### Connective Cognition Network for Directional Visual Commonsense Reasoning



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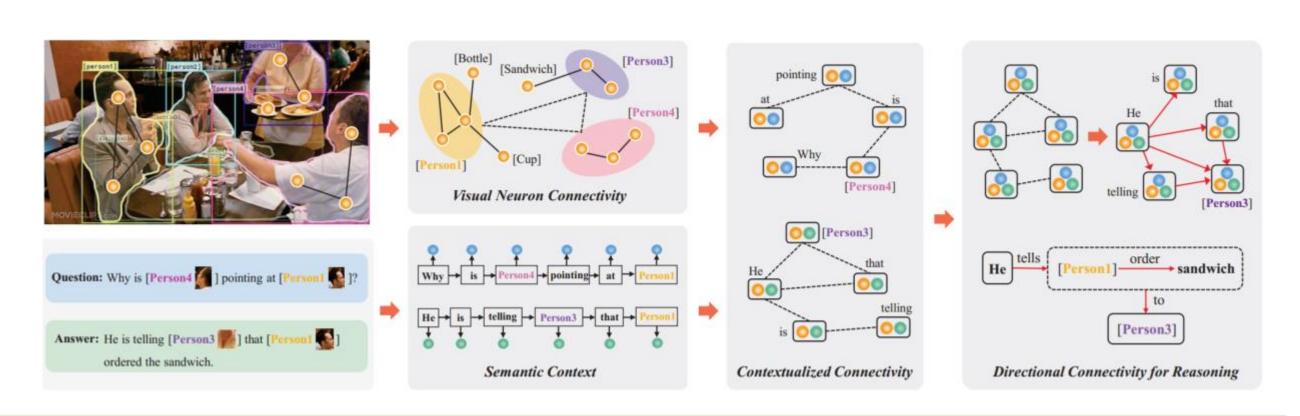
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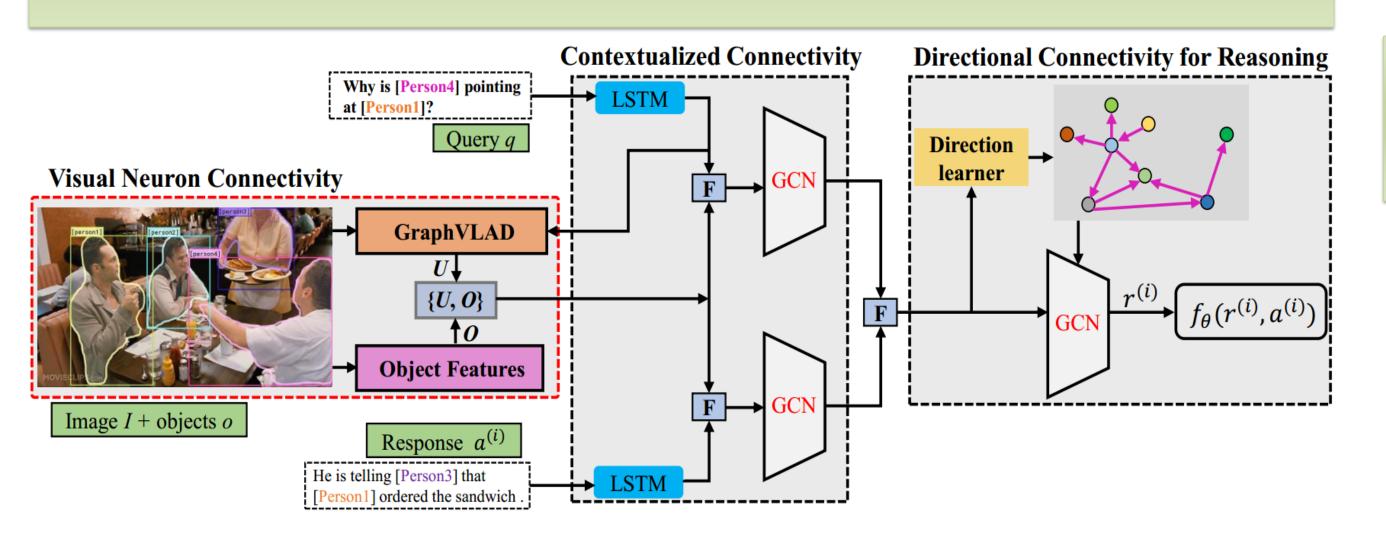


# Introduction

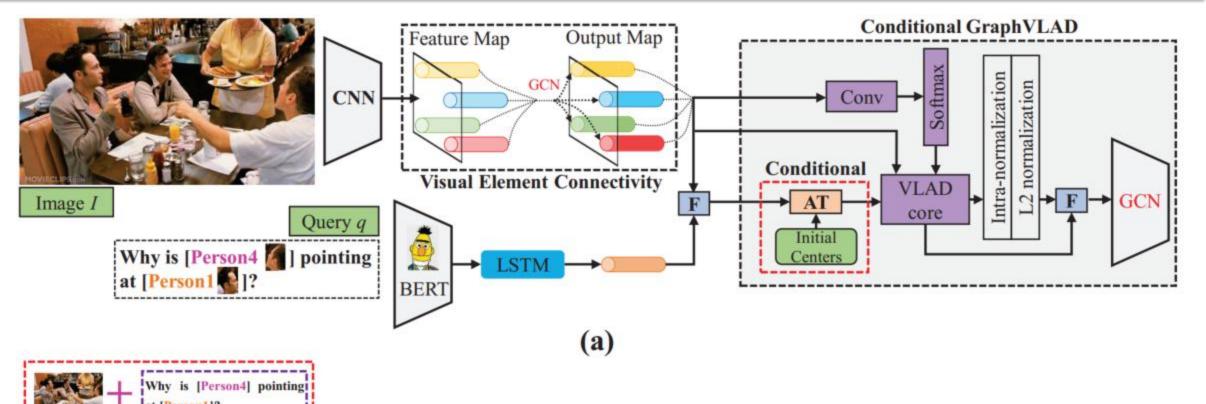
- Visual Commonsense Reasoning (VCR) answering challenging visual questions providing a rationale explaining why its answer is true.
- Connective Cognition Network (CCN) Inspired by neuroscience advances from brain connections to cognition, we propose CCN for VCR.

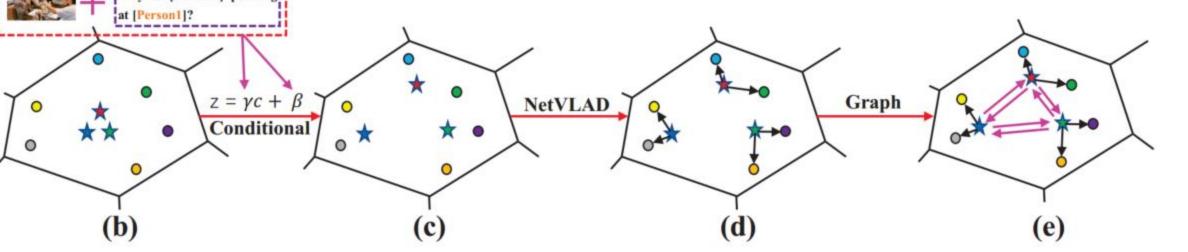


# **Connective Cognition Network**



#### Conditional GraphVLAD





$$M = \mathbf{A}\widetilde{X}, \qquad \widetilde{M} = \tanh(w_f^c * M + b_f^c) \odot \sigma(w_g^c * M + b_g^c),$$
 (1)

$$\gamma = f(|\widetilde{M}, \widetilde{Y}|), \quad \beta = h(|\widetilde{M}, \widetilde{Y}|), \quad z_i = \gamma c_i + \beta,$$
(2)

$$D_{j} = \sum_{i=1}^{N} \frac{e^{w_{j}^{T} \widetilde{M}_{i} + b_{j}}}{\sum_{j'} e^{w_{j'}^{T} \widetilde{M}_{i} + b_{j'}}} (\widetilde{M}_{i} - z_{j})$$

$$(3)$$

### **Contextualized and Directional** Connectivity

$$F_{qu} = softmax(QU^T), \qquad F_{qo} = softmax(QO^T), \qquad Q_U = F_{qu}U, \qquad Q_O = F_{qo}O, \quad (4)$$

$$D_{qa} = \emptyset(E_{qa}), \quad G_t = D_{qa}D_{qa}^T, \qquad D_t = sign(G_t), \qquad V_e = softmax(abs(G_t)),$$
 (5)

$$\mathbf{H} = D_t \odot V_e + I_d, \quad M_t = \mathbf{H} E_{qa}, \quad R_t = \tanh \left( w_f^r * M_t + b_f^r \right) \odot \sigma(w_g^r * M_t + b_g^r), \quad (6)$$

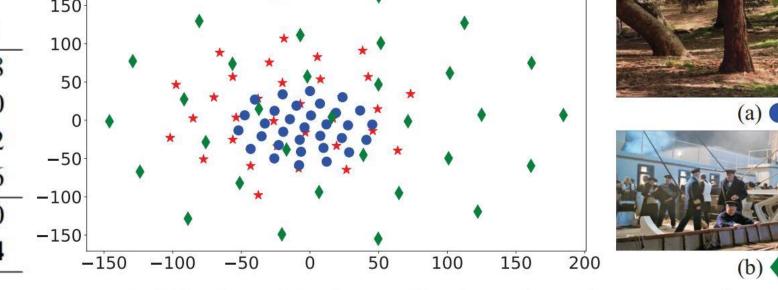
# **Experiments on VCR dataset**

- $\triangleright$  Q  $\longrightarrow$  A: given a question, select the correct answer.
- $\triangleright$  QA  $\longrightarrow$  R: given a question and correct answer, select the correct rationale.

 $\triangleright Q \longrightarrow AR$ : given a question, select the correct answer, then the correct rationale.

Table 1: The performance of our CCN model on the VCR dataset.

	Q -	$\rightarrow A$	QA	$\rightarrow R$	Q -	$\rightarrow AR$	
Model	Val	Test	Val	Test	Val	Test	
Revisited VQA [16]	39.4	40.5	34.0	33.7	13.5	13.8	-86
BottomUpTopDown [1]	42.8	44.1	25.1	25.1	10.7	11.0	
MLB [18]	45.5	46.2	36.1	36.8	17.0	17.2	
MUTAN [5]	44.4	45.5	32.0	32.2	14.6	14.6	
R2C (baseline) [38]	63.8	65.1	67.2	67.3	43.1	44.0	-0 -
CCN	67.4	68.5	70.6	70.5	47.7	48.4	-



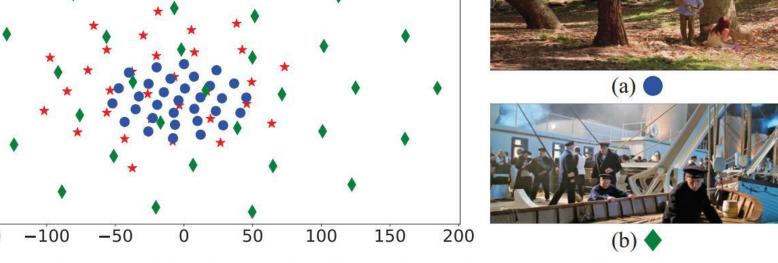
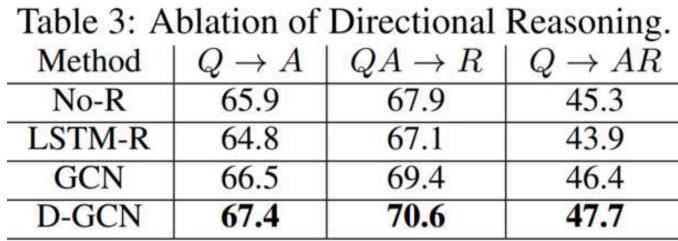
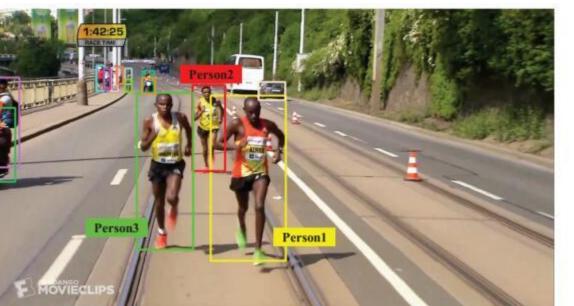


Table 2: Ablation analysis of GraphVLAD.

Method	$Q \to A$	$QA \rightarrow R$	$Q \to AR$
No-C + No-G	65.8	68.3	45.6
No-C	66.5	69.6	46.6
No-G	66.9	69.4	46.5
C + G	67.4	70.6	47.7







a) [Person1, Person3] will start to pick up their

If [Person2] closes the distance between himself erson2] looks like he is really picking up his legs erson1, Person3] flank [Person2] as he walks d) If [Person2] gets short, [Person1, Person3] will have chance of catching him on foot. 0.1%

e view while out for a walk. 10.9% b) [Person1, Person2] and [Person3] are employees o ) They enjoy reading books and consider reading to be d) They are on vacation at a resort. 61.4%

There is sand everywhere and the ocean is in the ackdrop. Many of the men have their shirts off and ding can be seen in the background. d) It is light food with a lot of fruits. It seems like in the

# Conclusion

- We propose a cognition connectivity network for directional visual commonsense reasoning.
- A conditional GraphVLAD module is proposed to represent an image.
- Experimental results demonstrate our method is effective.