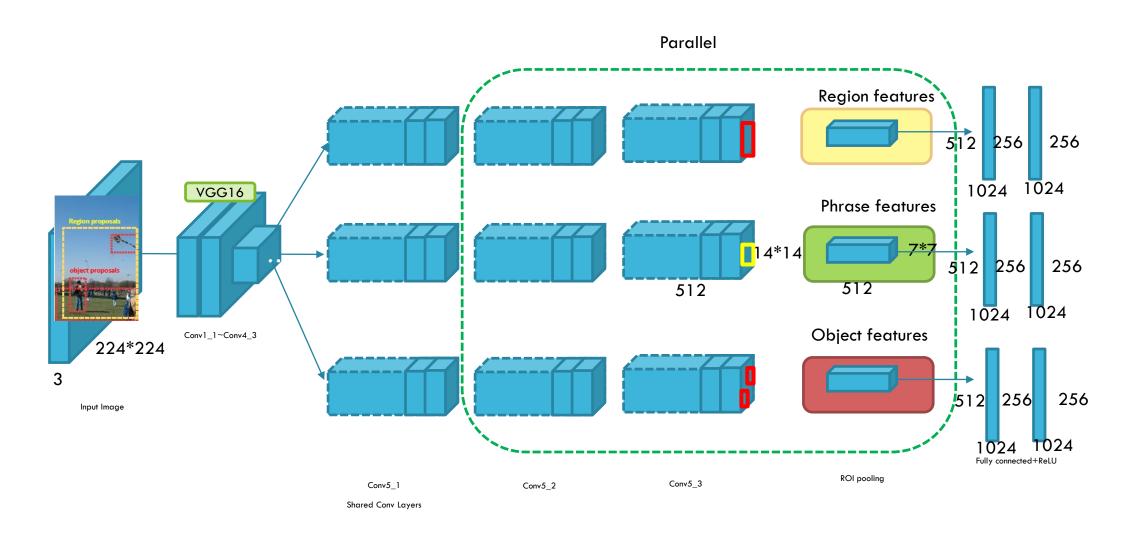




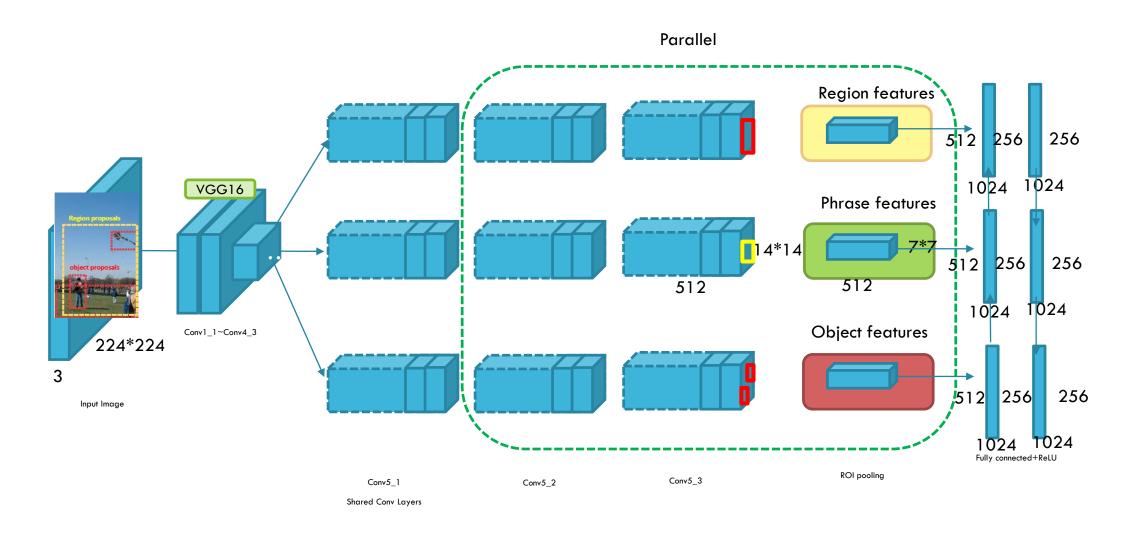


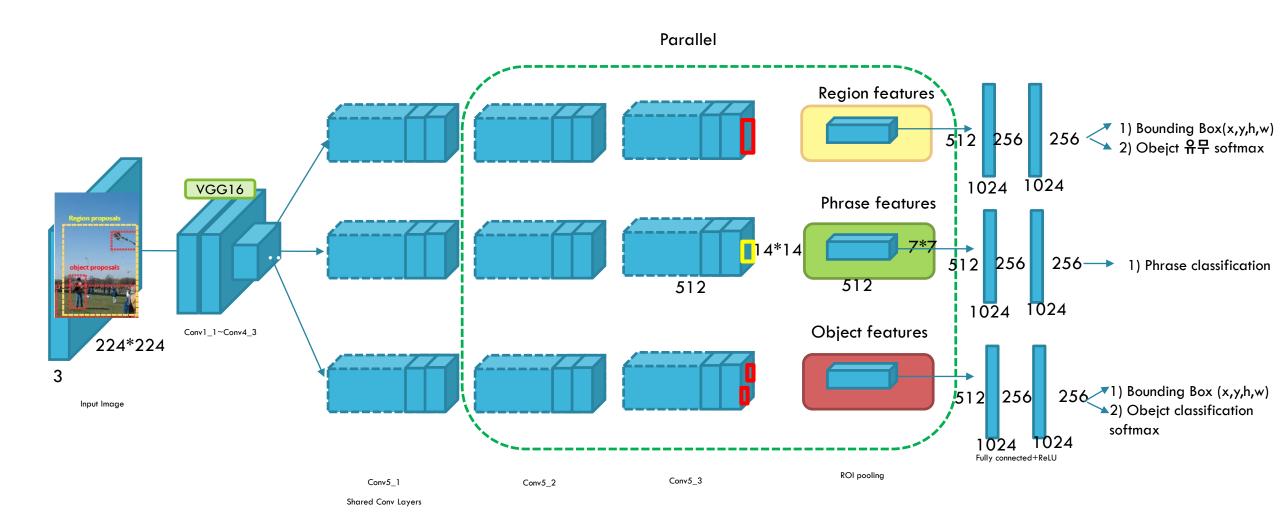




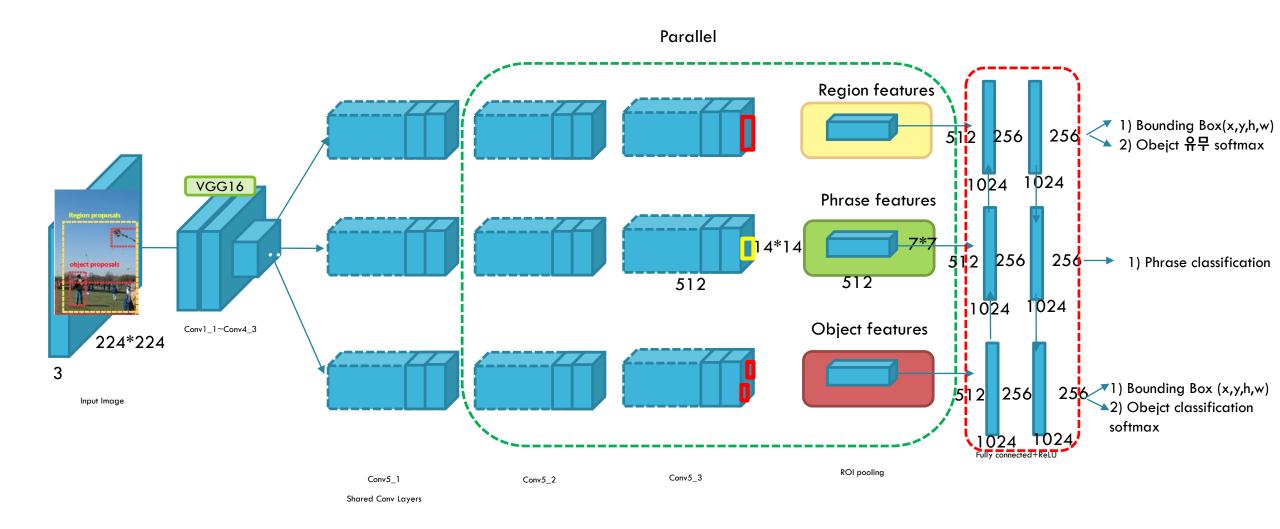




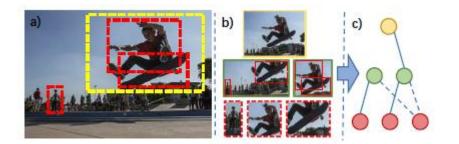




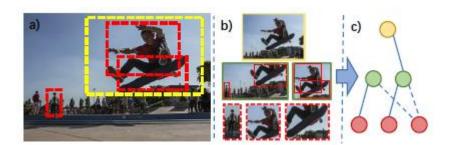


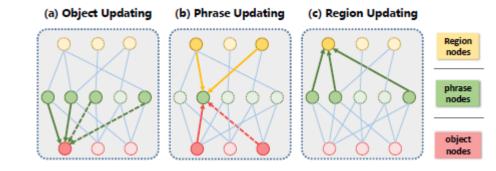




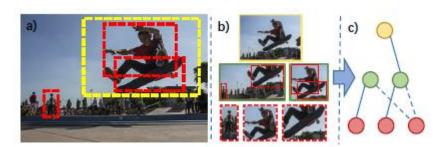


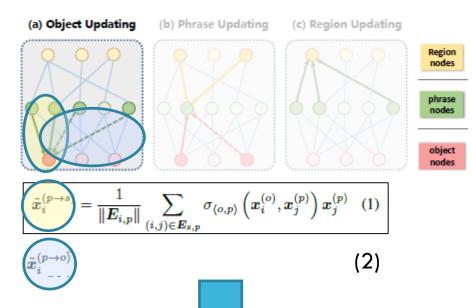










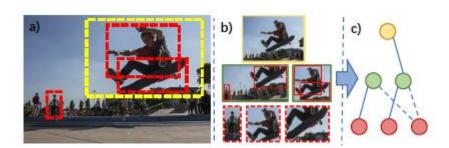


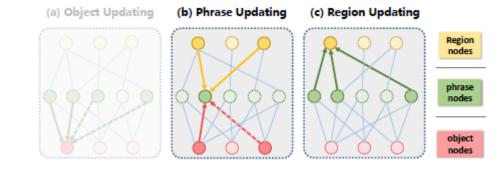
Object feature update위해 (1),(2) 값 사용하여 refining

$$\boldsymbol{x}_{i,t+1}^{(o)} = \boldsymbol{x}_{i,t}^{(o)} + \boldsymbol{F}^{(p \to s)} \left( \tilde{\boldsymbol{x}}_{i}^{(p \to s)} \right) + \boldsymbol{F}^{(p \to o)} \left( \tilde{\boldsymbol{x}}_{i}^{(p \to o)} \right)$$
(3)

$$\boldsymbol{F}(\cdot) = W \cdot ReLU(\cdot)$$

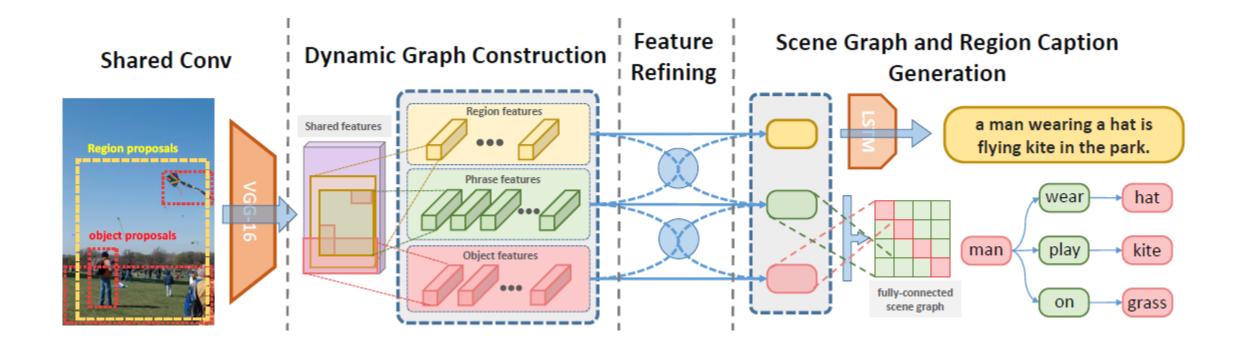


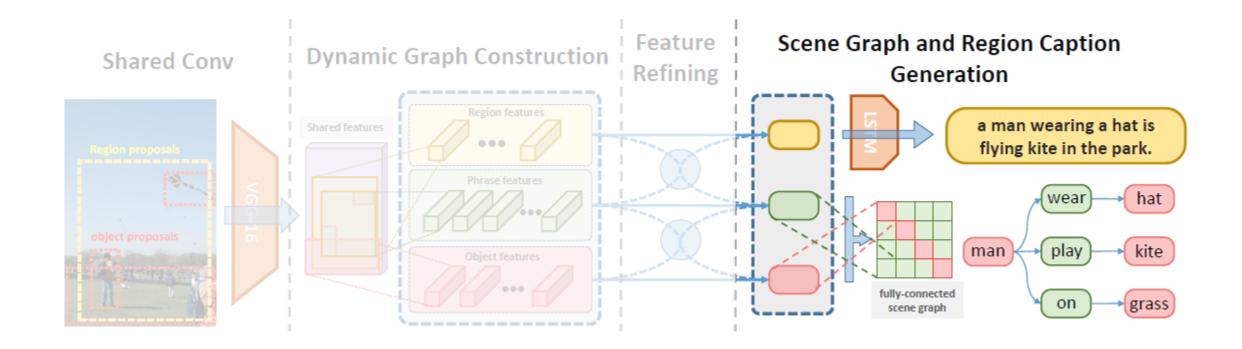


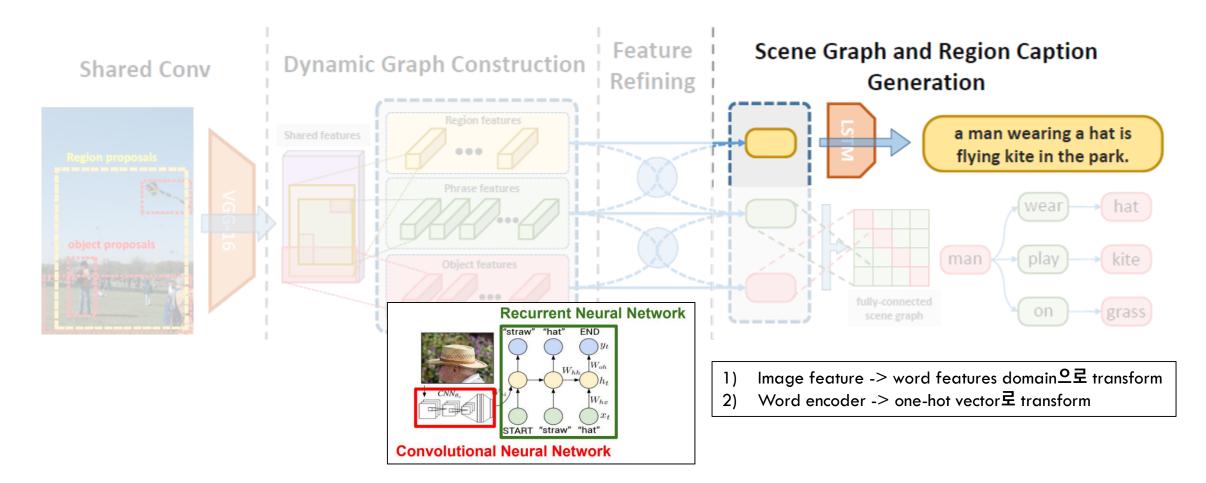


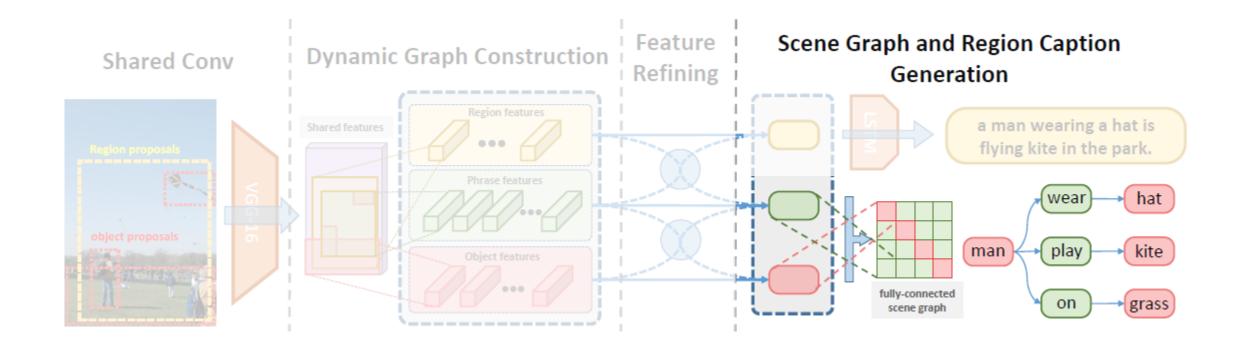
$$x_{j,t+1}^{(p)} = x_{j,t}^{(p)} + \mathbf{F}^{(s \to p)} \left( \tilde{x}_{j}^{(s \to p)} \right) + \mathbf{F}^{(o \to p)} \left( \tilde{x}_{j}^{(o \to p)} \right) + |\mathbf{F}^{(r \to p)} \left( \tilde{x}_{j}^{(r \to p)} \right), \quad (4)$$

$$x_{k,t+1}^{(r)} = x_{k,t}^{(r)} + \mathbf{F}^{(p \to r)} \left( \tilde{x}_{k}^{(p \to r)} \right),$$



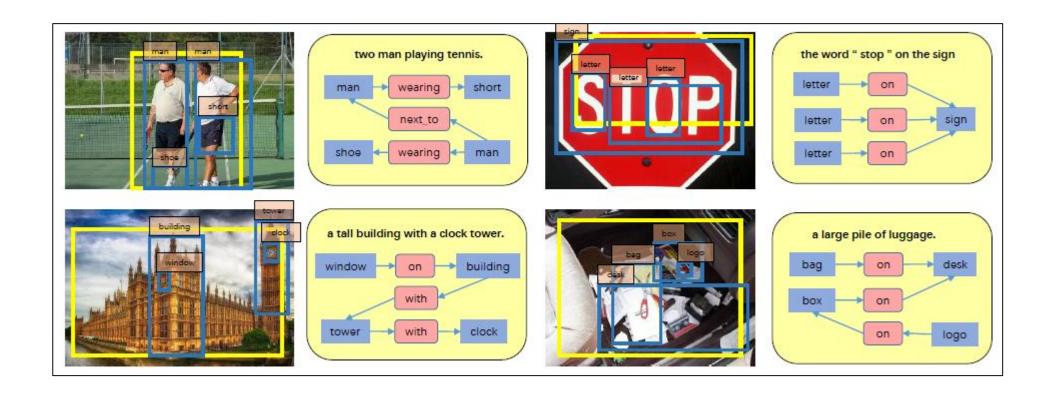








#### The connection is built by proposed dynamic graph generation



# **Experiment**

From the 95,998 images in the dataset, 25,000 images are sampled as the testing set and the remaining 70,998 images are used as the training set. (All the experiments are done on the Visual Genome.)

ID	Message Passing	Cap. branch	Cap. Supervision	FR-iters	PredCls		PhrCls		SGGen	
					Rec@50	Rec@100	Rec@50	Rec@100	Rec@50	Rec@100
1	-	-	-	0	49.28	52.69	7.31	10.48	2.39	3.82
2	✓	-	-	1	63.12	66.41	19.30	21.82	7.73	10.51
3	✓	✓	-	1	63.82	67.23	20.91	23.09	8.20	11.35
4	✓	✓	✓	1	66.70	71.02	23.42	25.68	10.23	13.89
5	✓	✓	✓	2	67.03	71.01	24.22	26.50	10.72	14.22
6	✓	✓	✓	3	66.23	70.43	23.16	25.28	10.01	13.62

- Top-K recall(Rec@K)이 main performance metric으로, top-K predictions가 ground truth instances의 부분일 경우

- PredCls: classification of the predicates

- PhrCls: classification of both predicate and object

- SGGen : Object bounding box가 0.5보다 크며 <Subject-Predicate-Object> classification

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Ta	ısk	LP [31]	ISGG [40]	Ours	
PredCls	R@50	26.67	58.17	67.03	
Fieucis	R@100	33.32	62.74	71.01	
PhrCls	R@50	10.11	18.77	24.34	
FIIICIS	R@100	12.64	20.23	26.50	
SGGen	R@50	0.08	7.09	10.72	
Social	R@100	0.14	9.91	14.22	

LP(Language Prior) : Object detection 후, 위치 관계를 사용하여 word embedding 된 predicate categories 결정 ISGG(Iterative Scene Graph Generation) : GRU-based feature refining scheme.(caption branch 부분이 없음)

- Scene Graph Generation에 있어서 Object, Phrase 외에 Caption 정보를 사용하 여 학습하면 좋은 성능을 냄
- 아쉬웠던 부분
- Feature Refining 부분에서 gate templates이 명확하지 않아서 edge 선택 기준이

모호했던 점