

Knowledge Graph Grounded Goal Planning for Open-Domain Conversation Generation

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Chitchat

Academia



Industry

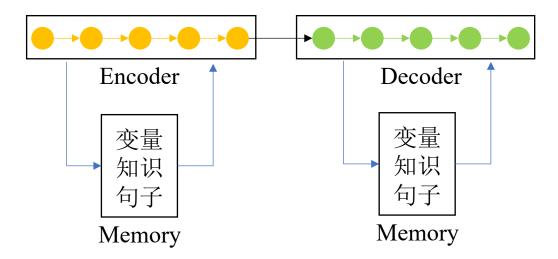


From http://www.ruiyan.me/pubs/tutorial-emnlp18.pdf



Chitchat

- Given context, generate or retrieve a response.
- Generally generated by Seq2seq based models (with memory) or retrieved by various matching networks.





Motivation

How about talking to real user?

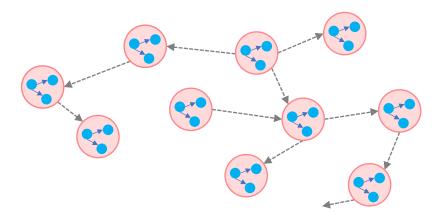
Start	80后 对此 肯定 不会 陌生, 当年 骗 父母 买 来 练 打字 的 举手!
User	举手,我就买过这样的,我觉得挺不错的
Bot	我家的是买过,不过现在还在用这个东东.老爸给了一台电脑!举手.
User	是啊,我也要举手,那时候我也给我家的老人家买过这个
Bot	我家的是开着卖,叫你买个给我们玩.哈哈哈!恭喜发财了啊.
User	好啊,恭喜发财,过年一起来打牌啊
Bot	好可爱的小刺猬,可爱死了.
User	是啊,很可爱的刺猬,就是刺太多了

• No effective mechanisms to manage chatting goals, and tend to produce less coherent and engaging dialogs.



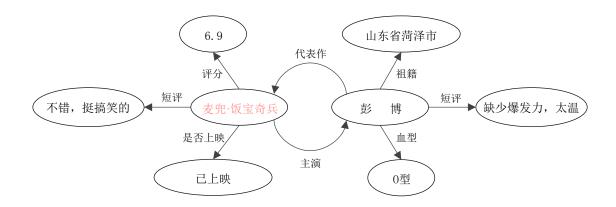
Motivation

• We introduce goal management into chitchat, to foster a more coherent and engaging multi-turn dialog.





- Goal (limited to this paper): chatting around a topic
 - Topic: a vertex in the KG
 - topic facet: an aspect of the given topic
 - 麦兜·饭宝奇兵 主演 彭博





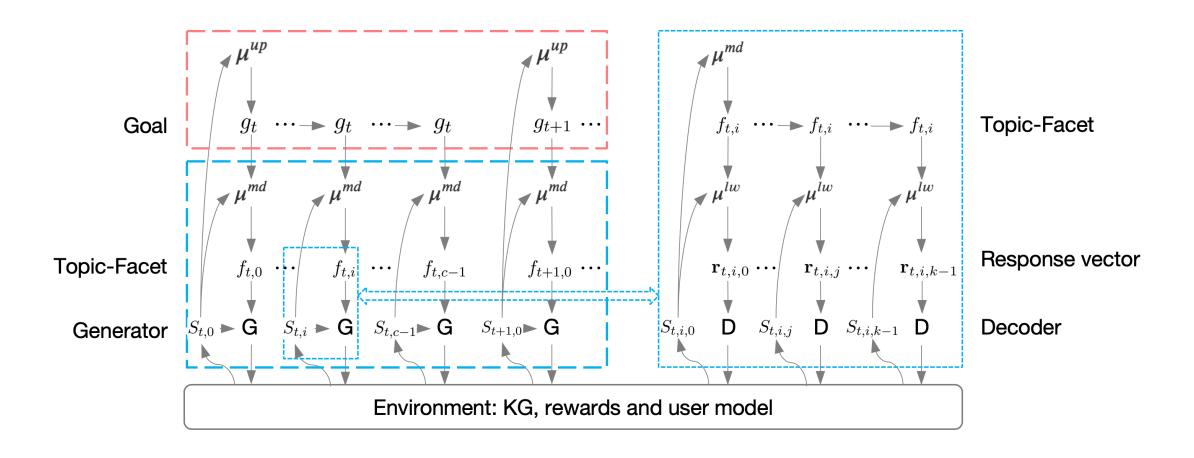
- We divide the task of multi-turn chitchat generation into two sub-tasks
 - Goal (chatting around a topic) sequence planning
 - Goal completion by topic elaboration
- And present a three-layer KG Grounded Hierarchical RL model
 - Upper-layer learns to plan topic (vertex) sequence over a KG
 - Middle-layer and Lower-layer work together to conduct topic elaboration
 - Middle-layer learns to select a neighboring vertex around the goal vertex as a topic facet
 - Lower-layer produces a multi-turn dialog based on topic and topic facet



Goal Planning	Topic Elaboration	
Upper-layer	Middle-layer	Lower-layer

Goals	Topic Facets	Utterances
<between calmness<="" td=""><td>Start_of_Session</td><td>M: Do you like love movies?</td></between>	Start_of_Session	M: Do you like love movies?
and Enthusiasm>		U: Yes, I often watch love movies recently.
		M: Then you can go to see the movie <between and<="" calmness="" td=""></between>
		Enthusiasm>(2001). It's a love movie.
		U: I've seen this movie. There are many cheaters in it. Another movie
		with the same rating is <the enough="" is="" not="" world="">(1999).</the>
<the is="" not<="" td="" world=""><td>Starring, Robert Carlyle</td><td>M: Robert Carlyle starred in this movie, whose constellation is Aries.</td></the>	Starring, Robert Carlyle	M: Robert Carlyle starred in this movie, whose constellation is Aries.
Enough>		U: Yes, he starred. His ancestral home is Glasgow, Scotland.
	Starring, Pierce Brosnan	M: The film is starred by Pierce Brosnan, who was born in May 1953.
		Do you know who it is?
		U: Of course. He starred with Robert Carlyle and won the 24th Saturn
		Award for Best Actor.







Rewards

- For Upper-layer
 - Coherence of the goal sequence: Average cosine distance with TransE
 - User interest consistency: Bot should follow new topic mentioned by users
 - Diversity: Balance between changing the goal too frequently and always sticking to the same goal
 - Sustainability: Give priority to vertices with lots of related knowledge, PageRank
 - Goal-completion information from the middle layer
- For Middle-layer
 - Topic-facet coherence
- For Lower-layer
 - Utterance relevance, Utterance informativeness, Topic-facet completion



Experiment

- Dataset: DuConv(publicly available knowledge-driven dialog dataset)
 - Consists of 30k dialogs with 120k dialogue turns
 - Provides a KG in the domain of movies and celebrities
 - 91k films and 51k film stars, and about 3.6 million knowledge triplets

Models

- CCM (Zhou et al. 2018), a state-of-the-art KG based conversation model
- CCM+LaRL (Zhao, Xie, and Eskenazi 2019), a latent variable based RL model
- KnowHRL, our model
- KnowHRL-liteReward: only use utterance relevance and informativeness as rewards.
- User simulator: multi-mapping based generation model (Chen et al. 2019)



Evaluation

- Evaluation Metrics at Session Level
 - Coherence: Intra-topic coherence, Inter-topic coherence
 - Engagement: User-interest consistency (Cons.)
 - Distinct
 - Knowledge accuracy (K.A.)
- Evaluation at Turn Level
 - Setting: Randomly sample 200 dialogs with user simulator or human.
 - Metrics: Appropriateness, Informativeness



Evaluation at Session Level

Setting: Randomly sample 100 dialogs with with user simulator or human, in which, on average, each dialog consists of 14 utterances.

Model	Intra.	Inter.	Dist-2	K.A.	Cons.
CCM	0.67	0.32	0.27	0.14	0.06
CCM+LaRL	0.94	0.44	0.32	0.17	0.09
KnowHRL	1.42	1.39	0.39	0.88	0.89
-liteReward	1.13	0.41	0.35	0.81	0.07

Model	Intra.	Inter.	Dist-2	K.A.	Cons.
CCM	0.72	0.40	0.29	0.13	0.09
CCM+LaRL	0.98	0.52	0.33	0.19	0.17
KnowHRL	1.40	1.45	0.41	0.90	0.93
-liteReward	1.16	0.46	0.38	0.84	0.13

Table 2: Results of session-level evaluations on dialogs with user simulator. KnowHRL outperforms all the baselines significantly (sign test, p-value < 0.01) in terms of all the metrics.

Table 3: Results of session-level evaluations on dialogs with human. KnowHRL outperforms all the baselines significantly (sign test, p-value < 0.01) in terms of all the metrics.

Findings:

- With the help of KG, and our strategy of divide-and-conquer, we obtain explainable dialog states and actions for KnowHRL.
- Thus, it is convenient to design goal related rewards to optimize the planning of goals and topic facets;
- These goals and topic facets can be used to guide response generation for better coherence and informativeness.



Evaluation at Turn Level

Setting: Randomly sample 200 dialogs with user simulator or human.

Models	Simu	lator	Human	
	Appr.	Infor.	Appr.	Info.
CCM	0.74	0.81	0.79	0.84
CCM+LaRL	0.77	0.78	0.81	0.81
KnowHRL	0.87	0.91	0.89	0.94
-liteReward	0.82	0.83	0.85	0.87

Table 4: Turn-level results on dialogs with user simulator and dialogs with human.

Findings: KnowHRL outperforms baselines in terms of turn-level response appropriateness and informativeness when chatting with user simulator and human.



B:	你喜欢看日本的电影么 [Do you like watching Japanese movies?]
U:	喜欢啊,你推荐几个看看呢 [Yes, could you recommend some movies?]
B:	给你推荐一部电影叫《龙猫》,是一部日本电影,你可以去看看 [My Neighbor Totoro, it's a Japanese movie. You can have a try.]
U:	好呀,有空就去看看,这个电影的口碑不错 [I'll try it whenever I have time, this movie has a good reputation.]
B:	我还知道一部日本的电影,是押井守导演的,我觉得也不错。 [I know another Japanese movie that is directed by Mamoru Oshii, which is a good movie.]
U:	你说的是哪部呀, 说说看呢, 我看下我听说过没 [Which one? Maybe I have heared of it.]
B:	《攻壳机动队2》, 我 觉得 挺 好看 的, 你 可以 看看 [Ghost in the Shell 2, I enjoy it, you can have a try.]
U:	哦, 这部电影我看过, 5年前上映的, 很不错呢 [I've seen this movie, it is a good movie that released five years ago.]
B:	这部电影的主演是大冢明夫,他是射手座的呢! [The film is starred by Akio tsuka who is a Sagittarius!]
U:	我知道有一位跟他同年出生的明星, 叫阿丽森·詹尼 [I remember another star who was born in the same year, named Allison Janney.]
B:	这是《我,花样女王》的主演吧,你知道她是什么星座吗? [She starred the <i>I, Tonya</i> , Do you know what constellation she is?]
U:	这个我倒是不清楚了,告诉我吧 [I don't know about that. Tell me, please.]
B:	是天蝎座,祖籍美国俄亥俄州 [Scorpio, and her ancestral home is Ohio, USA.]
U:	我记错了,原来是美国人啊![I'm mistaken. It turned out to be an American.]



- B: 你有没有关注中国辽宁沈阳的明星吗? [Did you take note of stars who come from Shenyang, Liaoning, China?]
- U: 哈哈, 我关注了一些, 你想说的是哪个呀 [Haha, I paid attention to some of them. Which star do you mean?]
- U: 恩恩, 我知道她呀, 是一个气质美女 [En, I know her. She's a temperamental beauty.]
- B: 她还出演过《黄金时代》呢, 是中央戏剧学院毕业的,你知道么 [She has been in *The Golden Era*, and do you know that she graduated from Central Academy of Drama.]
- U: 知道呀, 我还知道跟她同一个院校毕业的一位演员叫龚蓓苾 [Yes, I also know Gong Beiyun, an actor who graduated from the same college as her.]
- B: 嗯,人也长的不错。[Well, she's also good-looking.]
- U: 是的, 很漂亮, 也很有气质, 是金牛座的 [Yes, She's beautiful and temperamental, and she is Taurus.]
- B: 她的老公伍仕贤也是一个很有实力的人, 演技有个性的制作人呢! [Her husband, Wu Shixian, is also a very talented person. He is a producer with individuality in acting.
- B: 他还出演过一部电影,你知道吗? [He's also been in a movie, you know?]
- U: 这我倒是真不知道了,你说说看呢 [You fall me, tell me please.]
- B: 《将爱情进行到底》 , 是一部爱情片,我觉得他还可以, 推荐你去看看 [Eternal Moment, it's a a love movie which I recommend you to have a try.]
- U: 这部片子我听说过呀,看的人很多,口碑还行 [I've heard about this film. There are a lot of people watched it, and the reputation is not bad.]



Conclusion

- We propose a Knowledge graph grounded Hierarchical RL based conversational model (KnowHRL) to demonstrate how hierarchical goal planning and elaboration over a KG can facilitate chatting topic management and further response generation for chitchat.
- Results show that KnowHRL outperforms baselines in terms of dialog coherence, user interest consistency, and knowledge accuracy.
- In the future, we would like to investigate how to enrich the content of KG to cover more chatting goals.