

Chongqing University of Technology

ATAI Advanced Technique of Artificial Intelligence

Artificial

DialogueEIN: Emotion Interaction Network for Dialogue Affective Analysis

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Code: https://github.com/AIM3-RUC/DialogueEIN

2023. 5. 6 • ChongQing

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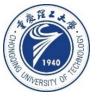


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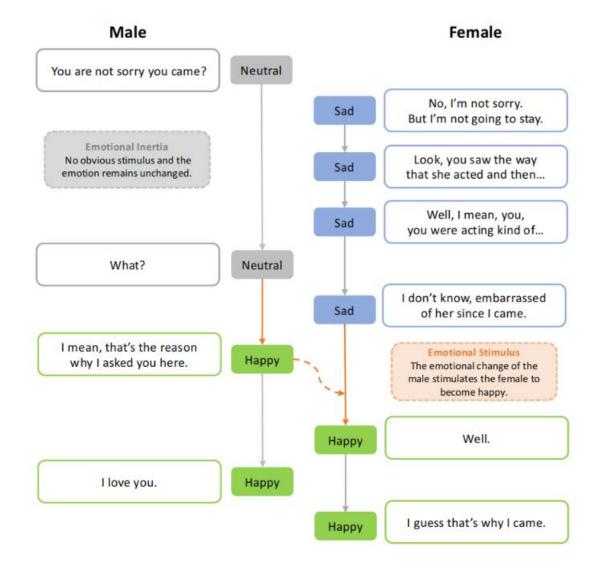






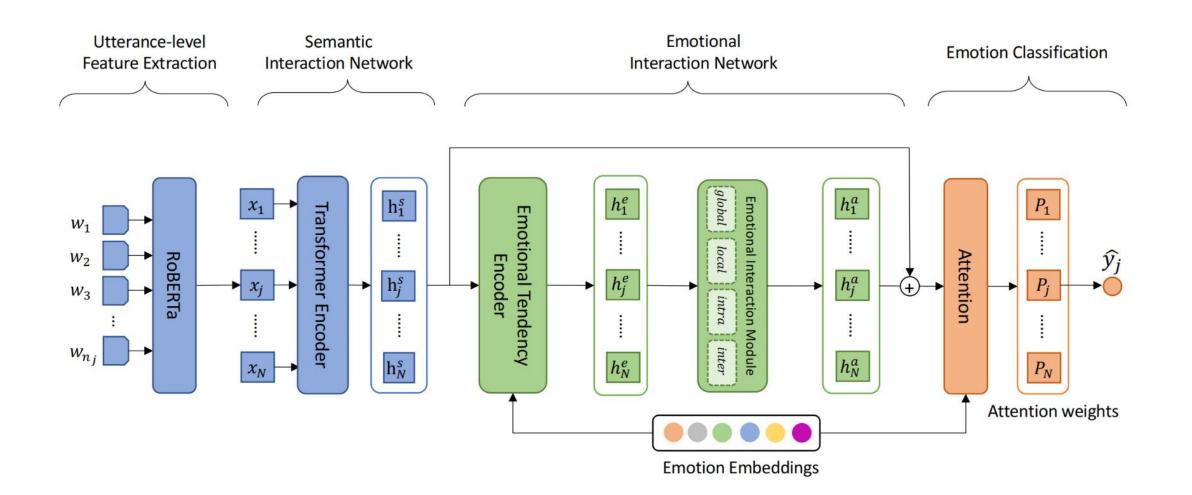


Introduction



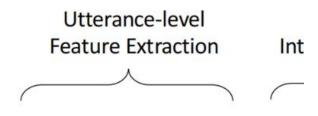


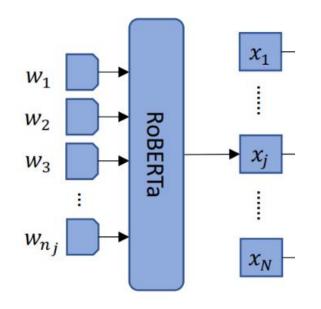
Overview











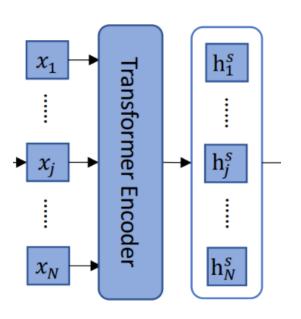
$$X_{j} = \text{RoBERTa}([CLS], w_{1}^{j}, w_{2}^{j}, ..., w_{n_{j}}^{j})$$
(1)

$$x_j = W_u X_{j,0} + b_u \tag{2}$$





Semantic Interaction Network

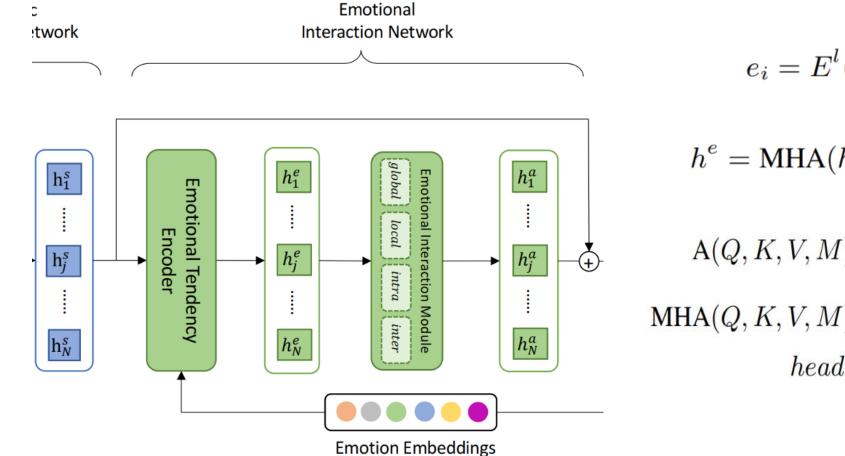


$$h^{0} = [x_{1}, x_{2}, ..., x_{N}] + \text{PosEnc}(0:N)$$
 (3)

$$h^{s} = \text{TRMEncoder}(h^{0}) \tag{4}$$







$$e_i = E^l(l_i) \tag{5}$$

$$h^{e} = \mathrm{MHA}(h^{s}, e, e) + h^{s} \tag{6}$$

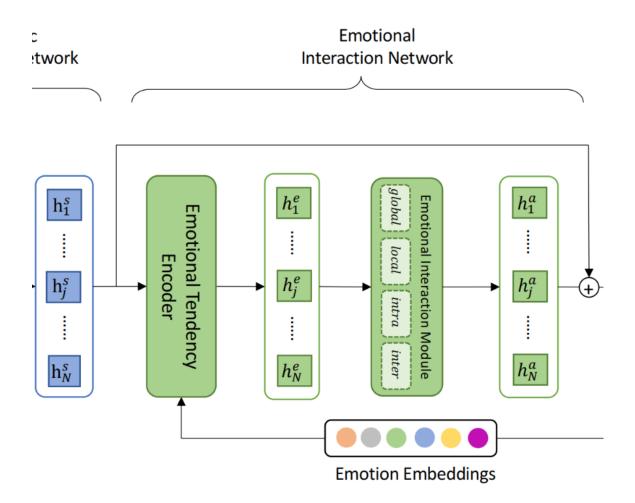
$$A(Q, K, V, M) = \operatorname{softmax}\left(\frac{QK^T}{\sqrt{d_h}} + M\right)V$$
(7)

 $MHA(Q, K, V, M) = Concat(head_1, ..., head_n)W^O$ (8)

 $head_i = \mathcal{A}(QW_i^Q, KW_i^K, VW_i^V, M) \quad (9)$







 $h^{ei} = \text{Concat}\{\text{MHA}_{EI}(h^e, h^e, h^e, m) | m \in M\}$ (10)

$$h^{a} = \text{LayerNorm}(h^{ei}W^{a} + b_{a} + h^{s})$$
(11)

$$m_{i,j}^{global} = 0 \tag{12}$$

$$m_{i,j}^{local} = \begin{cases} 0, & if |i-j| < w/2 \\ -\infty, & otherwise \end{cases}$$
(13)

$$m_{i,j}^{intra} = \begin{cases} 0, & if \ p(u_i) = p(u_j) \\ -\infty, & otherwise \end{cases}$$
(14)

$$m_{i,j}^{inter} = \begin{cases} 0, & if \ p(u_i) \neq p(u_j) \ or \ i = j \\ -\infty, & otherwise \end{cases}$$
(15)

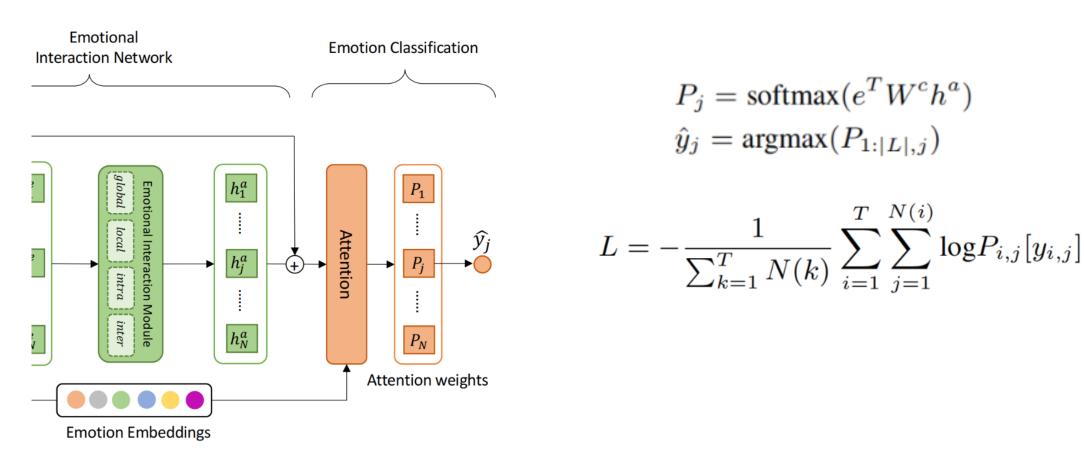


(16)

(17)

(18)







Experiments

Dataset	dialogues			utterances		
	train	val	test	train	val	test
IEMOCAP	100	20	31	4830	980	1623
MELD	1038	114	280	9989	1109	2610
EmoryNLP	713	99	85	9934	1344	1328
DailyDialog	11118	1000	1000	87170	8069	7740

Table 1: Data distribution of the four datasets.





	IEMOCAP	MELD	EmoryNLP	DailyDialog
	Avg(w)	Avg(w)	Avg(w)	Avg(micro)
KET	59.56	58.18	33.95	53.37
DialogueGCN	64.18	58.10	-	-
+RoBERTa	64.91	63.02	38.10	57.52
DialogXL	65.94	62.41	34.73	54.93
DAG-ERC	68.03	63.65	39.02	59.33
DialogueRNN	62.75	57.03	-	-
+RoBERTa	64.76	63.61	37.44	57.32
COSMIC	65.25	65.21	38.11	58.48
CESTa	67.10	58.36	-	63.12
DialogueEIN(Ours)	68.93	65.37	<u>38.92</u>	<u>62.58</u>



Experiments

	IEMOCAP	MELD
DialogueEIN	68.93	65.37
- intra-speaker attention	68.63 (0.30↓)	64.89 (0.48↓)
- inter-speaker attention	68.43 (0.40↓)	65.10 (0.27↓)
- intra-&inter-speaker attention	67.36 (1.57↓)	64.84 (0.51↓)
- global&local attention	67.71 (1.22↓)	64.70 (0.67↓)
- Emotional Tendency Encoder	67.68 (1.25↓)	64.85 (0.52↓)
- Emotional Interaction Network	66.04 (2.89 ↓)	64.59 (0.78 ↓)

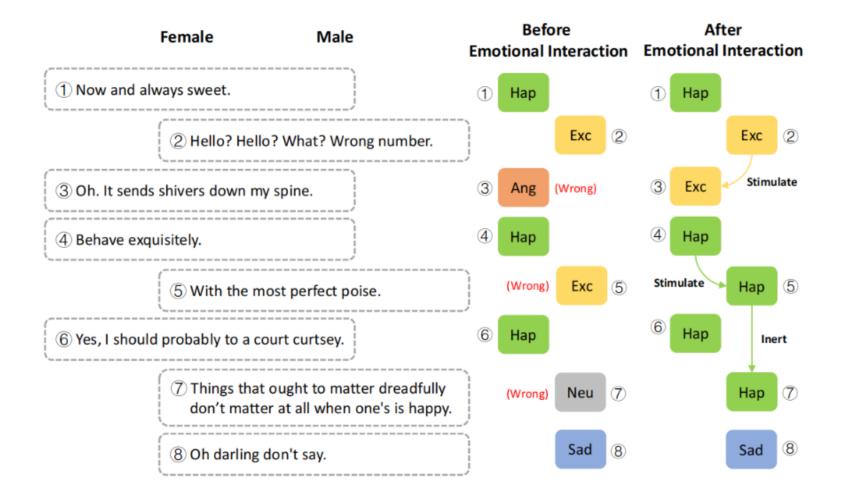




		IEMOCAP	MELD
1	RoBERTa	63.38	62.88
2	+TRM	66.04	64.59
3	+TRM&Attentions	67.55	64.95
4	+TRM&CRF	67.11	64.62
5	+TRM&Attentions&CRF	67.76	64.69
6	DialogueEIN	68.93	65.37



Experiments





Thanks!