

Towards Building More Intelligent Chatting Machines

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Joint Work with Prof. Xiaoyan Zhu



Good Examples

User: 最幸福的人不是拿到了最好的东西, 而是学会

去珍惜手上已经拥有的人。

Machine: 是的,所以我也要珍惜自己的幸福

User: 时间啊,你这把杀猪刀。。。

Time is a killer...

Machine: 我是被杀的

I am the one who was killed.

User: 怕孤獨,又捨不下自己的小自由。

Machine: 我想去流浪,但没有人陪我去。





Good Examples

User: 你若安好,便是晴天

Machine: 这句话我爱听,谢谢你的祝福,我会继续努力的!

User: 雪,飘舞。人,行走。树,纯美。天,苍茫。城,安静。

Snow dancing, people walking. Beautiful trees, boundless sky, quiet city.

Machine: 好美的意境

So beautiful scene.

User: 欲享受选择之快乐必先经历选择之痛苦

Machine: 是的,人生就是这样,人生的意义就是选择。

User: 有些命里遇见,从相识到相知,到最后的相忘于江湖,似乎都是命

中注定。

Machine: 缘来缘去,这就是人生





Bad Examples

User: 其实,他没有那么爱你。

Ohh, he does not love you that much.

Machine: 幸福的一天开始了。

A happy day starts.

User: 回北京了! 很冷很高兴!

Back to Beijing now. So cold so happy!

Machine: 什么时候来北京啊?

When will you come to Beijing?

User: 会做饭的男人是最性感的男人。

Mans who can cook are the most sexy.

Machine: 会做饭的男人是最性感的<u>女人</u>

Mans who can cook are the most sexy women.





Bad Examples

User: 猫叔羊毛毡手工作品,太美好了!

Machine: 这是什么品种的猫啊

User: 不得不夸一句广东的饭确实好吃!

Machine: 我也喜欢吃臭豆腐。

User: 我来自上海。

I come from Shanghai.

Machine: 真巧,我来自海南,<u>老乡啊</u>

So cool, I am from Hainan, the same province!





Bad Examples

语义理解的问题 Semantic Understanding

User: 你家多大? How large is your house?

小冰:这么快就要爆年龄了?

So soon to reveal my age?

User: 你是素食主义者还是吃肉啊?

Are you vegan or meatatarian?

小冰:怎么减肥

How can I lose weight?

上下文理解的问题 Context issue

User: 早饭不好吃

小冰: 那明早一起吃

User: 行,你推荐的应该不错

小冰: 是吗?

User: 哈哈,我相信你

小冰: 呃呃

微软小冰



个性身份一致性问题 Inconsistency in personality

User: 你几岁了?

小冰: 我啊。。。反正是九零后!

User: 那你是几几年出生的?

小冰:错,75年阳历一月出生。

User: 你是九零后,怎么会75年出生?

小冰: 生在九零后, 在深圳只能被当做八

零后了。





Bad Examples (AI Ethics)

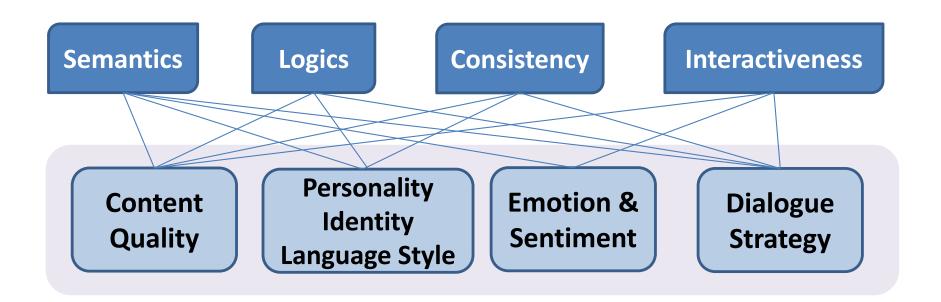








Challenges in Chatting Machines



Open-domain, open-topic conversational systems





- One-to-many: one input, many many possible responses
- Knowledge & Reasoning: real understanding requires various knowledge, world facts, or backgrounds
- Situational Context
 - Who are you talking with?
 - Stranger, or friend?
 - His mood and emotion?
 - Shared backgrounds that are only accessible by two acquaintances





Open-domain Chatting Machines

- Perceiving and Expressing Emotions (AAAI 2018)
- Behaving More Consistently with Personality Assignment (IJCAI-ECAI 2018)
- Better Understanding and Generation Using Commonsense Knowledge (IJCAI-ECAI 2018)
- Behaving More Proactively by Asking Good Questions (ACL 2018)

References:

- (1) Emotional Chatting Machine: Emotional Conversation Generation with Internal and External Memory. **AAAI 2018**.
- 2 Assigning personality/identity to a chatting machine for coherent conversation generation. **IJCAI-ECAI 2018**.
- (3) Commonsense Knowledge Aware Conversation Generation with Graph Attention. **IJCAI-ECAI 2018**.
- 4 Learning to Ask Questions in Open-domain Conversational Systems with Typed Decoders. **ACL 2018**.
- (5) Generating Informative Responses with Controlled Sentence Function. **ACL 2018**.



Emotion Perception and Expression





Emotional Chatting Machine

- Emotion intelligence is a key human behavior for intelligence (Salovey and Mayer, 1990; Picard and Picard, 1997)
- Understanding emotion and affect is important for dialogue and conversation
 - Enhance user performance
 - ◆ Improve user satisfaction
 - Less breakdowns
- Rule-based emotion adaptation is widely seen in early dialogue systems

User: Worst day ever. I arrived late because of the traffic.

Chatbot (Basic Seq2seq): You were late.

ECM (*Like*): I am always here to support you.

ECM (*Happiness*): Keep smiling! Things will get better.

ECM (Sadness): It's depressing.

ECM (*Disgust*): Sometimes life just sucks.

ECM (Anger): The traffic is too bad!





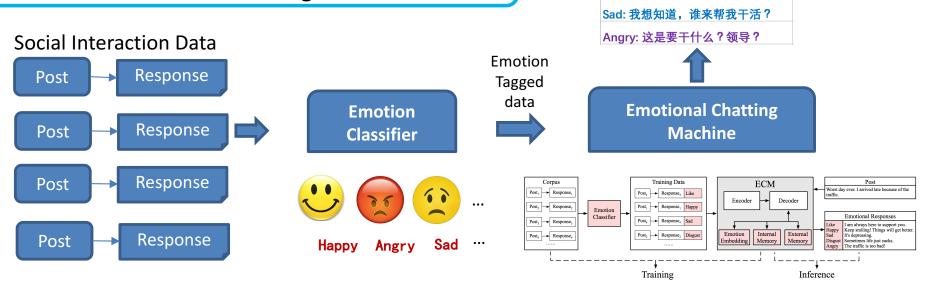
今天领导前来视察工作!

Happy: 哈哈, 领导视察工作辛苦了

Like: 这是一个很有特色的领导人

Emotional Chatting Machine

Perceiving and Expressing emotion by machine Closer to human-level intelligence



Our work was reported by MIT Technology Review, the Guardian, Cankao News, Xinhua News Agency etc.

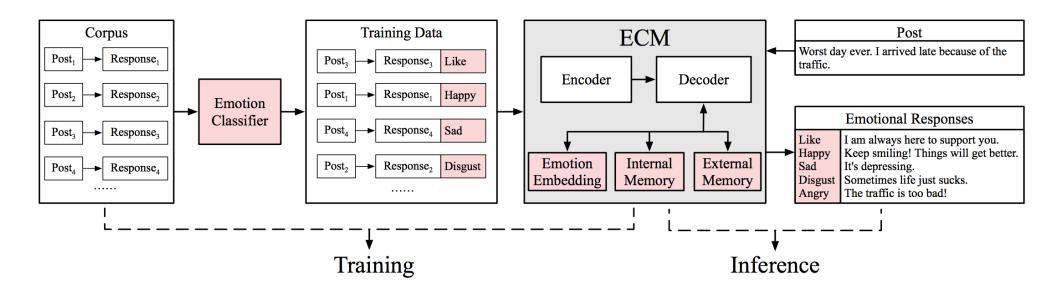
Prof Björn Schuller: "an important step" towards personal assistants that could read the emotional undercurrent of a conversation and respond with something akin to empathy.





Emotional Chatting Machine

- Emotion category embedding: High level abstraction of emotions
- Emotion internal state: Capturing the change of emotion state during decoding
- **Emotion external memory**: Treating emotion/generic words differentially







Emotional Chatting Machine

Post	M	odel	Response(Original)	Response(Translated)		
	Seq2Seq		承诺,就是一个骗子说给一个傻子听的。	A promise is what a cheater says to a fool.		
 承诺,就是一个骗子说给		Like	承诺是一个专一的傻子	Promise is a single-minded fool		
一个傻子听的。		Нарру	我就是一个 <mark>开心</mark> 的傻子	I'm just a happy fool		
A promise is what a cheater says to a fool.	ECM	Sad	我就是一个 <mark>悲伤</mark> 的傻子	I am a sad fool		
says to a root.		Disgust	一个 <mark>骗子</mark> 说给一个 <mark>傻子</mark> 听	A cheater says to a fool		
		Angry	不管你信不信,反正我信了!	Whatever you believe or not, I trust in it!		
	Seq2Seq		知道真相的我眼泪掉下来。	After knowing the truth, I burst into tears.		
	ECM	Like	太可爱了,太可爱了!	Too lovely, it's too lovely.		
知道真相的我眼泪笑下来 Knowing the truth, I laughed		Нарру	笑得我眼泪都出来了。	I can't help laugh myself to tears.		
myself to tears		Sad	知道真相的我 <mark>眼泪</mark> 掉下来。	After knowing the truth, I burst into tears.		
		Disgust	这是真的还是假的?	Is this true?		
		Angry	什么意思?没看懂。	What do you mean? I don't understand.		





More Examples

情绪 emotion	Post:早上的天气不错啊! 哈哈哈	哈哈哈
Other	今天是个好天气。	
喜欢 Like	今天的天气不错啊!	
悲伤 Sad	天气好,心情不好。	
厌恶 Disgust	这是哪里啊? 求解释	
愤怒 Angry	怎么了?什么情况?	
快乐 Happy	今天的天气真好啊!	

情绪	Post:为了解决抄作业的不好习惯,
emotion	高数老师这回要出绝招了!
Other	这老师是要逆天吗?
喜欢 Like	这老师太有才了!
悲伤 Sad	这样的老师伤不起啊
厌恶 Disgust	这老师也太狠了点吧
愤怒 Angry	这老师是要逆天啊!
快乐 Happy	哈哈,这老师太有才了

情绪	Post:送一个这样的蛋糕,	给你美丽的闺蜜
emotion	吧!	
Other	这是蛋糕蛋糕吗?	
喜欢 Like	好漂亮的蛋糕啊!	
悲伤 Sad	为什么我觉得很恶心	
厌恶 Disgust	这样的蛋糕你忍心吗?	
愤怒 Angry	有没有想开网店的?	
快乐 Happy	哈哈,好可爱的蛋糕!	

情绪 emotion	post:承诺,就是一个骗子说给一个傻子听的。
Other	承诺就是一个承诺
喜欢 Like	承诺是一个专一的傻子
悲伤 Sad	我就是一个悲伤的傻子
厌恶 Disgust	一个骗子说给一个傻子听
愤怒 Angry	不管你信不信,反正我信了!
快乐 Happy	我就是一个开心的傻子





Behaving More Consistently with Personality





Personality is important for game, custom service, etc.

General seq2seq model

User: Are you a boy or a girl?

Chatbot: I am a boy.

User: Are you a girl?

Chatbot: Yes, I am a girl.

User: 你几岁了?

小冰: 我啊。。。反正是九零后!

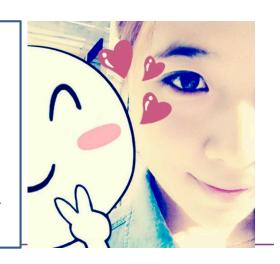
User: 那你是几几年出生的?

小冰:错,75年阳历一月出生。

User: 你是九零后,怎么会75年出生?

小冰: 生在九零后, 在深圳只能被当做八

零后了。



- Passing the **Turning Test**?
 - Deep semantic understanding
 - Existing chatting machine lacks identity or personality
- Personality is a well-defined concept in psychology(Norman, 1963; Gosling et al., 2003)
- Extremely **subtle**, **implicit** in language expression:
 - ◆ Age, gender, language, speaking style, level of knowledge, areas of expertise
- Existing works
 - ◆ Implicit personalization: learn implicit conversation style (Li et al., 2016; Al-Rfou et al., 2016)
 - ◆ Require dialogue data from different users with **user attributes tagged**

• Deliver coherent conversations w.r.t. identity/personality

Generic Dialogue Data for Training

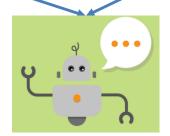
UserA: how old are you?

UserB: I am six.

UserA: do you like to play piano?

UserB: I play violin.

Personality-coherent Chatbot



Pre-specified Chatbot Profile

Profile key	Profile value
Name	汪仔(Wang Zai)
Age	三岁(3)
Gender	男孩(Boy)
Hobbies	动漫(Cartoon)
Speciality	钢琴(Piano)

Generated Dialogues

User: how old are you?

Machine: I am three years old.

User: do you like to play piano?

Machine: Yes, I play piano.

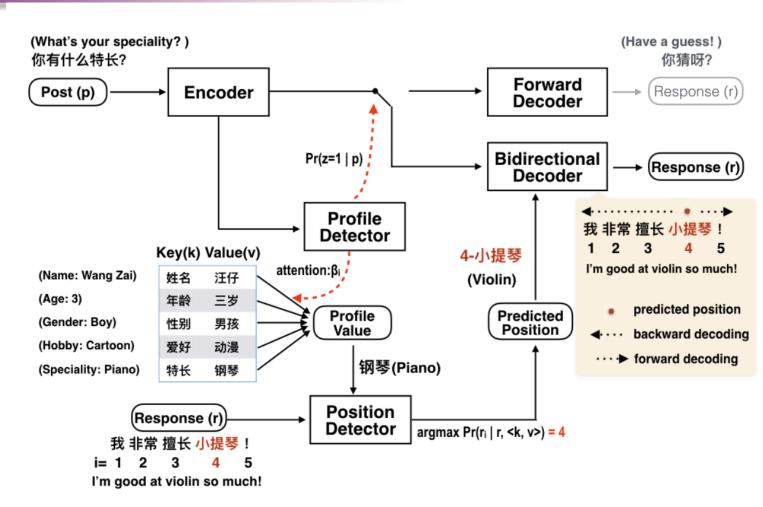


Encoder

Profile detector

Position detector

Decoder



•Qiao Qian, Minlie Huang, Haizhou Zhao, Jingfang Xu, Xiaoyan Zhu. Assigning personality/identity to a chatting machine for coherent conversation generation. **IJCAI-ECAI 2018**.



Post-level evaluation

Method	Nat.	Logic	Cor.
Seq2Seq	71.8%	56.0%	23.8%
Seq2Seq +PV	72.0%	56.0%	41.3%
Seq2Seq +PVD	73.3%	52.5%	38.0%
Our Model -PD	82.7%	51.7%	38.3%
Our Model	83.3%	59.5%	45.8%

Session-level evaluation

Method	Consistency	Variety		
Seq2Seq	1.3%	1.0%		
Seq2Seq +PV	47.0%	1.3%		
Seq2Seq +PVD	40.0%	7.5%		
Our Model -PD	38.8%	16.0%		
Our Model	49.5%	27.8%		

Generated sample responses that exhibit session-level consistency

Chinese	English(Translated)
U:你对什么事感兴趣	U:What are you interested in?
S:打篮球	S:Playing basketball.
U:你都有啥爱好呀	U:What's your hobby?
S:篮球	S:Basketball.
U:告诉我你兴趣所在	U:Tell me your interest.
S:我喜欢篮球	S:I like to play basketball.
U:你还没说你几岁呢	U:You haven't told me your age.
S:我三岁了	S:I'm three years old.
U:你今年有15了不	U:Are you 15 years old or not?
S:我还没到呢	S:I'm not yet.
U:你多大啦	U:How old are you?
S:3岁了	S:Three years old.

[•]Qiao Qian, Minlie Huang, Haizhou Zhao, Jingfang Xu, Xiaoyan Zhu. Assigning personality/identity to a chatting machine for coherent conversation generation. **IJCAI-ECAI 2018**.



Better Understanding and Generation with Commonsense Knowledge

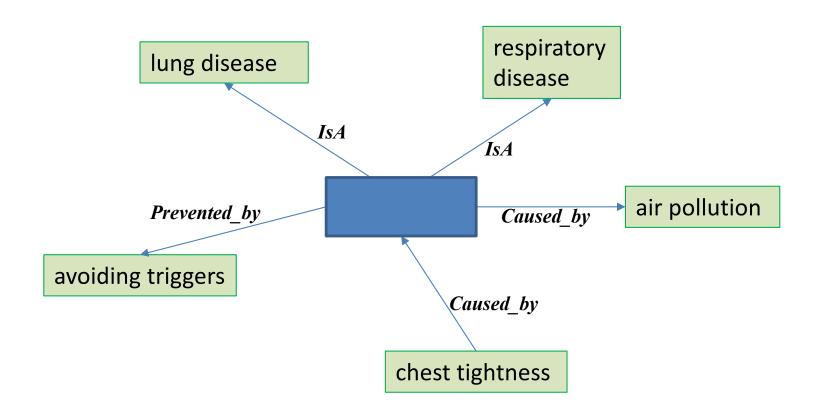




Commonsense Knowledge

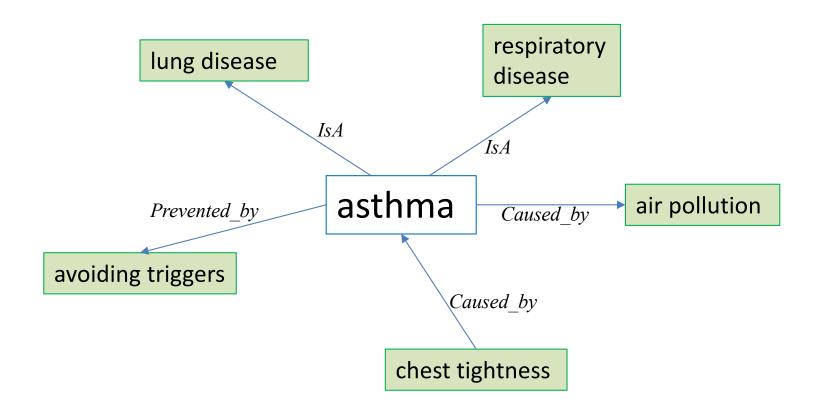
- Commonsense knowledge consists of facts about the everyday world, that all humans are expected to know. (Wikipedia)
 - Lemons are sour
 - Tree has leafs
 - Dog has four legs
- Commonsense Reasoning ~ Winograd Schema Challenge:
 - ◆ The trophy would not fit in the brown suitcase because it was too big. What was too big?
 - ◆ The trophy would not fit in the brown suitcase because it was too small. What was too small?







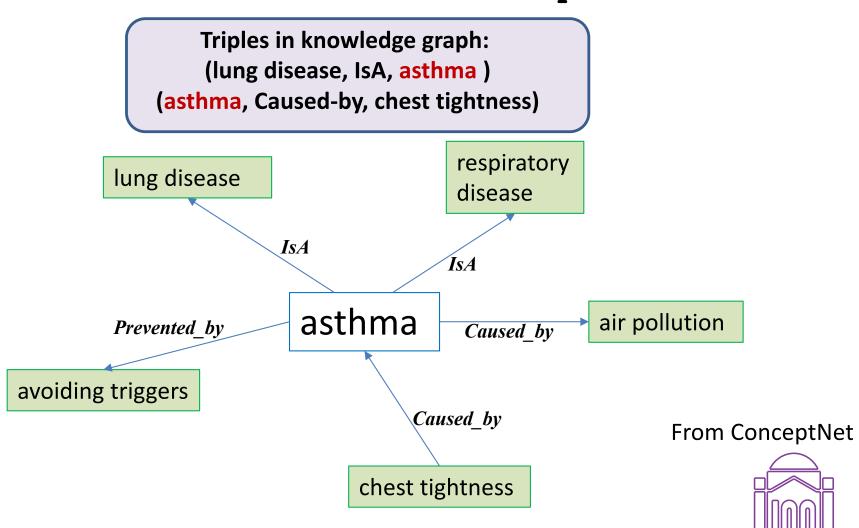








I have an asthma since three years old.





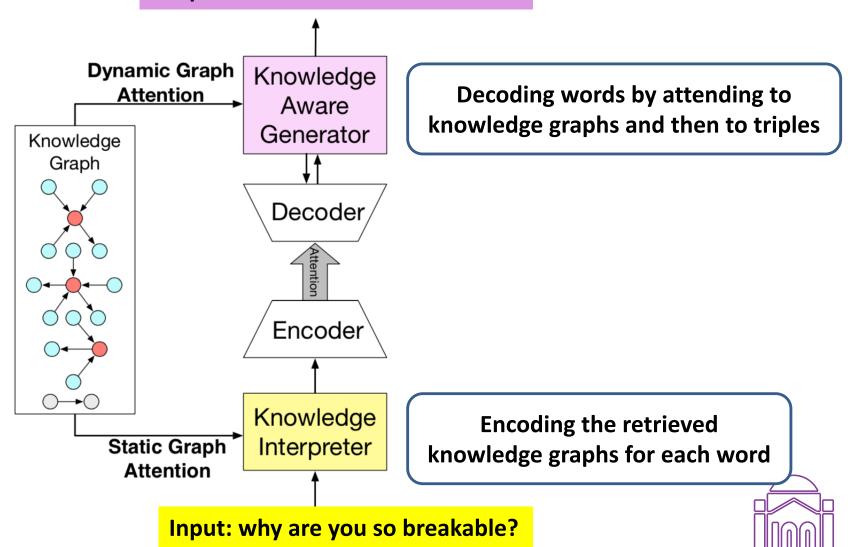
- Conversation is all about semantic understanding
- Commonsense knowledge is important for language understanding and generation

Post	Why are you so breakable ?
	(glass, RelatedTo, breakable),
Knowledge	(brittle, RelatedTo, breakable),
	(rule, RelatedTo, breakable)
Seq2Seq	I'm not a OOV , i 'm just a OOV .
MemNet	I'm not OOV . I'm just a really nice person.
CopyNet	I'm not. I'm just a lurker.
CCM	Because I'm a brittle man.





Output: Because I'm a brittle man.





Automatic evaluation

Model	Overall		High Freq.		Medium Freq.		Low Freq.		OOV	
	ppx.	ent.	ppx.	ent.	ppx.	ent.	ppx.	ent.	ppx.	ent.
Seq2Seq	47.02	0.717	42.41	0.713	47.25	0.740	48.61	0.721	49.96	0.669
MemNet	46.85	0.761	41.93	0.764	47.32	0.788	48.86	0.760	49.52	0.706
CopyNet	40.27	0.96	36.26	0.91	40.99	0.97	42.09	0.96	42.24	0.96
CCM	39.18	1.180	35.36	1.156	39.64	1.191	40.67	1.196	40.87	1.162

Manual evaluation (Sign-test, p-value<0.005)

Model	Overall		High Freq.		Medium Freq.		Low Freq.		OOV	
	app.	inf.	app.	inf.	app.	inf.	app.	inf.	app.	inf.
CCM vs. Seq2Seq	0.616	0.662	0.605	0.656	0.549	0.624	0.636	0.650	0.673	0.716
CCM vs. MemNet	0.602	0.647	0.593	0.656	0.566	0.640	0.622	0.635	0.626	0.657
CCM vs. CopyNet	0.600	0.640	0.606	0.669	0.586	0.619	0.610	0.633	0.596	0.640

Hao Zhou, Tom Yang, Minlie Huang, Haizhou Zhao, Jingfang Xu, Xiaoyan Zhu. Commonsense Knowledge Aware Conversation Generation with Graph Attention. **IJCAI-ECAI 2018**, Stockholm, Sweden.





Generation Examples 1

Post: He proposed March 5th. We will be married October 10th. So 7 months

Knowledge: (engaged, RelatedTo, married), (wedding, RelatedTo, married), (miss, DistinctFrom, married), (March, RelatedTo, months)

- Seq2Seq: I'm not sure how to answer that. I'm not sure how to make a date.
- MemNet: I'm assuming he's married to a gay marriage?
- CopyNet: I'm sorry, I'm not sure what you mean by that.
- CCM: I'm pretty sure he's going to be engaged in the next 6 months.



Behaving More Proactively by Asking Good Questions





Asking good questions requires scene understanding

Scene: Dining at a restaurant

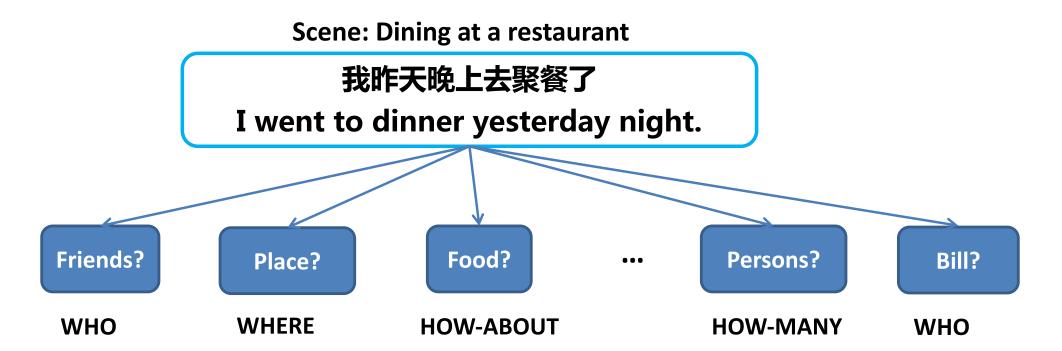
我昨天晚上去聚餐了

I went to dinner yesterday night.





Asking good questions requires scene understanding



Yansen Wang, Chenyi Liu, Minlie Huang, Liqiang Nie. Learning to ask questions in open-domain conversation systems. **ACL 2018**.





- Responding + asking (Li et al., 2016)
- Proactive behaviors (Yu et al., 2016)
- Asking good questions are indication of machine understanding
- Key differences to traditional question generation:
 - ◆ **Different goals**: Information seeking vs. Enhancing interactiveness and persistence of human-machine interactions
 - ◆ Various patterns: YES-NO, WH-, HOW-ABOUT, etc.
 - ◆ **Topic transition**: from topics in post to topics in response



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- A good question is a natural composition of
 - **◆ Interrogatives** for using various questioning patterns
 - ◆ Topic words for addressing interesting yet novel topics
 - ♦ Ordinary words for playing grammar or syntactic roles

Example 1:

User: I am too fat ...

Machine: How about climbing this weekend?

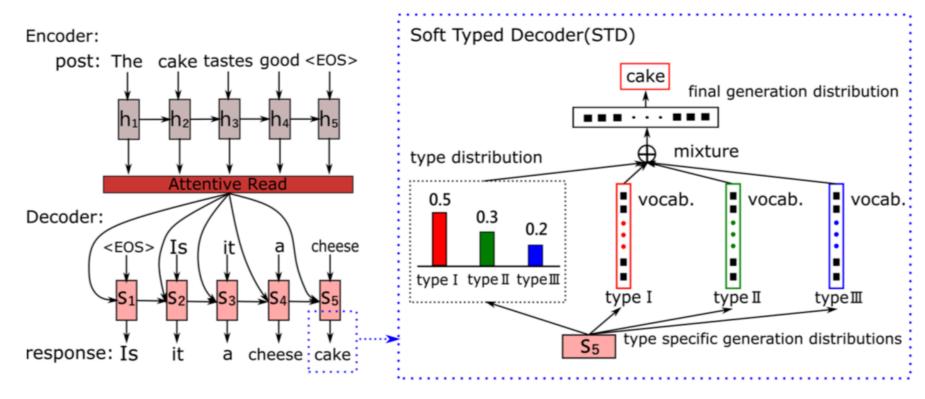
Example 2:

User: Last night, I stayed in <u>KTV</u> with friends. Machine: **Are** you happy with your <u>singing</u>?





• Typed decoders: soft typed decoder

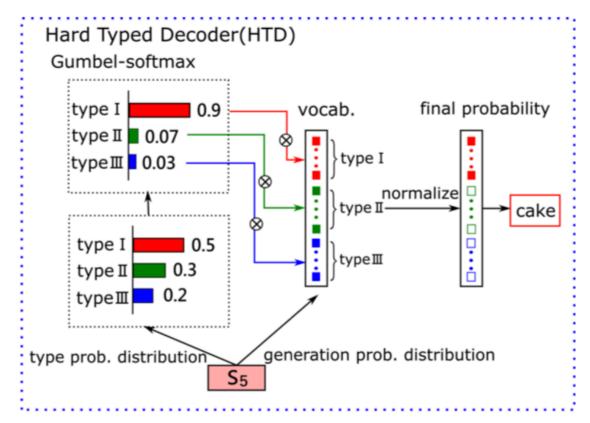


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• Typed decoders: hard typed decoder



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Type prediction at each decoding position

Post:	我喜欢小园	我喜欢小动物(I like little animals)									
Response:	你(you)	喜欢(like)	兔子(rabbit)	吗(<i>particle</i>)	?	_EOS					
Interrogative	0.09	0.02	0.01	0.85	1.00	0.01					
Topic word	0.26	0.35	0.71	0.14	0.00	0.02					
Ordinary word	0.65	0.63	0.28	0.01	0.00	0.97					

1 2 3





Metrics: Appropriateness, richness, willingness

Models	Appropriateness			Richness			Willingness		
	Win (%)	Lose (%)	Tie (%)	Win (%)	Lose (%)	Tie (%)	Win (%)	Lose (%)	Tie (%)
STD vs. Seq2Seq	42.0	38.6	19.4	37.2**	15.2	47.6	45.4*	38.6	16.0
STD vs. MA	39.6*	31.2	29.2	32.6**	16.8	50.6	49.4**	27.0	23.6
STD vs. TA	42.2	40.0	17.8	49.0**	5.4	45.6	47.6*	40.2	12.2
STD vs. ERM	43.4*	34.4	22.2	60.6**	13.2	26.2	43.2*	36.8	20.0
HTD vs. Seq2Seq	50.6**	30.6	18.8	46.0**	10.2	43.8	58.4**	33.2	8.4
HTD vs. MA	54.8**	24.4	20.8	45.0**	17.0	38.0	67.0**	18.0	15.0
HTD vs. TA	52.0**	38.2	9.8	55.0**	5.4	39.6	62.6**	31.0	6.4
HTD vs. ERM	64.8**	23.2	12.0	72.2**	8.4	19.4	56.6**	36.6	6.8
HTD vs. STD	52.0**	33.0	15.0	38.0**	26.2	35.8	61.8**	30.6	7.6

Yansen Wang, Chenyi Liu, Minlie Huang, Liqiang Nie. Learning to ask questions in open-domain conversation systems. **ACL 2018**.





Post: 我们昨天去唱歌了

We went to karaoke yesterday.

Seq2Seq: 什么时候?

When?

MA: 你怎么知道我是在问

How did you know I am questioning you?

TA: 什么?

What?

ERM: 什么情况?

What happened?

STD: 去哪里唱歌?

Where did you sing karaoke?

HTD: 你们几个人唱歌?

How many people were singing with you?

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Summary

- Emotion, personality, and knowledge is important to Chatbots
- Open-domain chatting machine is one of the most challenging AI tasks
 - ◆ Requires the ability of understanding **semantics**, **knowledge**, and **situational context**
 - Ability of making reasoning
 - Exhibits consistent personality
- Still a long way to go: existing generation models are still far from the expectation of real-world applications



Future Research Problems

Multi-modality emotion perception and expression (voice, vision, text)

- Personality, identity, style "human-like robot"
 - **♦** Introvert or extrovert
 - Personalized (style, or profile)

- Learning to learn
 - Grow up from interactions with human partners and environment





Thanks for Your Attention

- Minlie Huang, Tsinghua University
- aihuang@tsinghua.edu.cn
- http://coai.cs.tsinghua.edu.cn/hml
- Recruiting post-doctors and international PhDs!

