



Dialogue Discourse-Aware Graph Model and Data Augmentation for Meeting Summarization



Xiachong Feng, Xiaocheng Feng, Bing Qin, Xinwei Geng
Harbin Institute of Technology

Introduction

• Meeting Summarization

• Distill the most important information from a recorded meeting into a short textual passage.

• Problems

• Sequential text modeling: hinder the exploration of rich interactive relations between utterances.

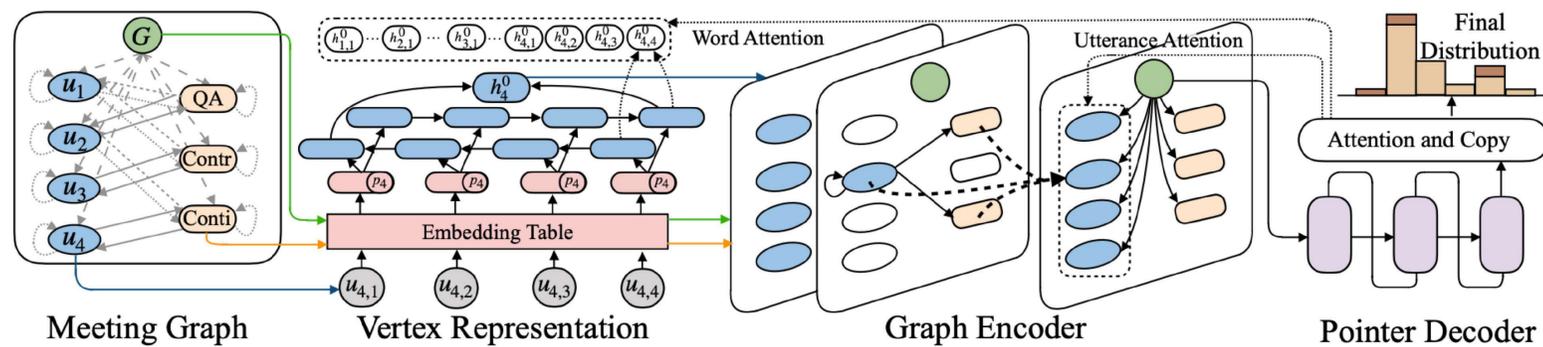
• Lack of sufficient training data: hinder the ability of data-hungry neural models.

• Solution: dialogue discourse

• Dialogue Discourse-Aware Meeting Summarizer: model utterances and discourse in a graph interaction manner.

• Dialogue Discourse-Aware Data Augmentation: construct a pseudo-summarization corpus based on QA discourse.

Dialogue Discourse-Aware Meeting Summarizer

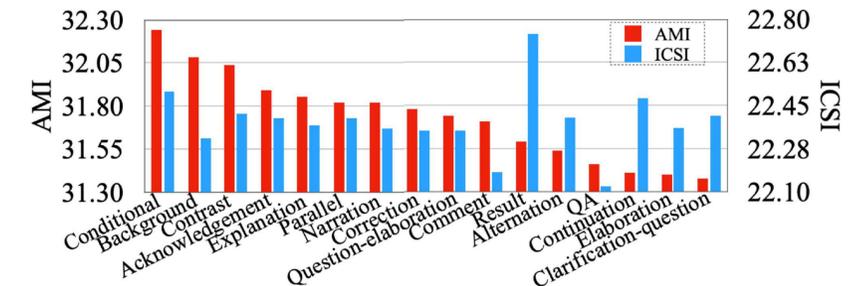
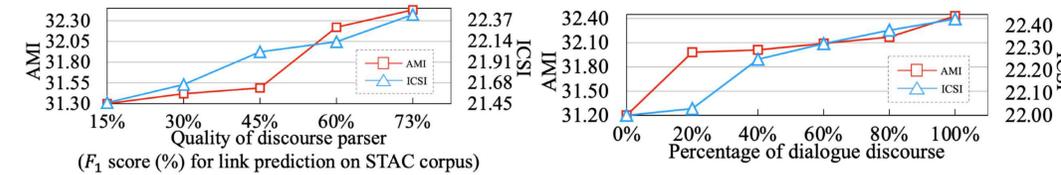


Experiments

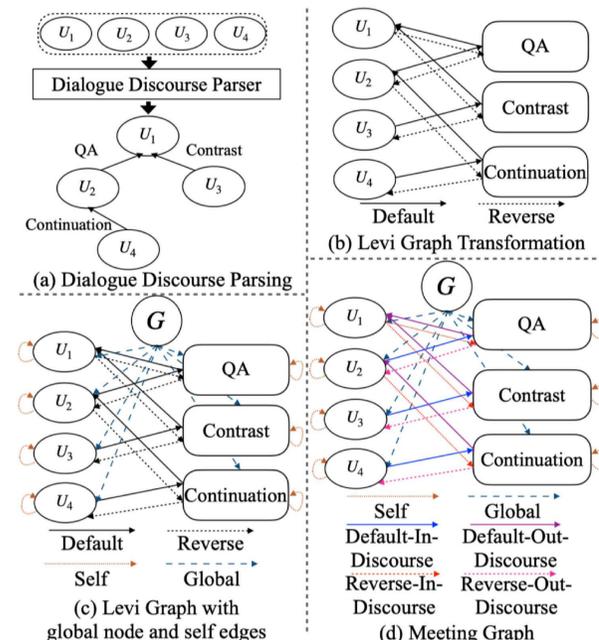
Automatic Evaluation

Model	R-1	AMI		R-L	ICSI		
		R-2	R-L		R-1	R-2	R-L
Extractive	TextRank [Mihalcea and Tarau, 2004]	35.19	6.13	15.70	30.72	4.69	12.97
	SummaRunner [Nallapati et al., 2017]	30.98	5.54	13.91	27.60	3.70	12.52
Abstractive	UNS [Shang et al., 2018]	37.86	7.84	13.72	31.73	5.14	14.50
	Pointer-Generator [See et al., 2017]	42.60	14.01	22.62	35.89	6.92	15.67
	HRED [Serban et al., 2016]	49.75	18.36	23.90	39.15	7.86	16.25
	Sentence-Gated [Goo and Chen, 2018]	49.29	19.31	24.82	39.37	9.57	17.17
	TopicSeg [Li et al., 2019]	51.53	12.23	25.47	-	-	-
	HMNet [Zhu et al., 2020]	52.36	18.63	24.00	45.97	10.14	18.54
Ours	DDAMS	51.42	20.99	24.89	39.66	10.09	17.53
	DDAMS + DDADA	53.15	22.32	25.67	40.41	11.02	19.18
	DDAMS + DDADA (w/o fine-tune)	28.35	4.67	14.92	25.94	4.18	13.92

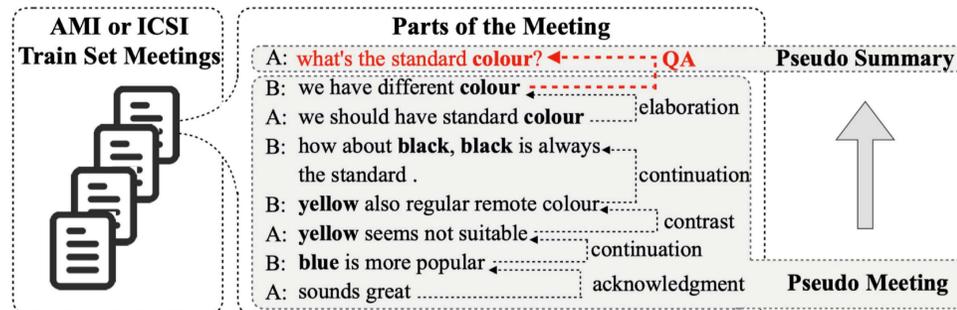
Analyses



Meeting Graph



Dialogue Discourse-Aware Data Augmentation



Conclusion

- We make the first attempt to successfully explore dialogue discourse to model the utterances interactions for meeting summarization;
- We devise a dialogue discourse-aware data augmentation strategy to alleviate the data insufficiency problem;
- Extensive experiments show that our model achieves SOTA performance.

Effect of Meeting Graph

Model	R-1	R-2	R-L
AMI DDAMS	51.42	20.99	24.89
AMI DDAMS (w/ Levi graph)	51.46	20.75	24.31
ICSI DDAMS	39.66	10.09	17.53
ICSI DDAMS (w/ Levi graph)	39.20	9.54	17.48

Effect of Pseudo Data

Model	R-1	R-2	R-L
AMI	DDAMS	51.42	24.89
	+ RBDA	52.94	25.05
	+ DDADA	53.15	25.67
ICSI	DDAMS	39.66	17.53
	+ RBDA	39.42	18.19
	+ DDADA	40.41	19.18

Paper



Code

