



HETERMPC: A Heterogeneous Graph Neural Network for Response Generation in Multi-Party Conversations

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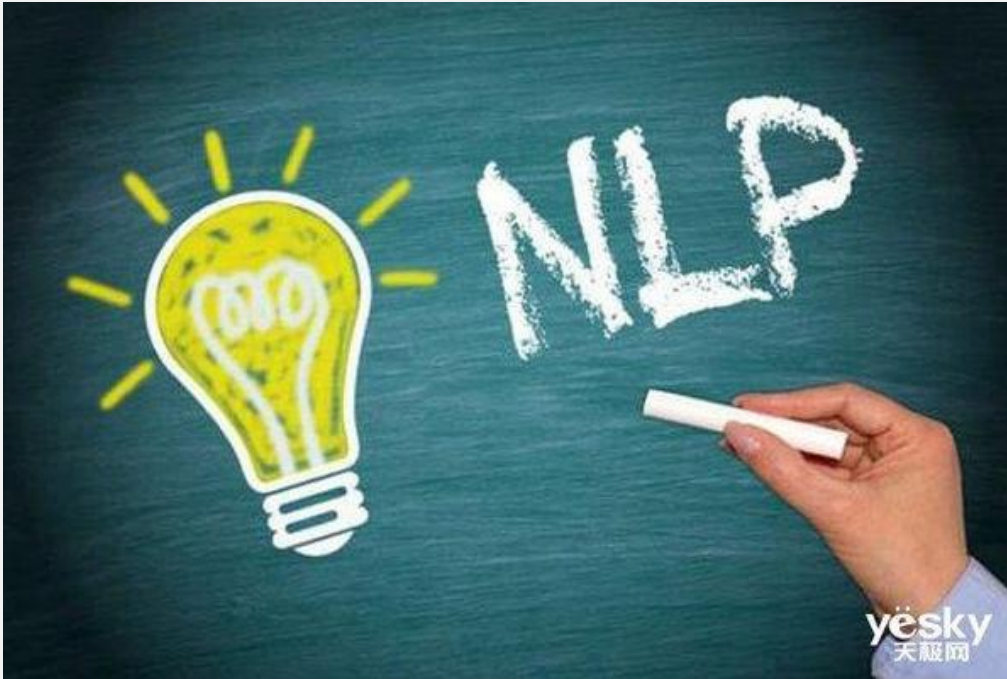
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Reported by Jia Wang



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Introduction

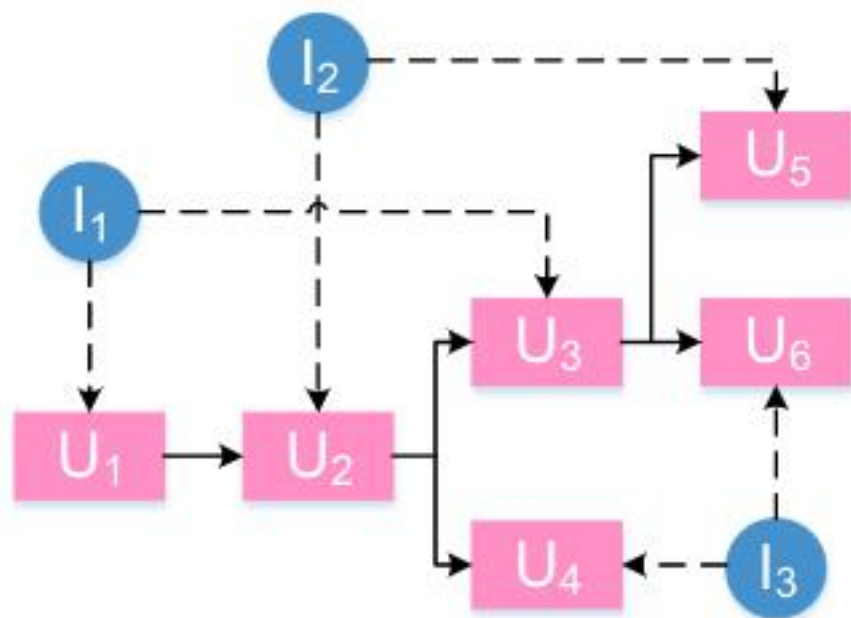


Figure 1: Illustration of a graphical information flow in an MPC. Pink rectangles denote utterances and blue circles denote interlocutors. Each solid line represents the “*replied-by*” relationship between two utterances. Each dashed line indicates the speaker of an utterance.



Introduction

In summary, our contributions in this paper are three-fold:

- To the best of our knowledge, this paper is the first exploration of using **heterogeneous graphs** for modeling conversations;
- A Transformer-based **heterogeneous graph architecture** is introduced for response generation in MPCs, in which two types of nodes, six types of meta relations, and node-edge-type-dependent parameters are employed to characterize the heterogeneous properties of MPCs;
- Experimental results show that our proposed model achieves a new state-of-the-art performance of response generation in MPCs on the Ubuntu IRC benchmark.

Approach

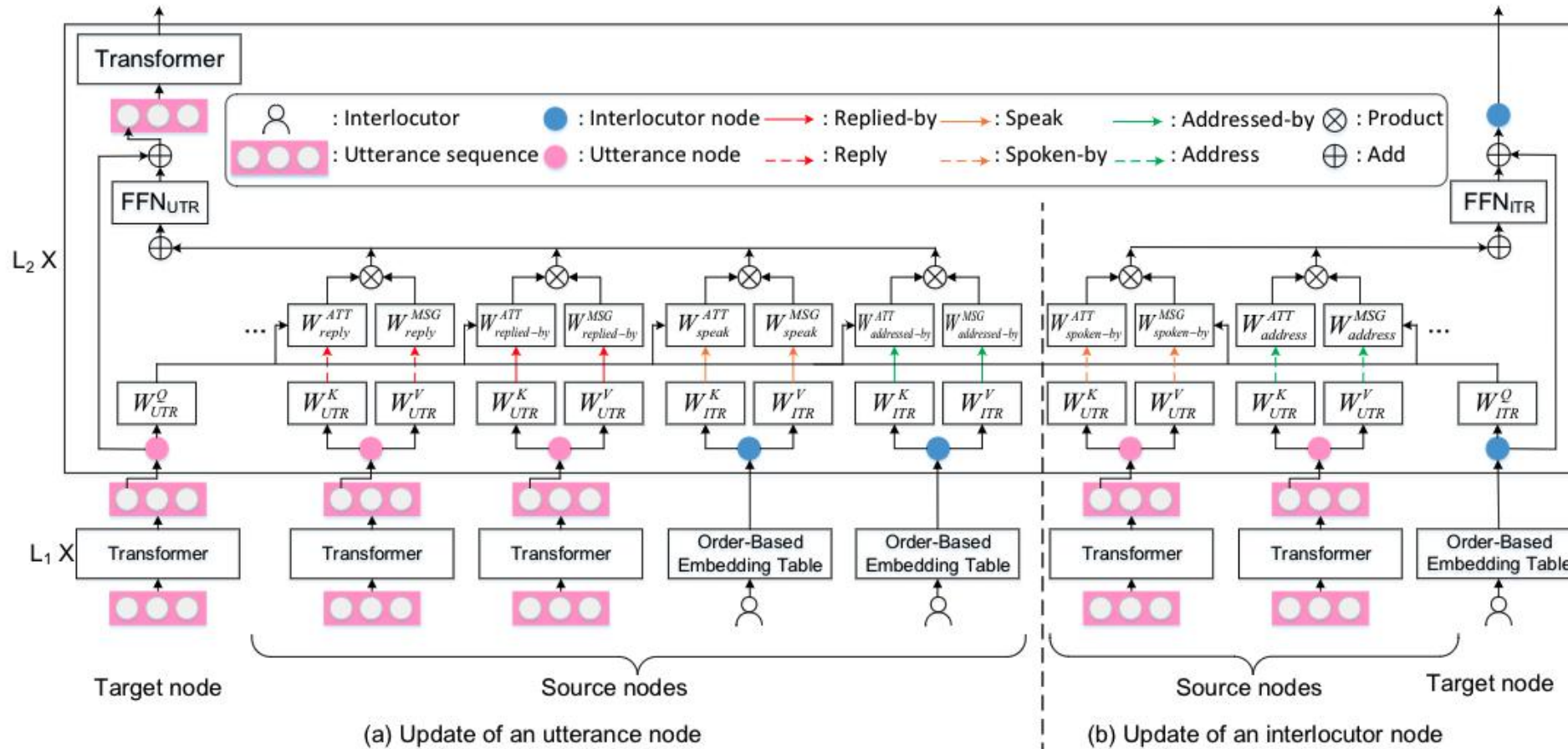


Figure 3: Model architecture of HeterMPC for (a) update of an utterance node and (b) update of an interlocutor node. “UTR” and “ITR” are abbreviations of “utterance” and “interlocutor” respectively. The set of W_* denotes the node-edge-type-dependent parameters.

Graph Construction

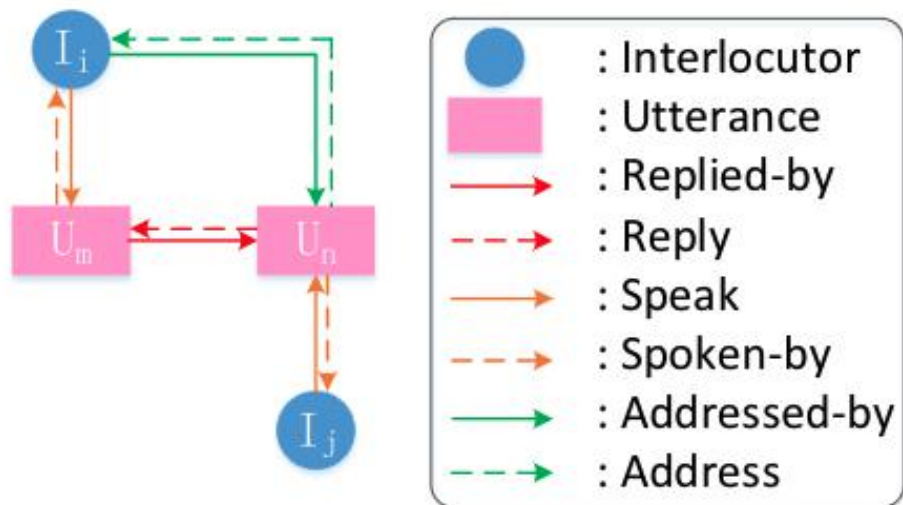
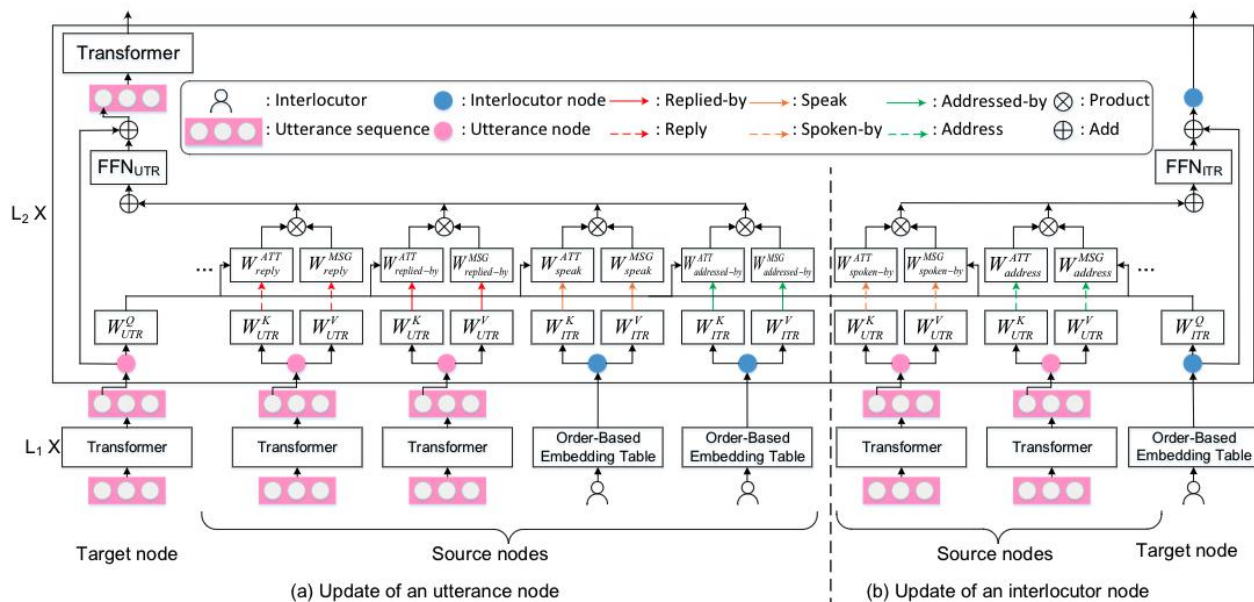


Figure 2: Illustration of the two types of nodes and six types of edges in a heterogeneous conversation graph. This example demonstrates that I_j speaks U_n replying U_m that is spoken-by I_i .

Given an MPC instance composed of M utterances and I interlocutors, a heterogeneous graph $\mathbb{G}(\mathbb{V}, \mathbb{E})$ is constructed. Specifically, \mathbb{V} is a set of $M + I$ nodes. Each node denotes either an utterance or an interlocutor. $\mathbb{E} = \{e_{p,q}\}_{p,q=1}^{M+I}$ is a set of directed edges. Each edge $e_{p,q}$ describes the connection from node p to node q .

Approach

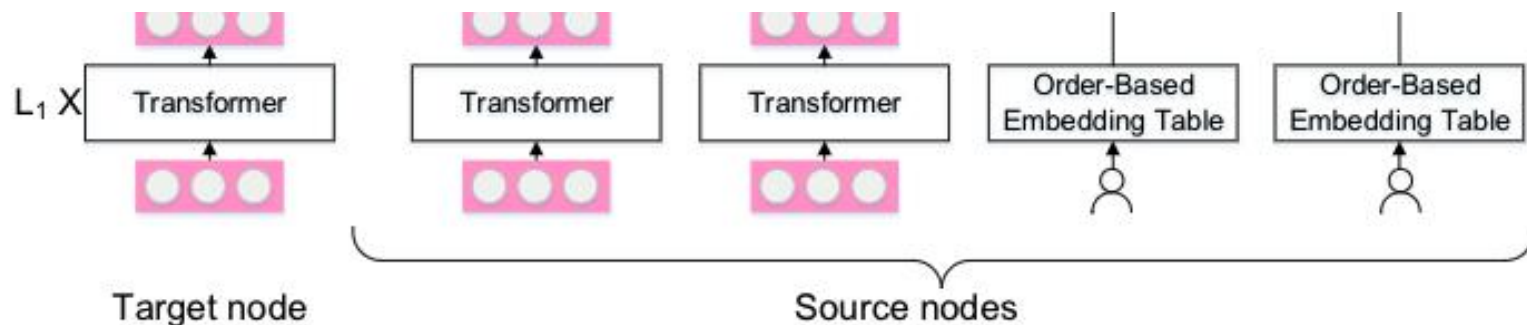
Node Initialization



Utterances:

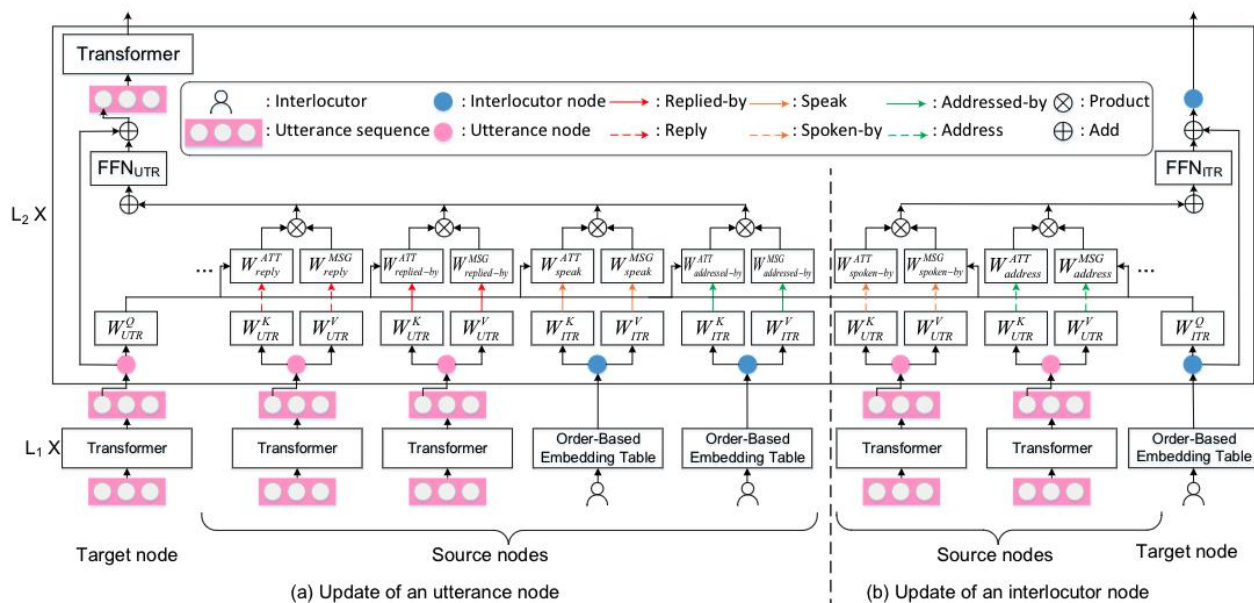
$$\mathbf{H}_m^{l+1} = \text{TransformerEncoder}(\mathbf{H}_m^l), \quad (2)$$

where $m \in \{1, \dots, M\}$, $l \in \{0, \dots, L_1 - 1\}$, L_1 denotes the number of Transformer layers for initialization, $\mathbf{H}_m^l \in \mathbb{R}^{k_m \times d}$, k_m denotes the length of an utterance and d denotes the dimension of embedding vectors.



Approach

Node Updating: Heterogeneous Attention



$$\mathbf{k}^l(s) = \mathbf{h}_s^l \mathbf{W}_{\tau(s)}^K + \mathbf{b}_{\tau(s)}^K, \quad (3)$$

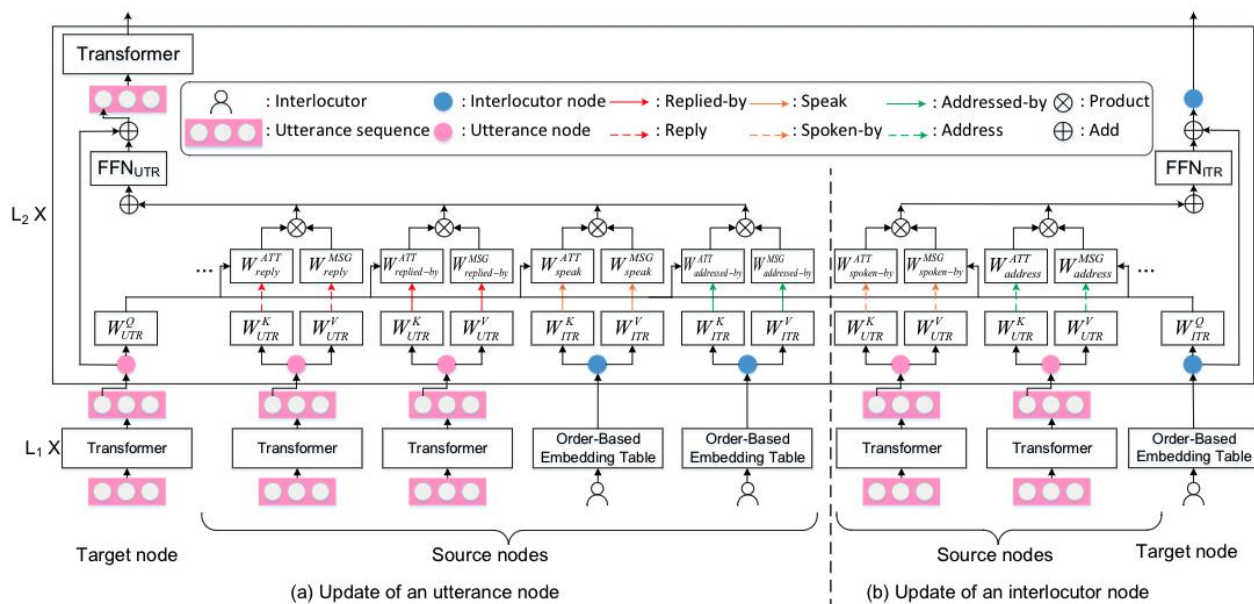
$$\mathbf{q}^l(t) = \mathbf{h}_t^l \mathbf{W}_{\tau(t)}^Q + \mathbf{b}_{\tau(t)}^Q, \quad (4)$$

$$w^l(s, e, t) = \mathbf{k}^l(s) \mathbf{W}_{e_s, t}^{ATT} \mathbf{q}^l(t)^T \frac{\mu_{e_s, t}}{\sqrt{d}}. \quad (5)$$

Here, $\tau(s), \tau(t) \in \{UTR, ITR\}$ distinguish utterance (*UTR*) and interlocutor (*ITR*) nodes. Eqs. (3) and (4) are node-type-dependent linear transformations. Eq. (5) contains an edge-type-dependent linear projection $\mathbf{W}_{e_s, t}^{ATT}$ where $\mu_{e_s, t}$ is an adaptive factor scaling to the attention. All $\mathbf{W}^* \in \mathbb{R}^{d \times d}$ and $\mathbf{b}^* \in \mathbb{R}^d$ are parameters to be learnt.

Approach

Node Updating: Heterogeneous Message Passing



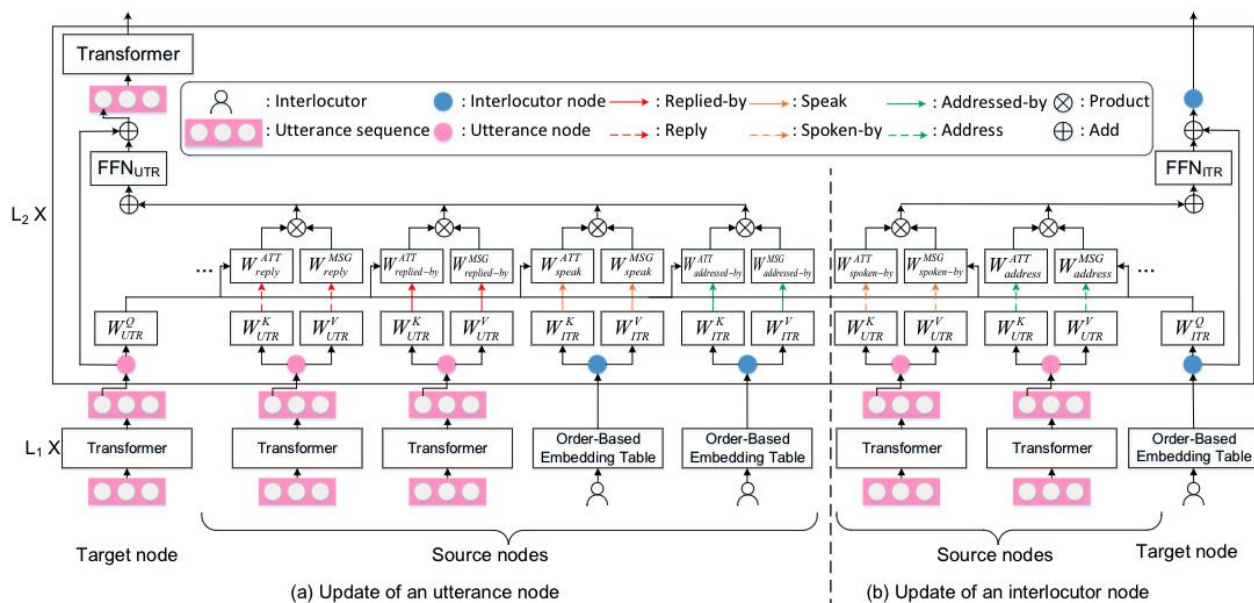
$$\mathbf{v}^l(s) = \mathbf{h}_s^l \mathbf{W}_{\tau(s)}^V + \mathbf{b}_{\tau(s)}^V, \quad (6)$$

$$\bar{\mathbf{v}}^l(s) = \mathbf{v}^l(s) \mathbf{W}_{e_{s,t}}^{MSG}, \quad (7)$$

where $\bar{\mathbf{v}}^l(s)$ is the passed message and all $\mathbf{W}^* \in \mathbb{R}^{d \times d}$ and $\mathbf{b}^* \in \mathbb{R}^d$ are parameters to be learnt.

Approach

Node Updating: Heterogeneous Aggregation



$$\bar{h}_t^l = \sum_{s \in S(t)} \text{softmax}(w^l(s, e, t)) \bar{v}^l(s), \quad (8)$$

$$h_t^{l+1} = FFN_{\tau(t)}(\bar{h}_t^l) + h_t^l, \quad (9)$$

$$\hat{h}_t^{l+1} = [h_t^l; h_t^{l+1}] \mathbf{W}_{com} + \mathbf{b}_{com}, \quad (10)$$

where $\mathbf{W}_{com} \in \mathbb{R}^{2d \times d}$ and $\mathbf{b}_{com} \in \mathbb{R}^d$ are parameters. Then, \hat{h}_t^{l+1} replaces the representation of [CLS] (i.e., h_t^l) in the sequence representations of the whole utterance.

Approach

Decoder

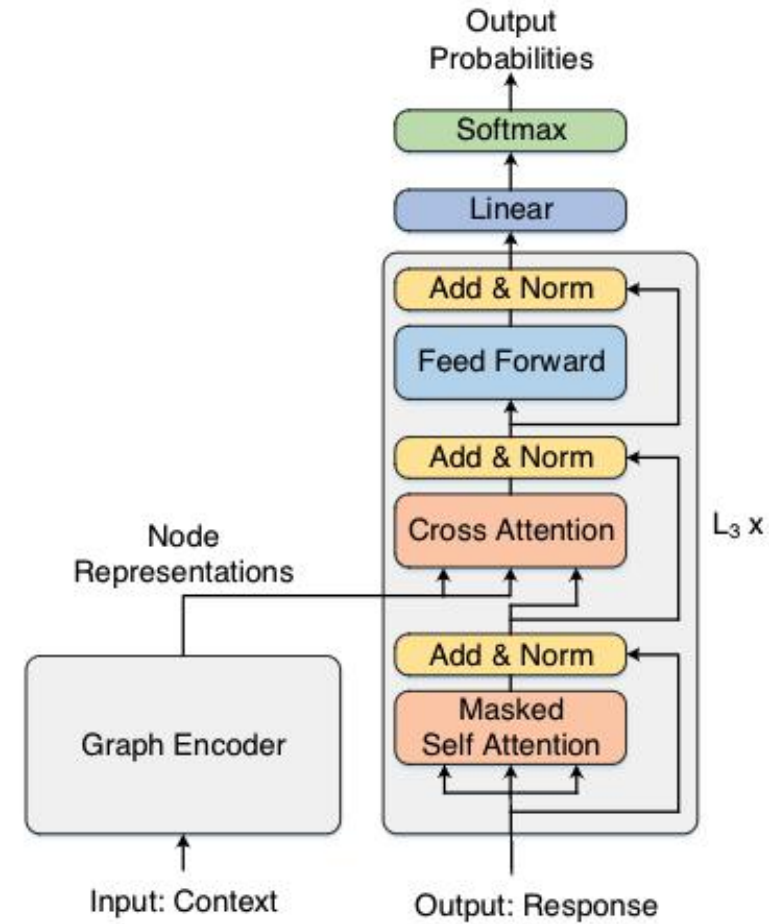
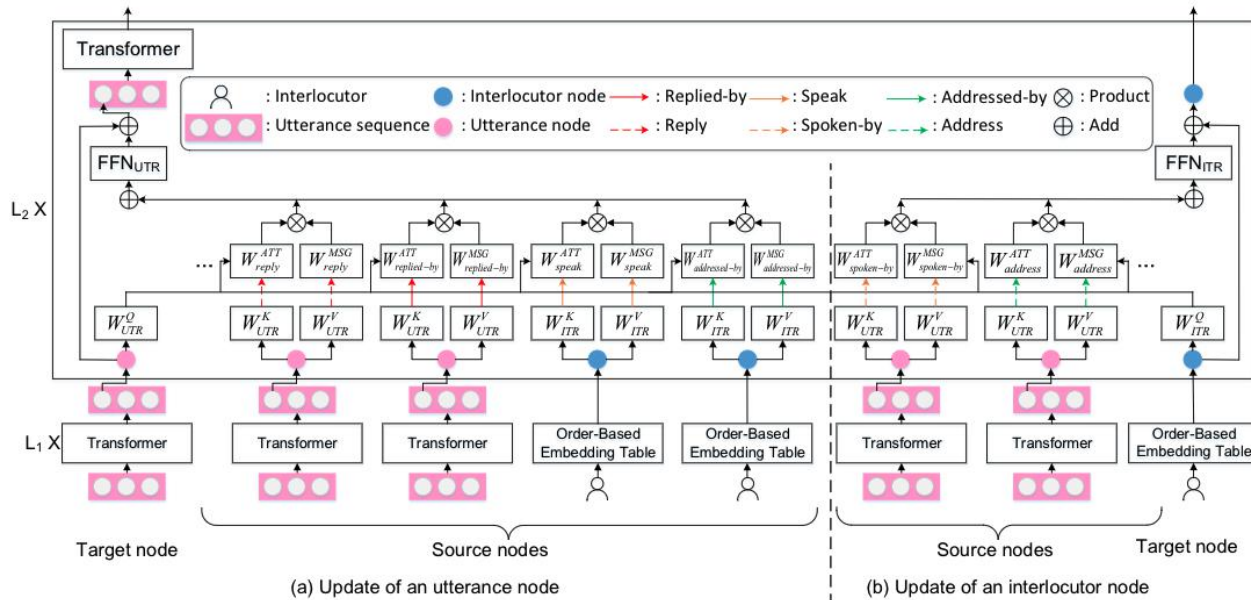


Figure 4: The decoder architecture of HeterMPC.



Experiments

Metrics Models	BLEU-1	BLEU-2	BLEU-3	BLEU-4	METEOR	ROUGE _L
Seq2Seq (LSTM) (Sutskever et al., 2014)	7.71	2.46	1.12	0.64	3.33	8.68
Transformer (Vaswani et al., 2017)	7.89	2.75	1.34	0.74	3.81	9.20
GSN (Hu et al., 2019b)	10.23	3.57	1.70	0.97	4.10	9.91
GPT-2 (Radford et al., 2019)	10.37	3.60	1.66	0.93	4.01	9.53
BERT (Devlin et al., 2019)	10.90	3.85	1.69	0.89	4.18	9.80
HeterMPC _{BERT}	12.61	4.55	2.25	1.41	4.79	11.20
HeterMPC _{BERT} w/o. node types	11.76	4.09	1.87	1.12	4.50	10.73
HeterMPC _{BERT} w/o. edge types	12.02	4.27	2.10	1.30	4.74	10.92
HeterMPC _{BERT} w/o. node and edge types	11.60	3.98	1.90	1.18	4.20	10.63
HeterMPC _{BERT} w/o. interlocutor nodes	11.80	3.96	1.75	1.00	4.31	10.53
BART (Lewis et al., 2020)	11.25	4.02	1.78	0.95	4.46	9.90
HeterMPC _{BART}	12.26	4.80	2.42	1.49	4.94	11.20
HeterMPC _{BART} w/o. node types	11.22	4.06	1.87	1.04	4.57	10.63
HeterMPC _{BART} w/o. edge types	11.52	4.27	2.05	1.24	4.78	10.90
HeterMPC _{BART} w/o. node and edge types	10.90	3.90	1.79	1.01	4.52	10.79
HeterMPC _{BART} w/o. interlocutor nodes	11.68	4.24	1.91	1.03	4.79	10.65

Table 1: Performance of HeterMPC and ablations on the test set in terms of automated evaluation. Numbers in bold denote that the improvement over the best performing baseline is statistically significant (t-test with p -value < 0.05).



Experiments

Models \ Metrics	Score	Kappa
Human	2.81	0.55
GSN (Hu et al., 2019b)	2.00	0.50
BERT (Devlin et al., 2019)	2.19	0.42
BART (Lewis et al., 2020)	2.24	0.44
HeterMPC _{BERT}	2.39	0.50
HeterMPC _{BART}	2.41	0.45

Table 2: Human evaluation results of HeterMPC and some selected systems on a randomly sampled test set.

Experiments

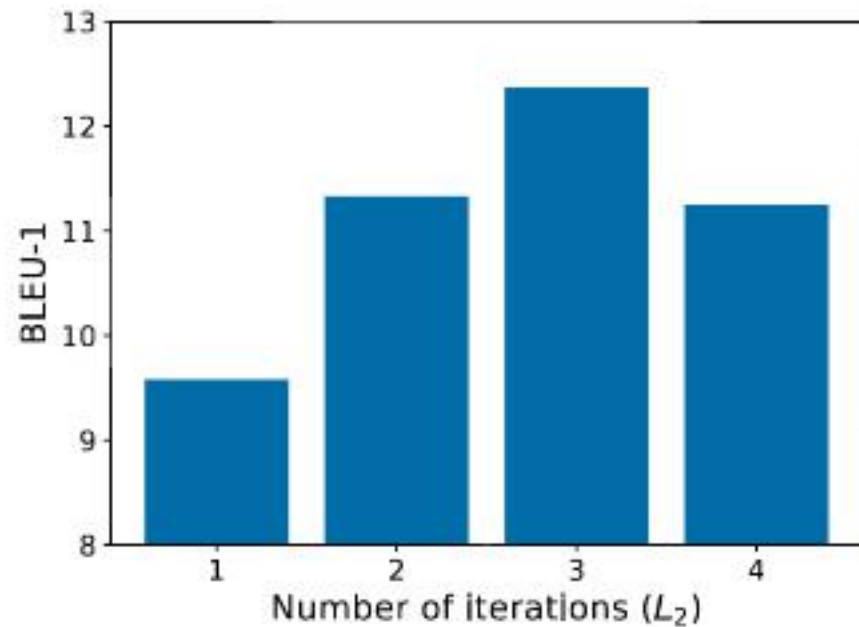


Figure 5: Performance of HeterMPC_{BERT} under different numbers of iterations (L_2) on the test set.

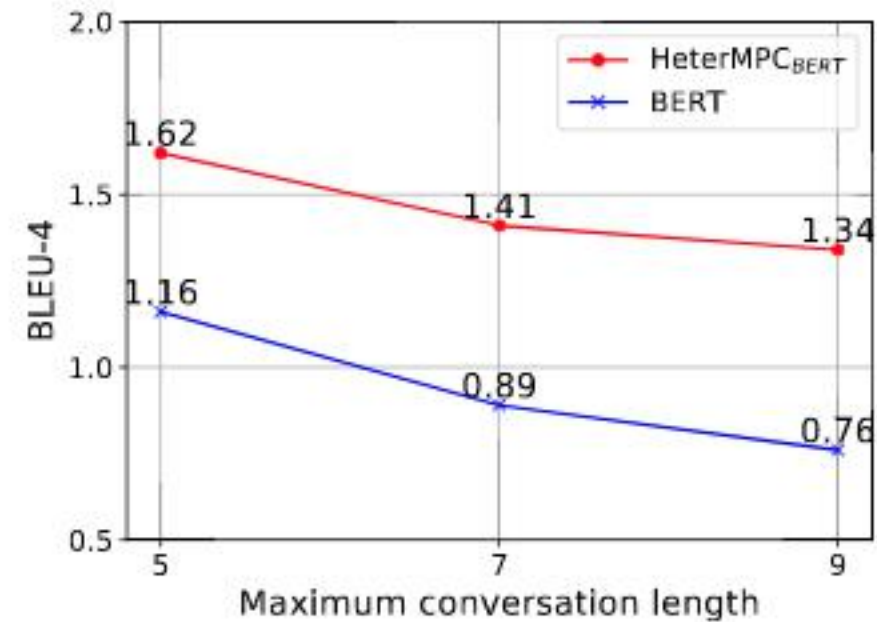


Figure 6: Performance of BERT and HeterMPC_{BERT} on test samples with different session lengths.

Experiments

Speaker	Utterance	Addressee
Case 1		
I.1	geev: in gparted now?	-
I.2	there is no such command in my computer	I.1
I.1	open a terminal, type: sudo apt-get install gparted	I.2
I.1	after: gparted aksks for your pw, opens a gui	I.2
I.2	ok if found fail i will contact you	I.1
I.3	does it say "to install type sudo apt-get install gparted"? (Human)	I.2
	do you have remote access to remote desktop? (GSN)	
	ok, i will try that. (BERT)	
	you can use the command "sudoapt-get install gparted" to install the gparted livecd. (HeterMPC_{BERT})	
	i'm not sure what you mean by "in gparted" (BART)	
	you need to install gparted, it is a gui tool (HeterMPC_{BART})	

Case 2		
I.1	got a bit of a mount problem. fstab is still the same but the secondary harddrive shows as busy when doing mount	-
I.2	you can see what is mounted with : mount	I.1
I.1	yes it is not mounted	I.2
I.3	how would i check what module the cam uses?	I.2
I.2	lsmod less will show you te loaded modules	I.3
I.1	lmount - t ext3 filepath fails cause already mounted or busy	I.2
I.2	then try to mount it at command line (Human)	I.1
	i'm not sure how to do that (GSN)	
	i'm not sure what the problem is (BERT)	
	you need to mount it as a mount point (HeterMPC_{BERT})	
	i'm not sure what the problem is (BART)	
	you need to check the filepath file (HeterMPC_{BART})	



Thanks !