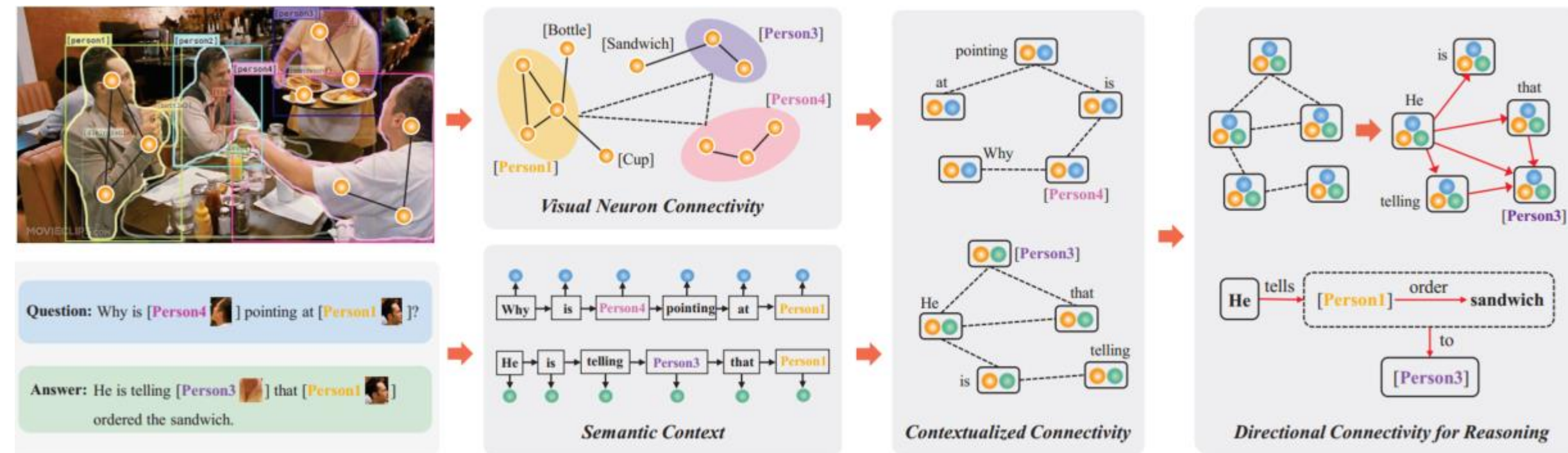
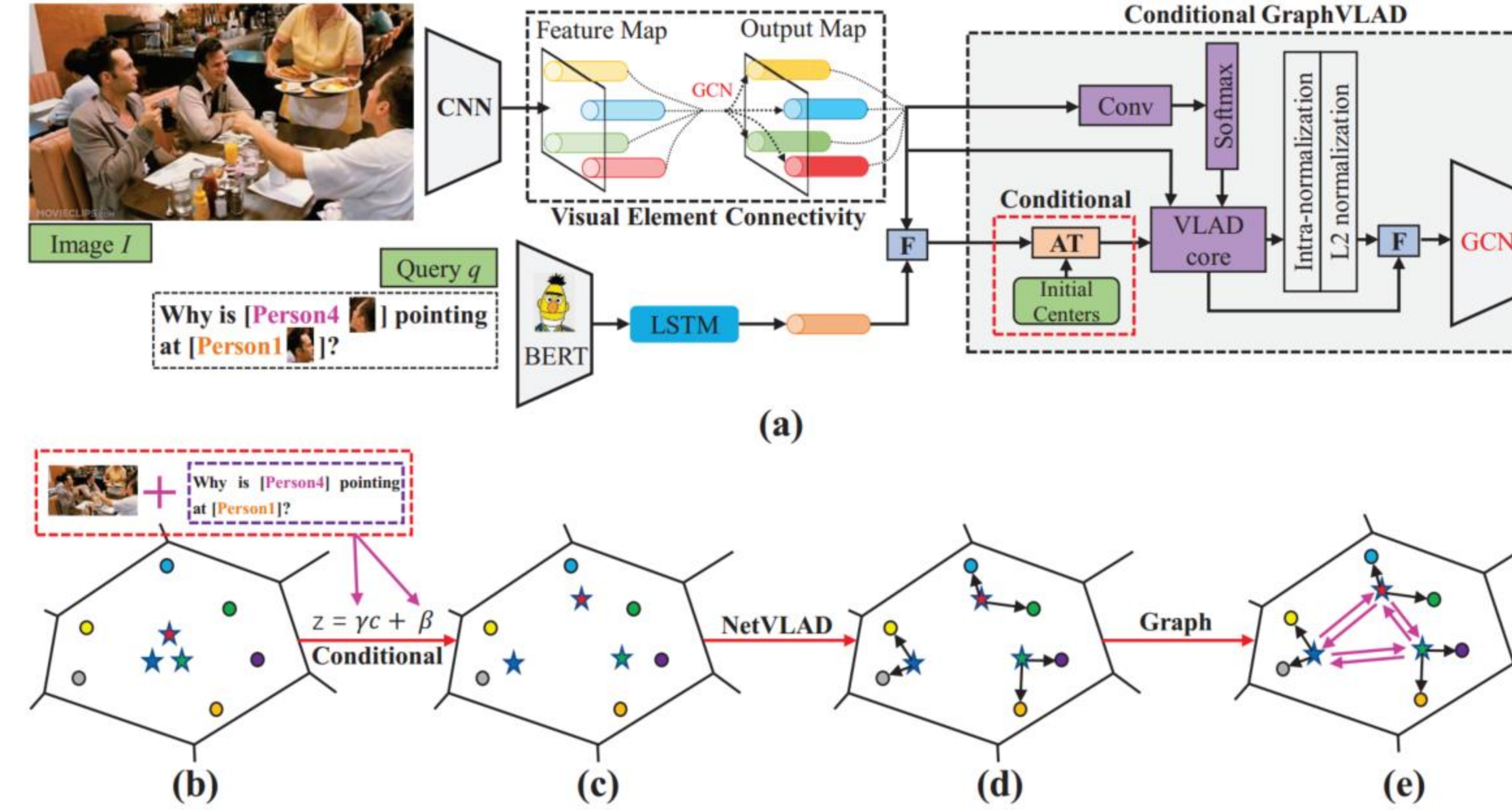


Introduction

- **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.



Conditional GraphVLAD



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

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

$$D_j = \sum_{i=1}^N \frac{e^{w_j^T \tilde{M}_i + b_j}}{\sum_{j'} e^{w_{j'}^T \tilde{M}_i + b_{j'}}} (\tilde{M}_i - z_j) \quad (3)$$

Contextualized and Directional Connectivity

$$F_{qu} = \text{softmax}(\tilde{Q}U^T), \quad F_{qo} = \text{softmax}(\tilde{Q}O^T), \quad Q_U = F_{qu}U, \quad Q_O = F_{qo}O, \quad (4)$$

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

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

Experiments on VCR dataset

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

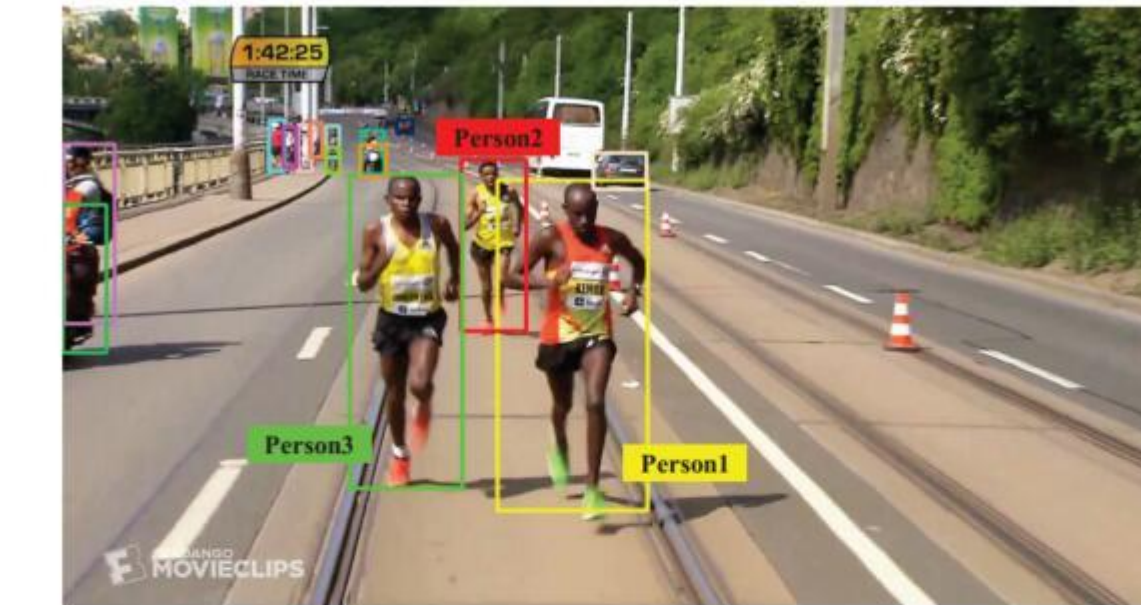
- **$Q \rightarrow AR$** : given a question, select the correct answer, then the correct rationale.

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

Model	$Q \rightarrow A$		$QA \rightarrow R$		$Q \rightarrow AR$	
	Val	Test	Val	Test	Val	Test
Revisited VQA [16]	39.4	40.5	34.0	33.7	13.5	13.8
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
CCN	67.4	68.5	70.6	70.5	47.7	48.4

Table 2: Ablation analysis of GraphVLAD.

Method	$Q \rightarrow A$	$QA \rightarrow R$	$Q \rightarrow AR$
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



What will [Person1, Person3] do if [Person2] catches up to them?

a) [Person1, Person3] will start to pick up their paces and run faster if [Person2] catches up. 97.5%

b) [Person1, Person3] will fly away. 1.2%

c) [Person1, Person3] will scream for [Person2]. 1.0%

d) [Person1, Person3] hug, and follow [Person2] to their destination. 0.3%

The rationale is ...

a) If [Person2] closes the distance between himself and [Person1, Person3], then [Person1, Person3] will be concerned that [Person2] is going to push ahead of them, so they will run faster because they want to keep in the lead. 99.3%

b) [Person2] looks like he is really picking up his legs to try to set himself apart from the other runners. 0.2%

c) [Person1, Person3] flank [Person2] as he walks between them. 0.4%

d) If [Person2] gets short, [Person1, Person3] will have a chance of catching him on foot. 0.1%

Why are [Person1, Person2] and [Person3] have brunch?

a) [Person1, Person2] and [Person3] are just enjoying the view while out for a walk. 10.9%

b) [Person1, Person2] and [Person3] are employees of [Person1]. 0.4%

c) They enjoy reading books and consider reading to be a worthwhile hobby. 27.3%

d) They are on vacation at a resort. 61.4%

The rationale is ...

a) There is sand everywhere and the ocean is in the backdrop. Many of the men have their shirts off and there are women wearing bikinis. 5.2%

b) They have just married and it is custom to enjoy a vacation after a wedding. 23.7%

c) They are dressed in clothing someone might wear on vacation. There is a lake behind them and a large building can be seen in the background. 41.9%

d) It is light food with a lot of fruits. It seems like in the afternoon which is brunch time. 29.2%

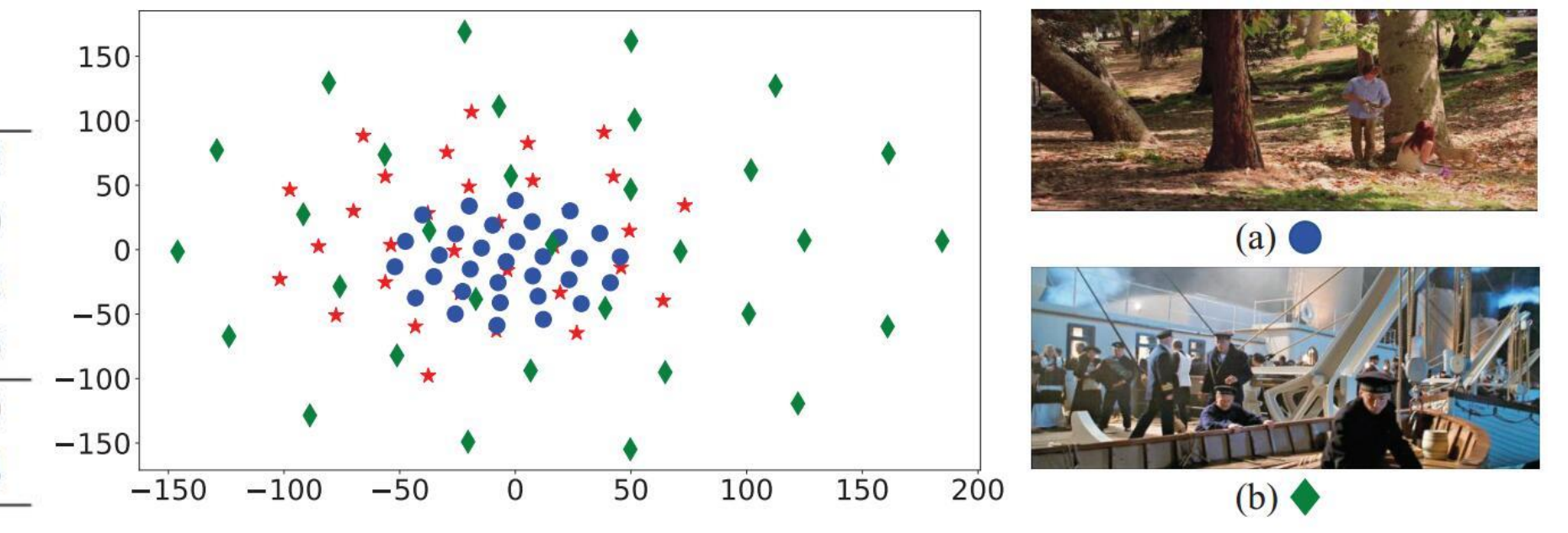
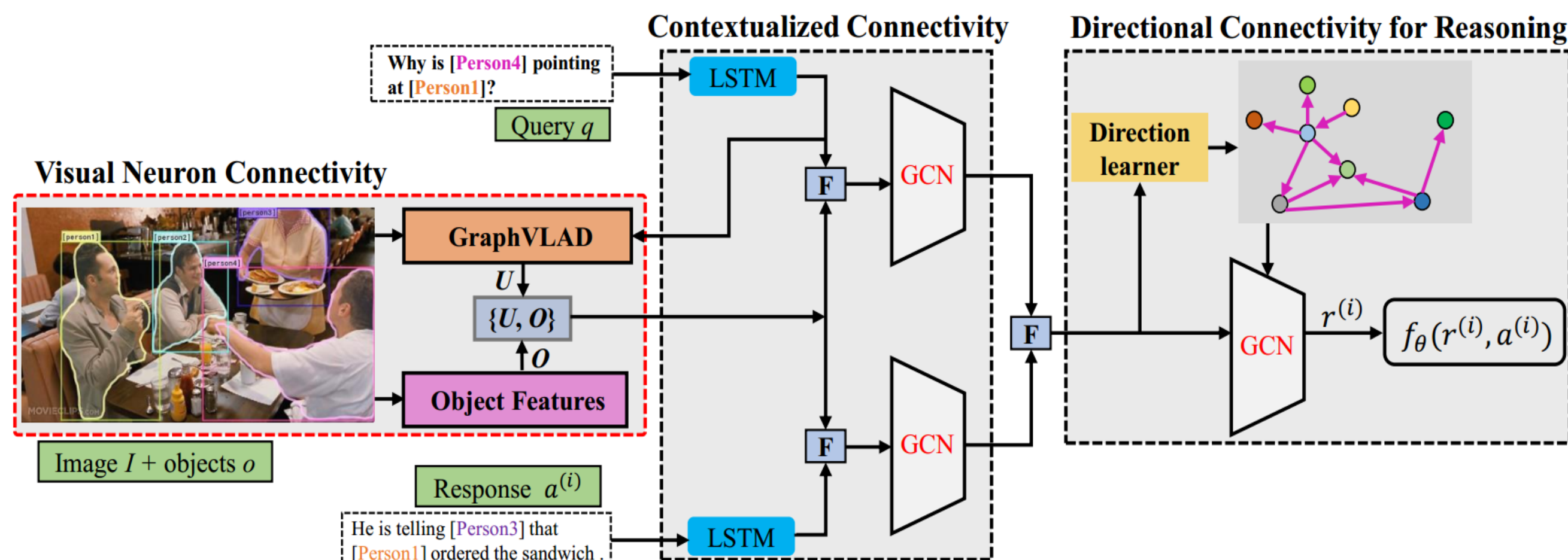


Table 3: Ablation of Directional Reasoning.

Method	$Q \rightarrow A$	$QA \rightarrow R$	$Q \rightarrow AR$
No-R	65.9	67.9	45.3
LSTM-R	64.8	67.1	43.9
GCN	66.5	69.4	46.4
D-GCN	67.4	70.6	47.7



Connective Cognition Network



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.