



Conversation Understanding using Relational Temporal Graph Neural Networks with Auxiliary Cross-Modality Interaction

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Reported by Renhui Luo





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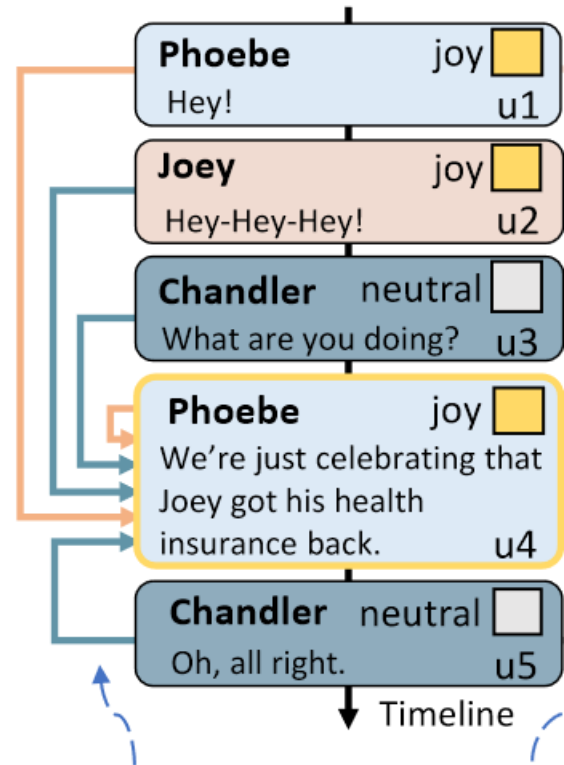
Introduction

Emotion	Utterance text	Utterance text	Emotion
excited	Just a couple days ago.	Oh, sure this is standing on the beach, this is waiting, fighting.	angry
excited	Oh my gosh.	Right.	neutral
excited	...I can't believe it. I never thought you would get married.	This isn't anything like I thought anything would be.	sad
excited		I know me neither.	sad
excited	Oh my gosh.	This is just this...	sad

ID: Ses01M_script02_2 ID: Ses05F_impro03

 : Speaker 1	 : Speaker 2
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Introduction



Simplified dependencies

Intra-Speaker →

Inter-Speaker →

Self and Inter dependency

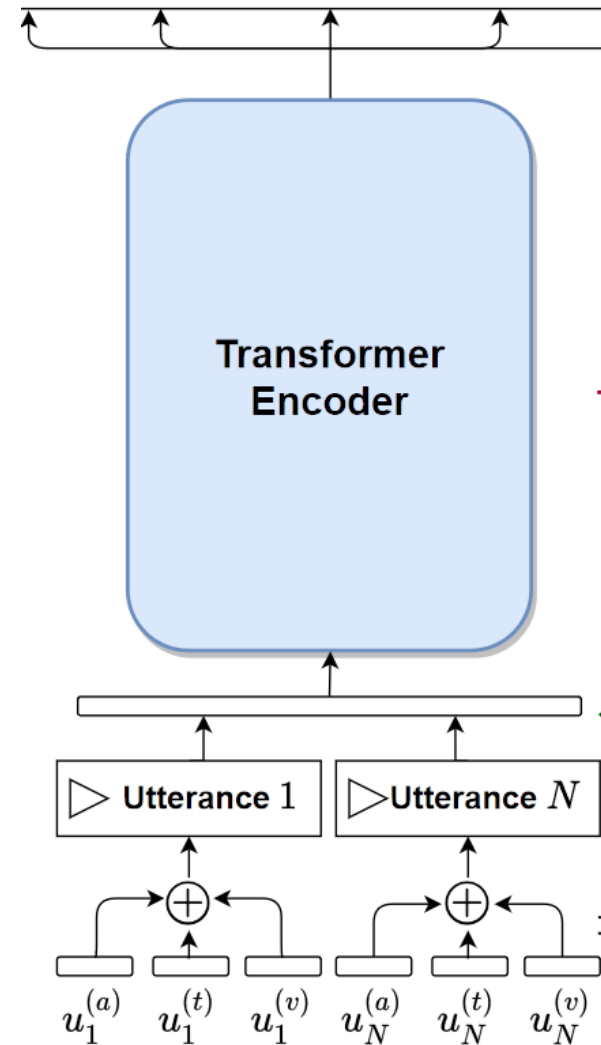
Pho→Pho —

Cha→Pho —

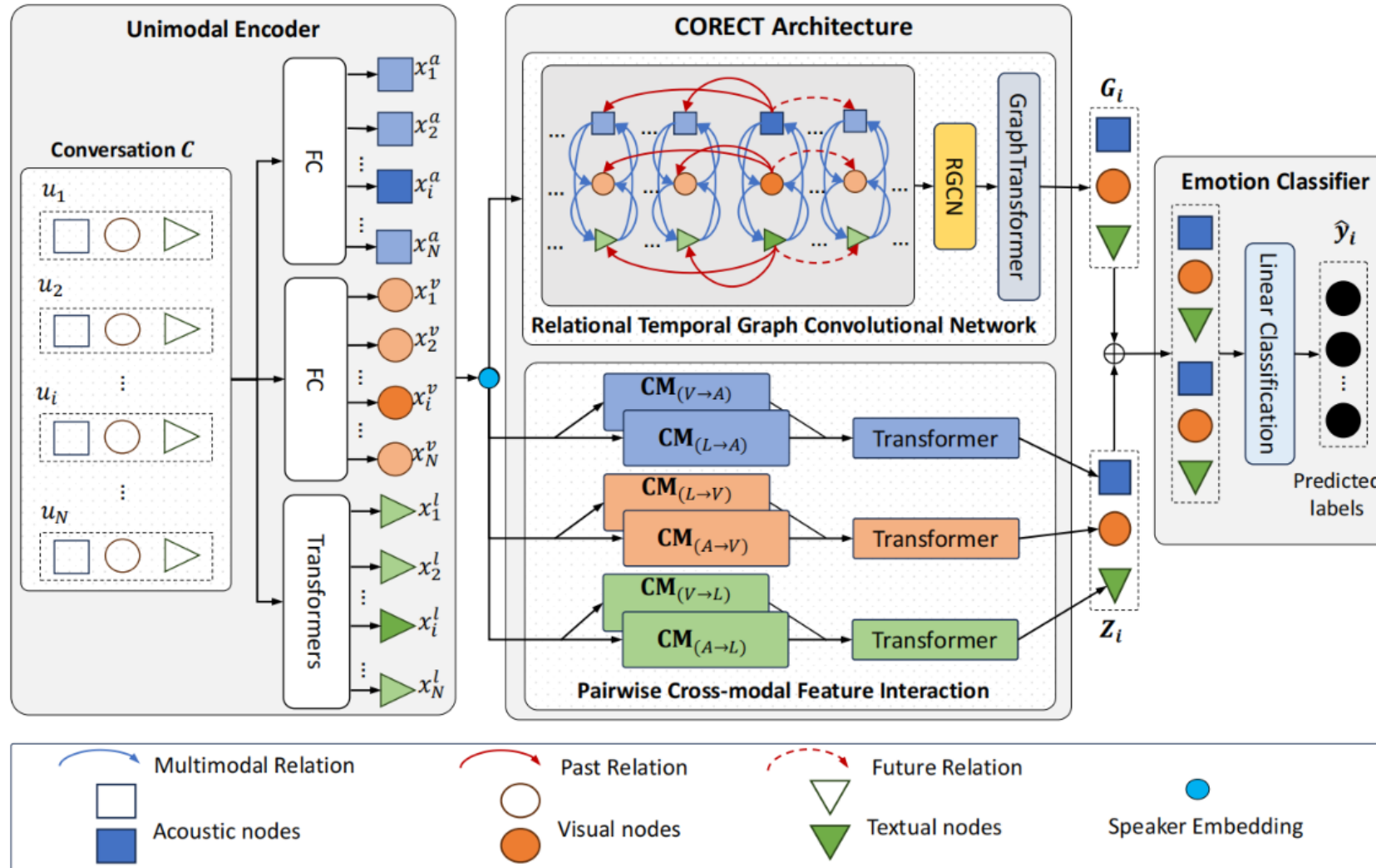
Joe→Pho —

Cha→Cha —

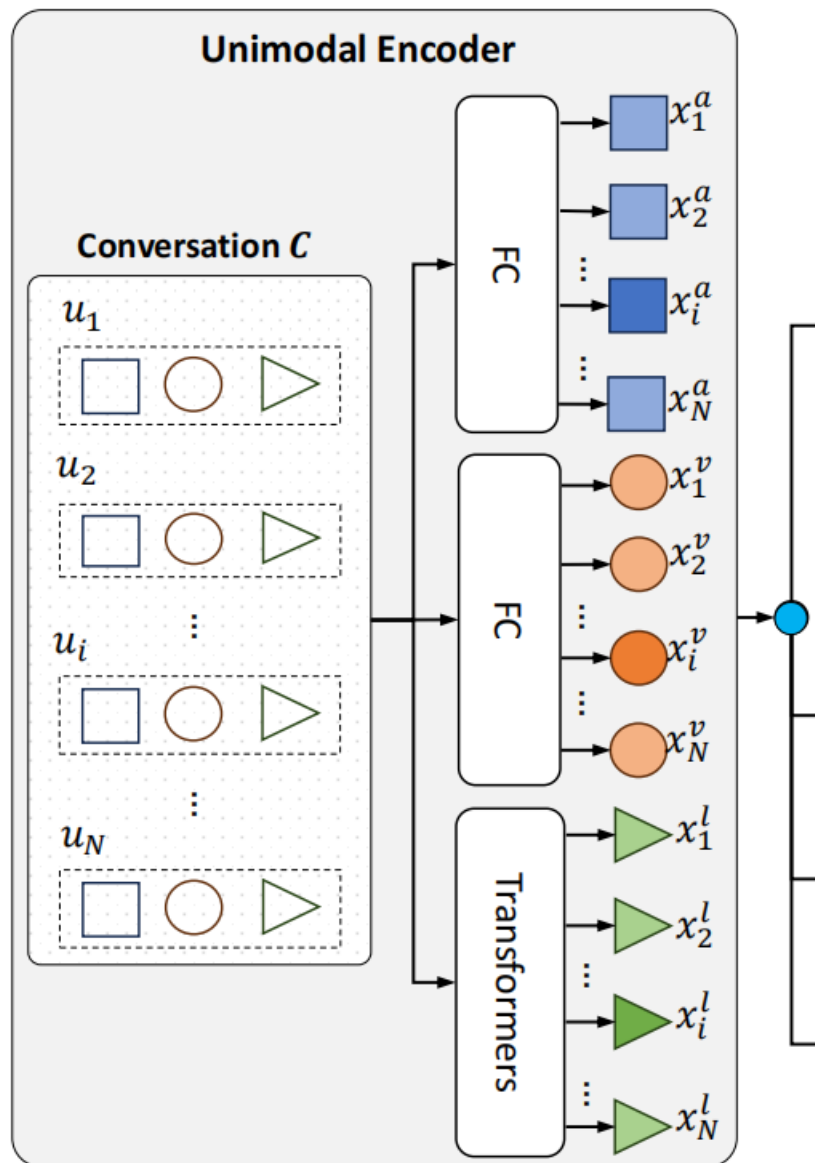
Pho→Cha —



Overview



Method



audio (a), visual (v), and textual (l)

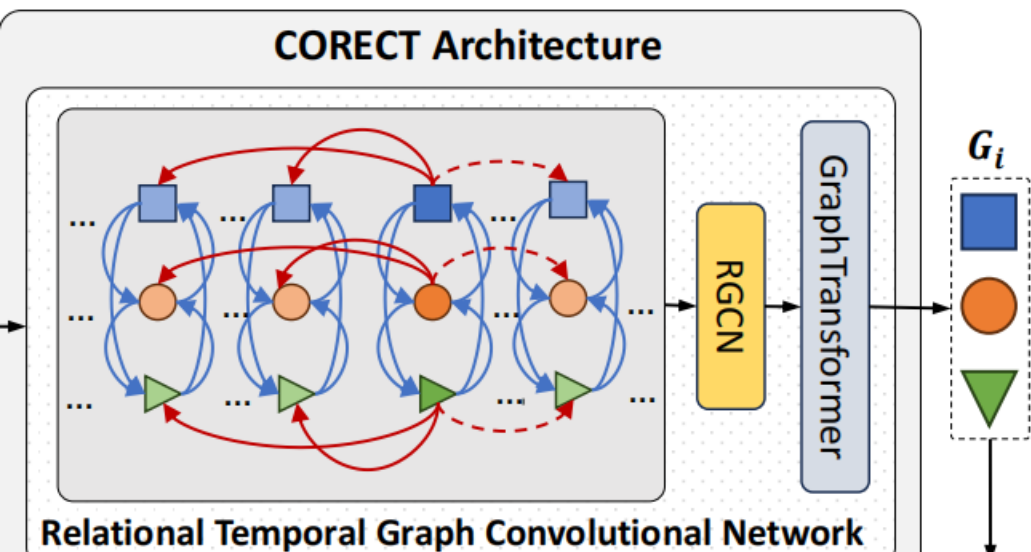
$$x_i^l = \mathbf{Transformer}(u_i^l, \mathbf{W}_{trans}^l) \quad (1)$$

$$x_i^\tau = \mathbf{FC}(u_i^\tau; \mathbf{W}_{fc}^\tau), \tau \in \{a, v\} \quad (2)$$

$$\mathcal{S}_{emb} = \mathbf{Embedding}(S, \mathcal{N}_S) \quad (3)$$

$$\mathbf{X}_\tau = \eta \mathcal{S}_{emb} + \mathcal{X}_\tau, \tau \in \{a, v, l\} \quad (4)$$

Method



$$\mathcal{R}_{multi} = \begin{cases} \{(u_i^a, u_i^v), (u_i^v, u_i^a), (u_i^a, u_i^a)\} \\ \{(u_i^v, u_i^l), (u_i^l, u_i^v), (u_i^v, u_i^v)\} \\ \{(u_i^l, u_i^a), (u_i^a, u_i^l), (u_i^l, u_i^l)\} \end{cases} \quad (5)$$

$$\mathcal{R}_{temp} = \begin{cases} \{(u_j \xrightarrow{\text{past}} u_i)^\tau \mid i - \mathcal{P} < j < i\} \\ \{(u_i \xleftarrow{\text{future}} u_j)^\tau \mid i < j < i + \mathcal{F}\} \end{cases} \quad (6)$$

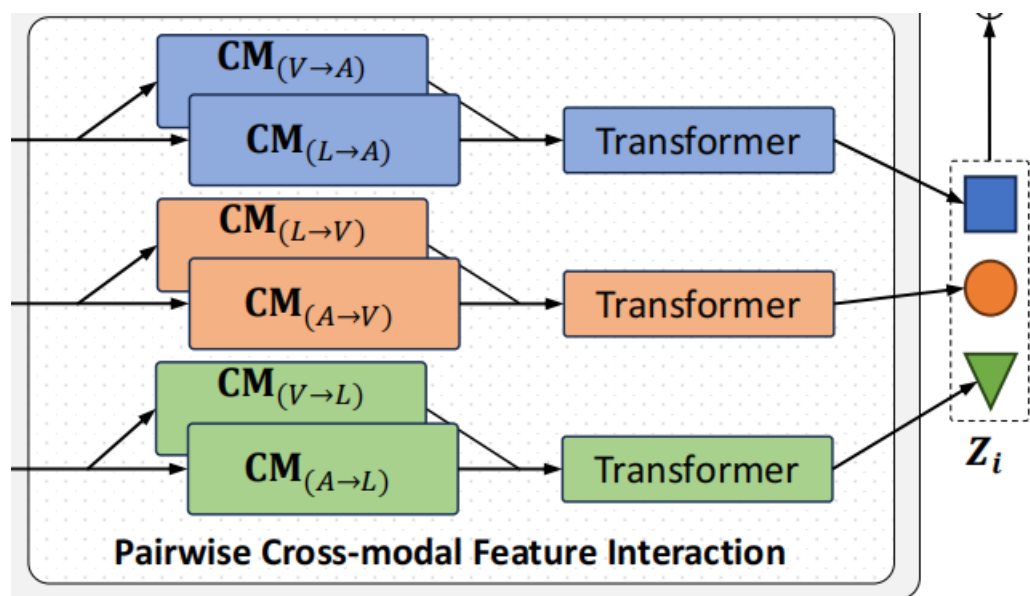
$$g_i^\tau = \sum_{r \in \mathcal{R}} \sum_{j \in \mathcal{N}_r(i)} \frac{1}{|\mathcal{N}_r(i)|} \mathbf{W}_r \cdot x_j^\tau + \mathbf{W}_0 \cdot x_i^\tau \quad (7)$$

$$o_i^\tau = \|\|_{c=1}^C [\mathbf{W}_1 g_i^\tau + \sum_{j \in \mathcal{N}(i)} \alpha_{i,j}^\tau \mathbf{W}_2 g_j^\tau] \quad (8)$$

$$\alpha_{i,j}^\tau = \text{softmax} \left(\frac{(\mathbf{W}_3 g_i^\tau)^\top (\mathbf{W}_4 g_j^\tau)}{\sqrt{d}} \right) \quad (9)$$

$$\mathbf{G}^\tau = \{o_1^\tau, o_2^\tau, \dots, o_N^\tau\} \quad (10)$$

Method



$$\mathbf{CM}^{l \rightarrow a} = \sigma \left(\frac{\mathbf{X}^a \mathbf{W}_{Q^a} (\mathbf{W}_{K^l})^\top (\mathbf{X}^l)^\top}{\sqrt{d_k}} \right) \mathbf{X}^l \mathbf{W}_{V^l} \quad (11)$$

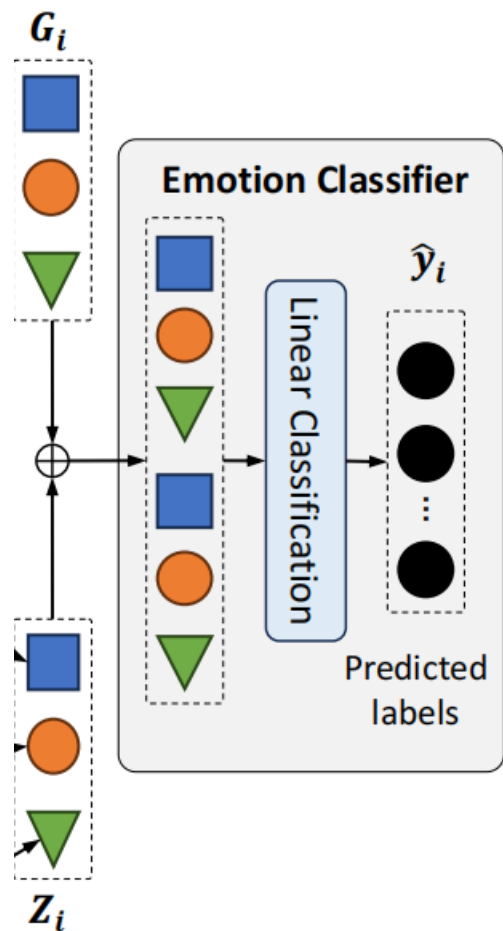
$$\mathbf{Z}_{[0]}^{l \rightarrow a} = \mathbf{Z}_{[0]}^a$$

$$\begin{aligned} \bar{\mathbf{Z}}_{[i]}^{l \rightarrow a} &= \mathbf{CM}_{[i]}^{l \rightarrow a} (LN(\mathbf{Z}_{[i-1]}^{l \rightarrow a}), LN(\mathbf{Z}_0^{l \rightarrow a})) \\ &+ LN(\mathbf{Z}_{[i-1]}^{l \rightarrow a}) \end{aligned}$$

$$\mathbf{Z}_{[i]}^{l \rightarrow a} = (LN(\bar{\mathbf{Z}}_{[i]}^{l \rightarrow a}))^{FFN} + LN(\bar{\mathbf{Z}}_{[i]}^{l \rightarrow a}) \quad (12)$$

$$\begin{aligned} LN(\bar{\mathbf{Z}}_{[i]}^{l \rightarrow a})^{FFN} &= \max(0, LN(\bar{\mathbf{Z}}_{[i]}^{l \rightarrow a})) \Omega_1 \\ &+ \mathbf{b}_1) \Omega_2 + \mathbf{b}_2 \quad (13) \end{aligned}$$

Method



$$\mathbf{H} = \text{Fusion}([\mathbf{G}, \mathbf{Z}]) \quad (14)$$

$$= [\{o_1^\tau, o_2^\tau, \dots, o_N^\tau\}, \{\mathbf{Z}_{a \leftrightarrow v}^{[D]}, \mathbf{Z}_{v \leftrightarrow l}^{[D]}, \mathbf{Z}_{l \leftrightarrow a}^{[D]}\}]$$

$$v_i = \text{ReLU}(\Phi_0 h_i + b_0) \quad (15)$$

$$p_i = \text{softmax}(\Phi_1 v_i + b_1) \quad (16)$$

$$\hat{y}^i = \text{argmax}(p_i) \quad (17)$$



Experiments

Dataset	Dialogues			Utterances		
	train	valid	test	train	valid	test
IEMOCAP (6-way)	108	12	31	5,146	664	1,623
IEMOCAP (4-way)	108	12	31	3,200	400	943
MOSEI	2,249	300	646	16,327	1,871	4,662



Experiments

Methods	IEMOCAP (6-way)						Acc. (%)	w-F1 (%)
	Happy	Sad	Neutral	Angry	Excited	Frustrated		
bc-LSTM (Poria et al., 2017)	32.63	70.34	51.14	63.44	67.91	61.06	59.58	59.10
CMN (Hazarika et al., 2018b)	30.38	62.41	52.39	59.83	60.25	60.69	56.56	56.13
ICON (Hazarika et al., 2018a)	29.91	64.57	57.38	63.04	63.42	60.81	59.09	58.54
DialogueRNN (Majumder et al., 2019)	33.18	78.80	59.21	65.28	71.86	58.91	63.40	62.75
DialogueGCN (Ghosal et al., 2019)	47.10	80.88	58.71	66.08	70.97	61.21	65.54	65.04
MMGCN (Wei et al., 2019)	45.45	77.53	61.99	<u>66.70</u>	72.04	<u>64.12</u>	65.56	65.71
DialogueCRN (Hu et al., 2021)	51.59	74.54	62.38	67.25	73.96	59.97	65.31	65.34
COGMEN (Joshi et al., 2022)	<u>55.76</u>	80.17	<u>63.21</u>	61.69	74.91	63.90	<u>67.04</u>	<u>67.27</u>
CORECT (Ours)	59.30	<u>80.53</u>	66.94	69.59	<u>72.69</u>	68.50	69.93 (↑ 2.89)	70.02 (↑ 2.75)



Experiments

Modality Settings	IEMOCAP (4-way)	
	Acc. (%)	w-F1 (%)
bc-LSTM (Poria et al., 2017)	75.20	75.13
CHFusion (Majumder et al., 2018)	76.59	76.80
COGMEN (Joshi et al., 2022)	<u>82.29</u>	<u>82.15</u>
CORECT (Ours)	84.73 (↑ 2.44)	84.64 (↑ 2.49)



Experiments

Methods	Sentiment Classification Accuracy (%)		Emotion Classification (Binary, 1 vs. all) weighted F1-score (%)					
	2 Class	7 Class	Happiness	Sadness	Angry	Fear	Disgust	Surprise
Multilouge-Net (Shenoy and Sardana, 2020)	82.88	44.83	67.84	65.34	67.03	87.79	74.91	86.05
TBJE (Delbrouck et al., 2020)	82.40	43.91	65.91	70.78	70.86	87.79	<u>82.57</u>	86.04
COGMEN (Joshi et al., 2022)	<u>82.95</u>	<u>45.22</u>	<u>70.88</u>	<u>70.91</u>	<u>74.20</u>	87.79	81.83	86.05
CORECT (Ours)	83.66	46.31	71.35	72.86	76.77	87.90	84.26	86.48



Experiments

Modality Settings	IEMOCAP (6-way)		IEMOCAP (4-way)	
	Acc. (%)	w-F1 (%)	Acc. (%)	w-F1 (%)
A	52.31	51.49	67.02	65.48
T	67.22	67.26	82.82	82.65
V	38.63	37.67	49.73	47.97
A+T	<u>68.27</u>	<u>68.36</u>	<u>83.14</u>	<u>83.13</u>
T+V	65.50	65.61	81.76	81.75
V+A	54.16	53.82	69.03	68.21
CORECT (A+T+V)	69.93	70.02	84.73	84.64



Thanks!