Artificial

Knowledge-Bridged Causal Interaction Network for Causal Emotion Entailment

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Code: https://github.com/circle-hit/KBCIN.

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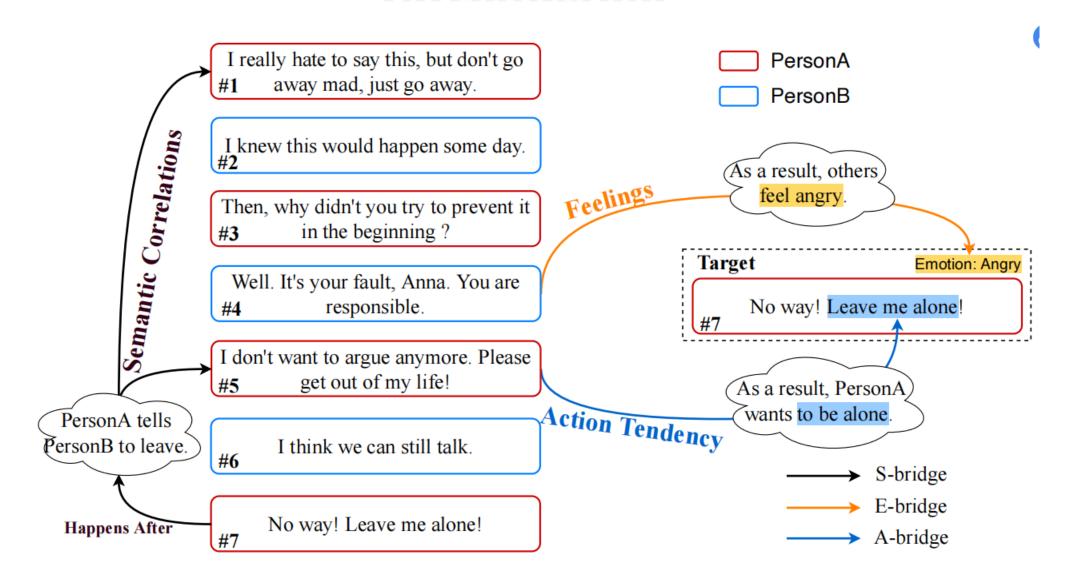




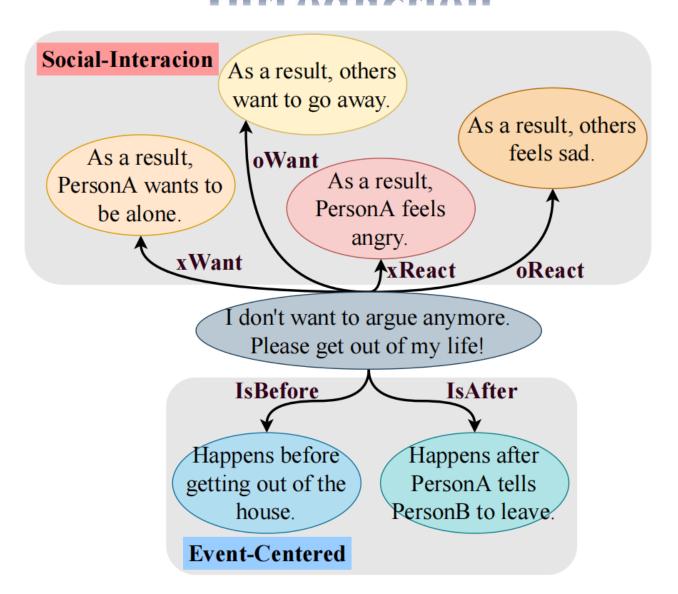




Introduction

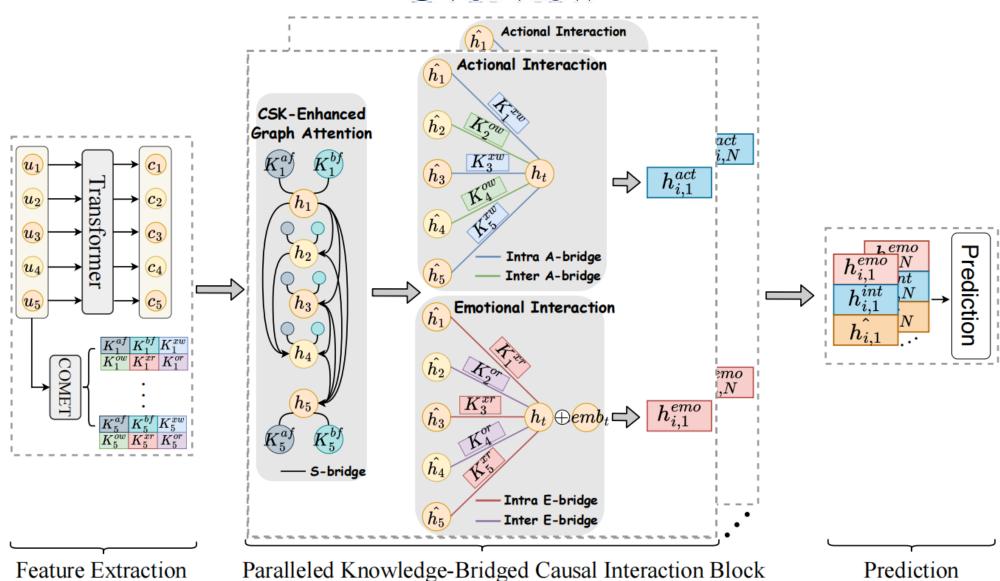


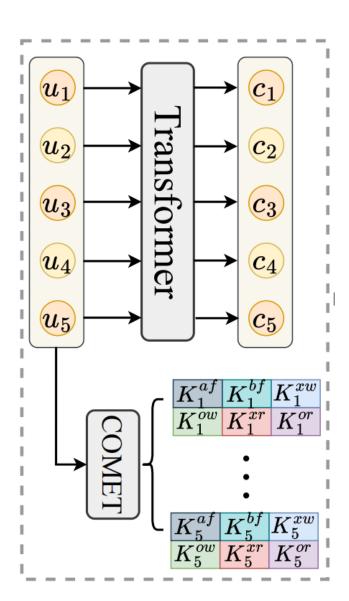
Introduction



Intelligence

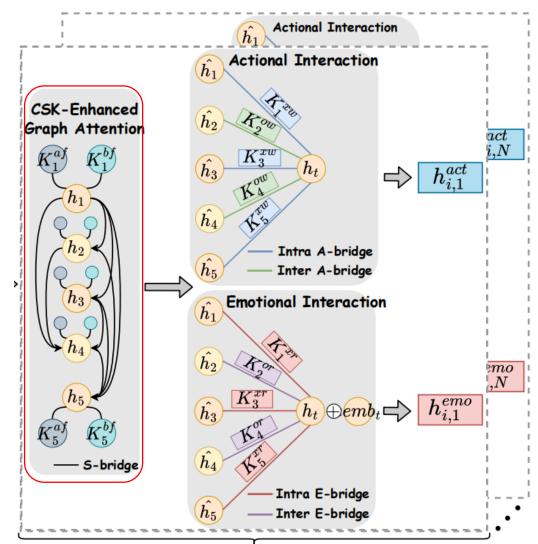
Overview





$$c_i = \text{Maxpooling}(\text{Transformer}([CLS], w_1, w_2, \cdots, w_L))$$
(1)

 $K_i^r, r \in \{af, bf, xr, or, xw, ow\}$ and af, bf, xr, or, xw, ow are short for the relation type isAfter, isBefore, xReact, oReact, xWant and oWant, respectively.

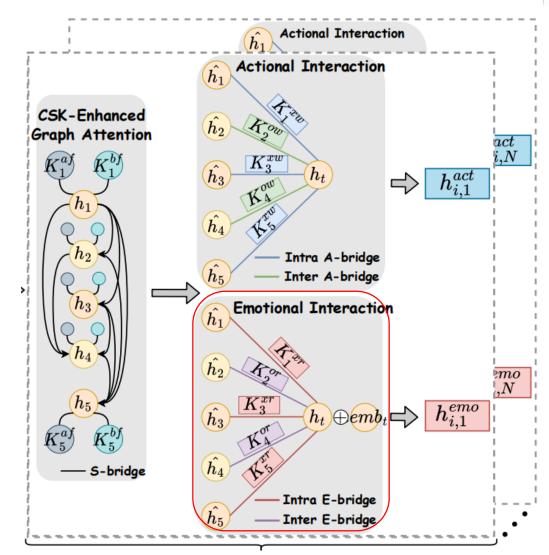


$$h_i = c_i \oplus pemb_i \oplus eemb_i \tag{2}$$

$$\hat{h_i} = \sigma(\sum_{j \in N_i} \alpha_{ij} W_h h_j) \tag{3}$$

$$\alpha_{ij} = \frac{\exp(\mathcal{F}(h_i, h_j, K_j^{af}, K_j^{bf}))}{\sum_{j' \in N_i} \exp(\mathcal{F}(h_i, h_{j'}, K_{j'}^{af}, K_{j'}^{bf}))}$$
(4)

$$\mathcal{F}(h_i, h_j, K_j^{af}, K_j^{bf}) = \text{LeakyReLU}(a^{\top}[W_h h_i \parallel W_h h_j + W_e K_j^{af} + W_e K_j^{bf}])$$
(5)

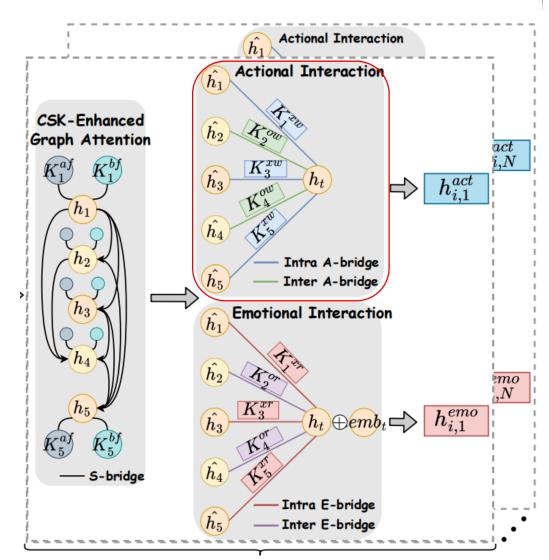


Paralleled Knowledge-Bridged Causal Interaction Block

$$Q^{emo} = f_q(h_t + eemb_t), \ K^{emo} = f_k(\hat{h_i}) + f_e(K_i^r)$$
$$s_i^{emo} = \operatorname{softmax}(\frac{Q^{emo}(K^{emo})^{\top}}{\sqrt{d_h}})$$
(6)

$$h_i^{emo} = s_i^{emo} V^{emo} + s_i^{emo} Q^{emo}$$

$$V^{emo} = f_v(\hat{h}_i) + f_e(K_i^r)$$
(7)

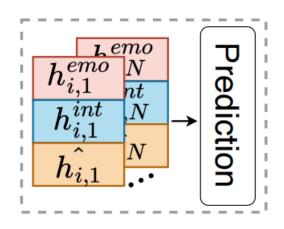


$$Q^{act} = f'_q(h_t), \ K^{act} = f'_k(\hat{h}_i) + f'_e(K_i^r)$$
$$s_i^{act} = \operatorname{softmax}(\frac{Q^{act}(K^{act})^{\top}}{\sqrt{d_h}})$$
(8)

$$h_i^{act} = s_i^{act} V^{act} + s_i^{act} Q^{act}$$

$$V^{act} = f_v'(\hat{h}_i) + f_e'(K_i^r)$$
(9)

Paralleled Knowledge-Bridged Causal Interaction Block



$$\tilde{h_i} = \hat{h_i} + h_i^{emo} + h_i^{act} \tag{10}$$

$$\hat{y}_i = \operatorname{sigmoid}(\operatorname{MLP}(\|_{n=1}^N \tilde{h}_{i,n}))$$
 (11)

	Train	Valid	Test
Positive Causal Pairs	7,027	328	1,767
Negative Causal Pairs	20,646	838	5,330
Num. of Dialogue	834	47	225
Num. of Utterance	8,206	493	2,405

Table 1: Dataset statistics

Model	Neg. F1	Pos. F1	macro F1		
ECE Methods					
KAG	86.35	58.18	72.26		
Adapted	88.18	64.53	76.36		
ECPE Methods					
ECPE-2D	94.96	55.50	75.23		
ECPE-MLL	94.68	48.48	71.59		
RankCP	97.30	33.00	65.15		
CEE Methods					
RoBERTa-Base	88.74	64.28	76.51		
RoBERTa-Large	87.89	66.23	77.06		
KEC	88.85	66.55	77.70		
KBCIN (Ours)	89.65	68.59	79.12		

Table 2: Comparison of our model against state-of-the-art baselines of CEE, ECE and ECPE.

S-bridge	E-bridge	A-bridge	Pos. F1	macro F1
·	V	/	68.59	79.12
×	✓	~	67.47	78.63
✓	×	~	66.33	77.78
~	✓	×	66.92	78.37
×	×	×	57.59	71.81

Table 3: Results of ablation study.

Emotion	Pos. F1	macro F1
Gold Emotion	68.59	79.12
Predicted Emotion	67.51	78.43
No Emotion	64.05	76.73

Table 4: Results of our proposed KBCIN with different ways of using emotion information.

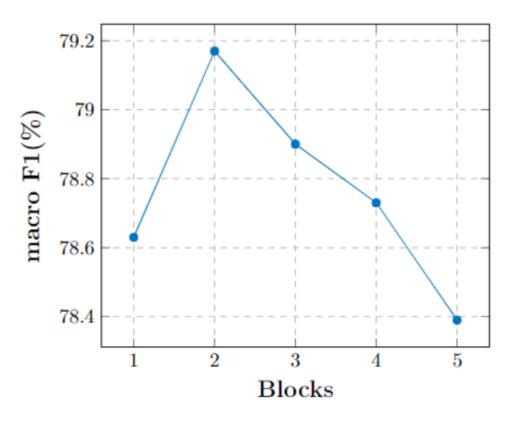


Figure 4: Results of our proposed model with different numbers of Knowledge-Bridged Causal Interaction block.

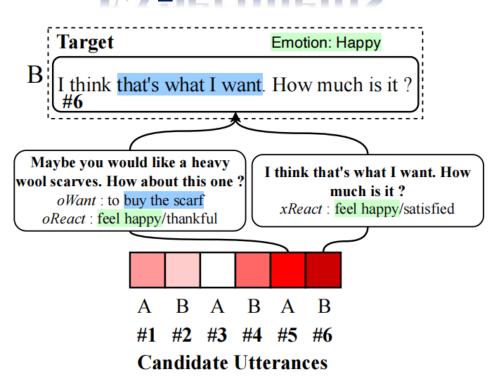


Figure 5: A case that our model makes the right prediction, along with the visualizations for attention weights of the candidate utterances (obtained by summing up the measuring results after the process of Emotional Interaction and Actional Interaction from each block). The darker color mean larger attention weights.



Thanks!