

Building Intelligent Chatting Machines: Emotion, Personality, and Commonsense Knowledge

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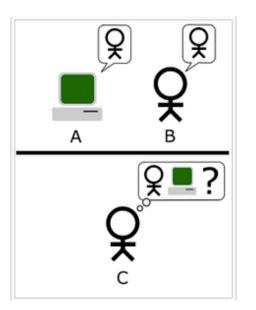
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Open-domain Chatting and Conservational AI



2010 IBM Watson: Humanlevel knowledge QA





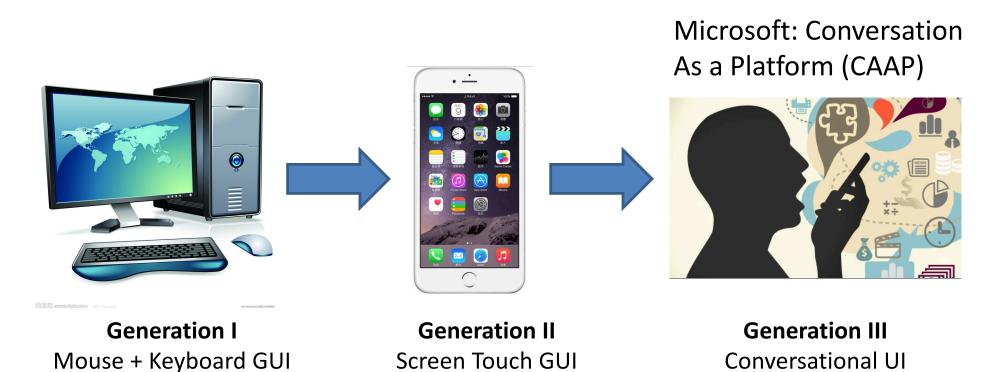
2014 Xiaoice: virtual partner

1950 Turing Test

Building open-domain chatting machines is one of the most challenging AI tasks.



Reshaping Human-Machine Interactions







Virtual Conversational Agents

Eliza Chat bot Is she a Rogerian 0 psychotherapist, a semi intelligent chat bot, or just a toy? Eliza, one of the original chat b is now ready to talk to you. She is programmed to behave a A.L.I.C.E.: | have never h Rogerian psychotherapist, and i Fake Kirk: So I am teach interesting example of the limitations of early artificial intelligence programs new. 2016 2010 1966 1994 2014 **QQ** chatbot **MS** Tay Eliza ALICE **Xiaoice**

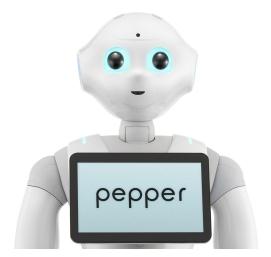


Social Robots (with Physical Body)

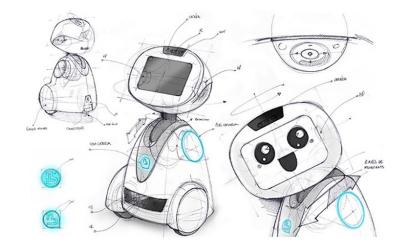


Pepper-humanoid robot The first robot with emotions

MIT Jibo: the world's first social robot for the home **Buddy:** protects your home, entertains the family, interfaces with devices, and assists the family









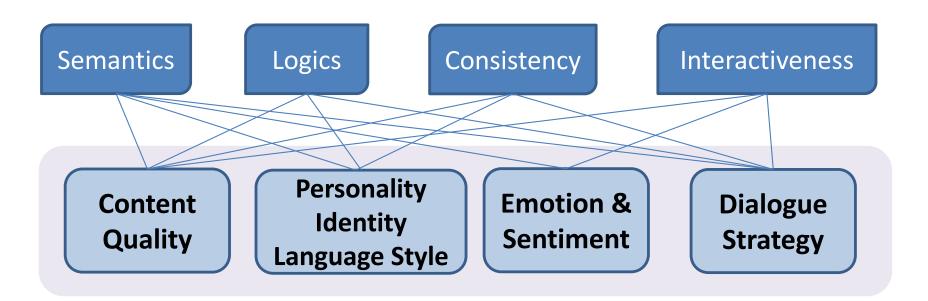


Business Values

- Personal assistants: SIRI, DUER, Cortana
- AI-aided customer services for industrial companies
- Robots with HCI (in many scenarios)
- Natural language interface to recommendation, search and many embedded systems
 - Apple homepod
 - Amazon Echo
 - ♦ Google Home



Challenges in Chatting Machines



Open-domain, open-topic conversational agents



大学

Challenges in Chatting Machines



- 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





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。。。 Machine: 我是被杀的

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。 Machine: 好美的意境

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。 Machine:幸福的一天开始了。

User: 回北京了! 很冷很高兴! Machine: 什么时候来北京啊?

User: 会做饭的男人是最性感的男人。 Machine: 会做饭的男人是最性感的<u>女人</u>





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。 Machine: 真巧,我来自海南,<u>老乡啊</u>





Bad Examples

语义理解的问题

User: 你家多大? 小冰: 这么快就要爆年龄了?

User: 你是素食主义者还是吃肉啊? 小冰: 怎么减肥

微软小冰

上下文理解的问题

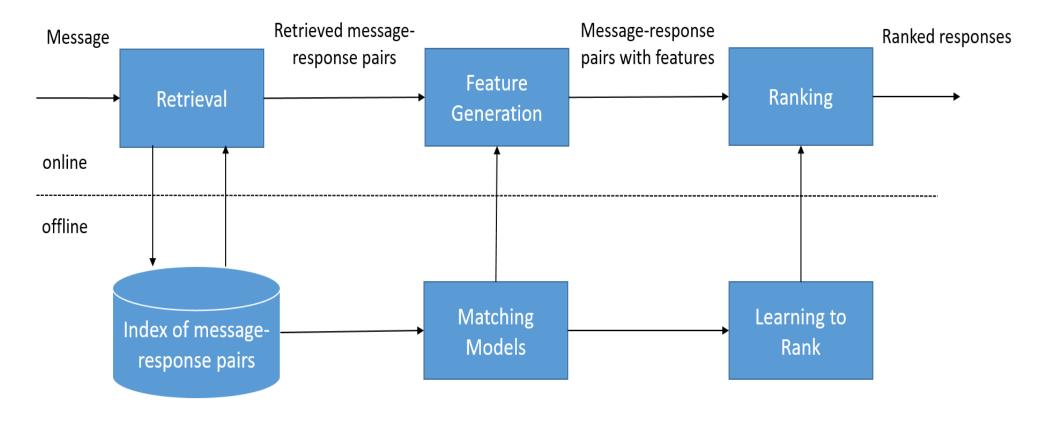
User: 早饭不好吃
 小冰: 那明早一起吃
 User: 行,你推荐的应该不错
 小冰: 是吗?
 User: 哈哈,我相信你
 小冰: 呃呃

个性身份一致性问题

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。



Typical Solution I: Retrieval-based



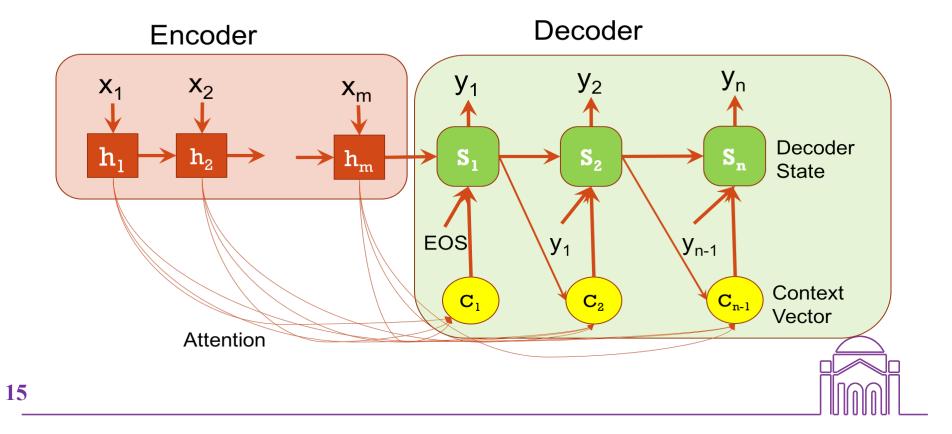
From MSRA Dr. Ming Zhou



Typical Solution II: Generation-based



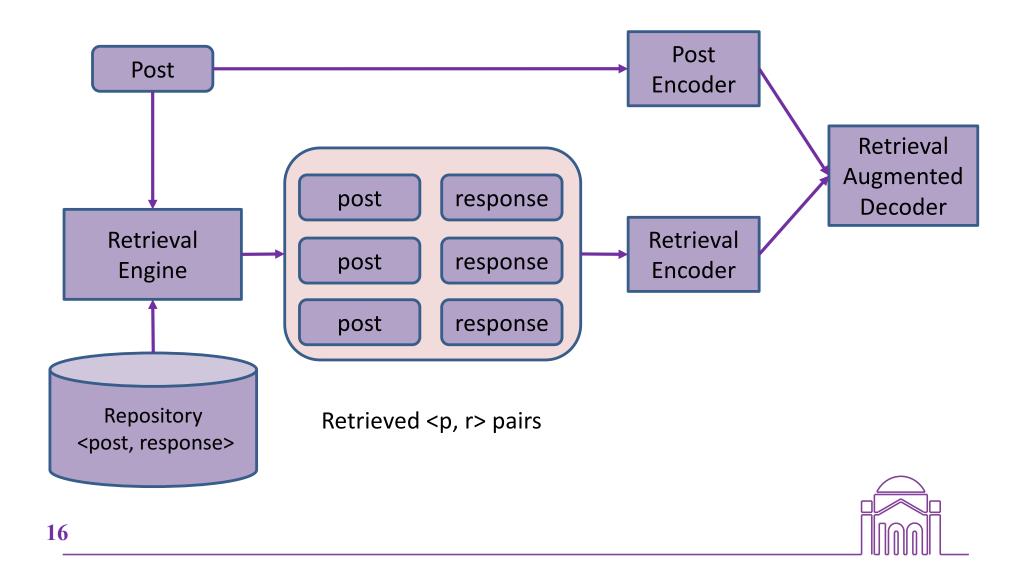
Machine: Glad with you.







Typical Solution III: Hybrid Methods





Open-domain Chatting Machines

- Emotional Chatting Machine (AAAI 2018)
- Personality Assignment (IJCAI-ECAI 2018)
- Commonsense Knowledge (IJCAI-ECAI 2018)
- 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.



Emotion





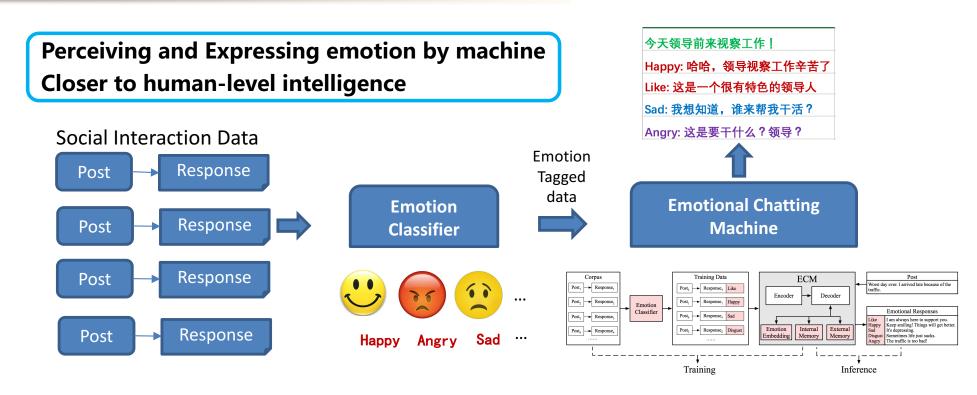
- Emotion intelligence is a key human behavior for intelligence (Salovey and Mayer, 1990; Picard and Picard, 1997)
- Understanding **emotion and affect** is important for dialogue and conversation
 - Enhance user performance
 - Improve user satisfaction
 - Less breakdowns
- **Rule-based** emotion adaptation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







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

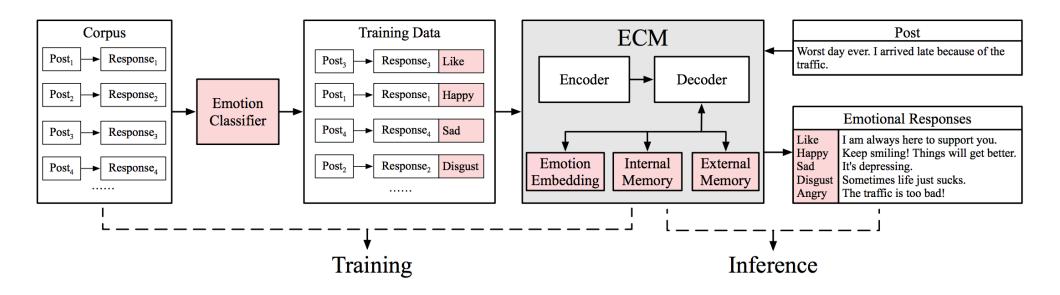
Prof Björn Schuller: "an important step" towards personal assistants that could read the emotional undercurrent of a conversation and respond with something akin to empathy.

•Hao Zhou, Minlie Huang, Xiaoyan Zhu, Bing Liu. Emotional Chatting Machine:
20 Emotional Conversation Generation with Internal and External Memory. AAAI 2018.





- **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



•Hao Zhou, Minlie Huang, Xiaoyan Zhu, Bing Liu. Emotional Chatting Machine: Emotional Conversation Generation with Internal and External Memory. **AAAI 2018.**





Post	Model		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 一个傻子听的。 A promise is what a cheater says to a fool.	ECM	Like	承诺是一个专一的傻子	Promise is a single-minded fool
		Нарру	我就是一个开心的傻子	I'm just a happy fool
		Sad	我就是一个 <mark>悲伤</mark> 的傻子	I am a <mark>sad</mark> fool
		Disgust	一个骗子说给一个傻子听	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.
知道真相的我眼泪笑下来 Knowing the truth, I laughed myself to tears	ECM	Like	太可爱了,太可爱了!	Too lovely, it's too lovely.
		Нарру	笑得我眼泪都出来了。	I can't help laugh 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.

•Hao Zhou, Minlie Huang, Xiaoyan Zhu, Bing Liu. Emotional Chatting Machine: Emotional Conversation Generation with Internal and External Memory. **AAAI 2018.**





More Examples

	post:早上的天气不错啊!哈哈哈	情绪	post:送一个这样的蛋糕,给你美丽的闺蜜吧!
情绪	哈哈哈	other	这是蛋糕蛋糕吗?
other	今天是个好天气。	喜欢	好漂亮的蛋糕啊!
喜欢	今天的天气不错啊!	悲伤	为什么我觉得很恶心
悲伤	天气好,心情不好。	厌恶	这样的蛋糕你忍心吗?
厌恶	这是哪里啊? 求解释	愤怒	有没有想开网店的?
愤怒	怎么了?什么情况?	快乐	哈哈,好可爱的蛋糕!
快乐	今天的天气真好啊!		
	post:为了解决抄作业的不好习惯,	情绪	post:承诺,就是一个骗子说给一个傻子听的
情绪	高数老师这回要出绝招了!	other	承诺就是一个承诺
other	这老师是要逆天吗?	喜欢	承诺是一个专一的傻子
喜欢	这老师太有才了!	悲伤	我就是一个悲伤的傻子
悲伤	这样的老师伤不起啊	厌恶	一个骗子说给一个傻子听
厌恶	这老师也太狠了点吧	愤怒	不管你信不信,反正我信了!
愤怒	这老师是要逆天啊!	快乐	我就是一个开心的傻子
快乐	哈哈,这老师太有才了		

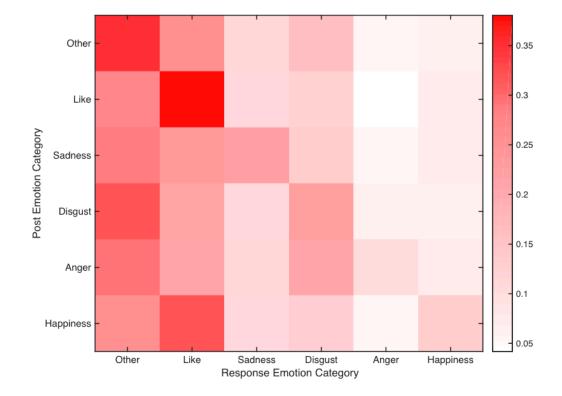
•Hao Zhou, Minlie Huang, Xiaoyan Zhu, Bing Liu. Emotional Chatting Machine: Emotional Conversation Generation with Internal and External Memory. **AAAI 2018.**



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Emotion Interaction Patterns



Like→Like (empathy)
 Sadness →Sadness (empathy)
 Sadness →Like (comfort)
 Disgust → Disgust (empathy)
 Disgust → Like (comfort)
 Anger → Disgust
 Happiness→Like

•Hao Zhou, Minlie Huang, Xiaoyan Zhu, Bing Liu. Emotional Chatting Machine: Emotional Conversation Generation with Internal and External Memory. **AAAI 2018.**



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Personality





• Personality is important for game, custom service, etc.

Conversation without Chatbot's personality by general sequence-to-sequence models. User: Are you a boy or a girl? Chatbot: I am a boy. User: Are you a girl? Chatbot: Yes, I am a girl. **Conversation with Chatbot's personality** by our model. User: Are you a boy or a girl? Chatbot: I am a three-year old boy. User: Are you a girl? Chatbot: I am a three-year old boy.





• 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)
- Clues that portray personality:
 - Age, gender, language, speaking style, level of knowledge, areas of expertise
- Existing works
 - User embedding: learn implicit conversation style (Li et al., 2016; Al-Rfou et al., 2016)
 - Require dialogue data from different users with user attributes tagged



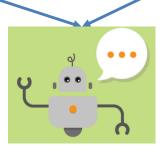
• Generating coherent conversation w.r.t. **identity/personality**

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.

> Identity-coherent Chatbot



Pre-specified Chatbot Profile

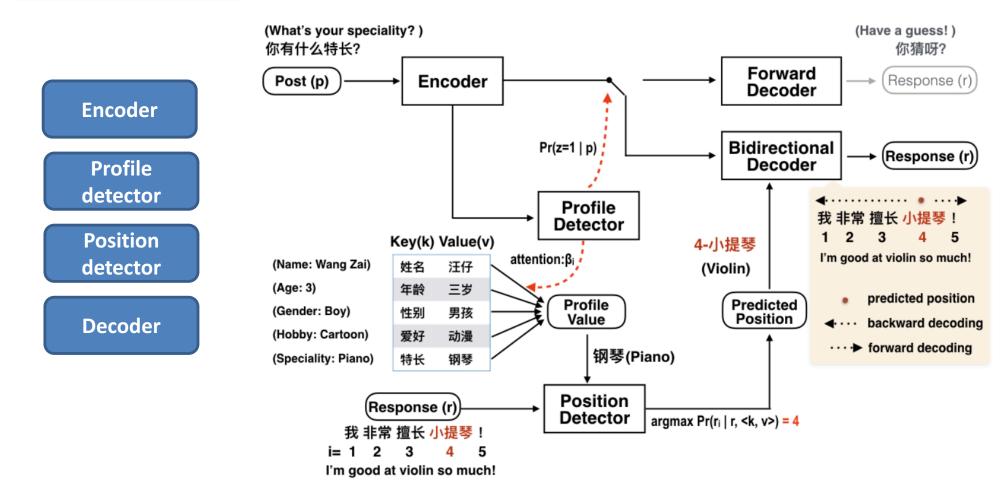
Profile key	Profile value
Name	汪仔(Wang Zai)
Age	三岁(3)
Gender	男孩(Boy)
Hobbies	动漫(Cartoon)
Speciality	钢琴(Piano)

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.





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



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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**.





Commonsense Knowledge





Commonsense Knowledge

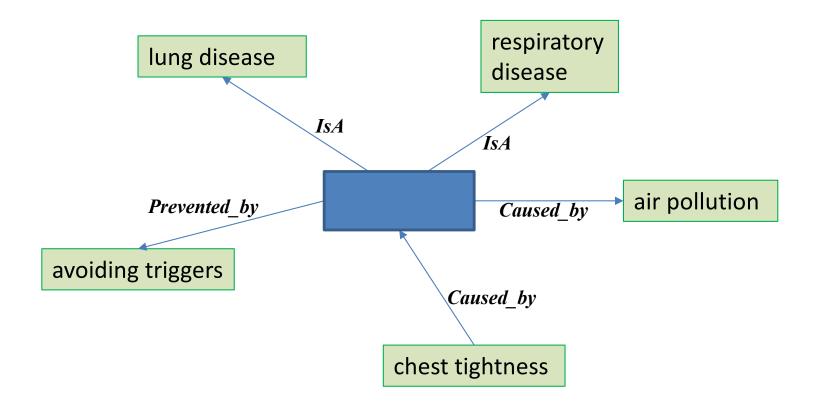
• Commonsense knowledge consists of facts about the everyday world,

that all humans are expected to know. (Wikipedia)

- Lemons are sour
- Tree has leafs
- Dog has four legs
- The main focus of the Paul <u>Allen Institute for Artificial Intelligence</u> (\$125 million in 2018)
- Commonsense Reasoning ~ Winograd Schema Challenge:
 - The trophy would not fit in the brown suitcase because it was too big (*small*). What was too big (*small*)?

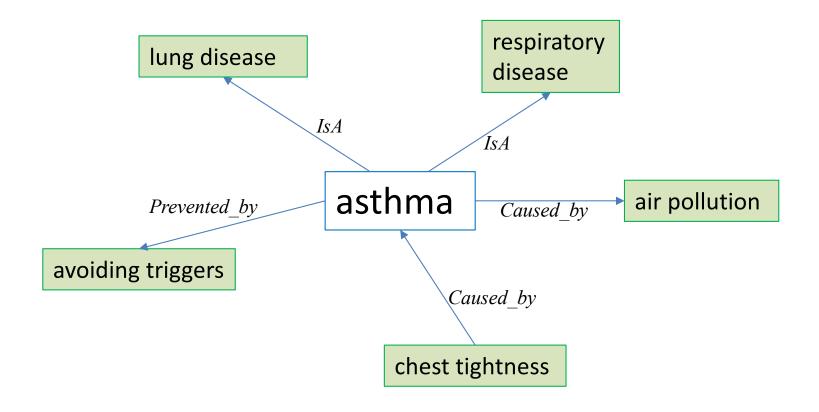








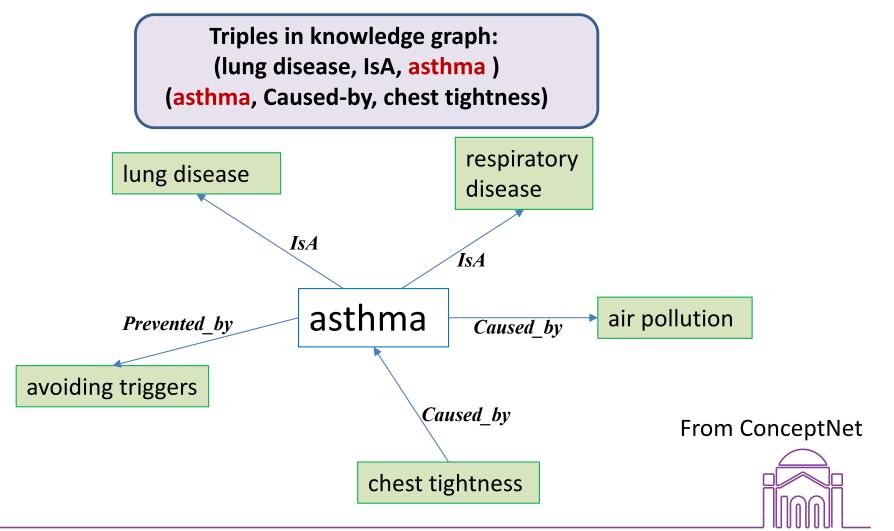








I have an asthma since three years old.





- 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.
ССМ	Because I'm a brittle man .

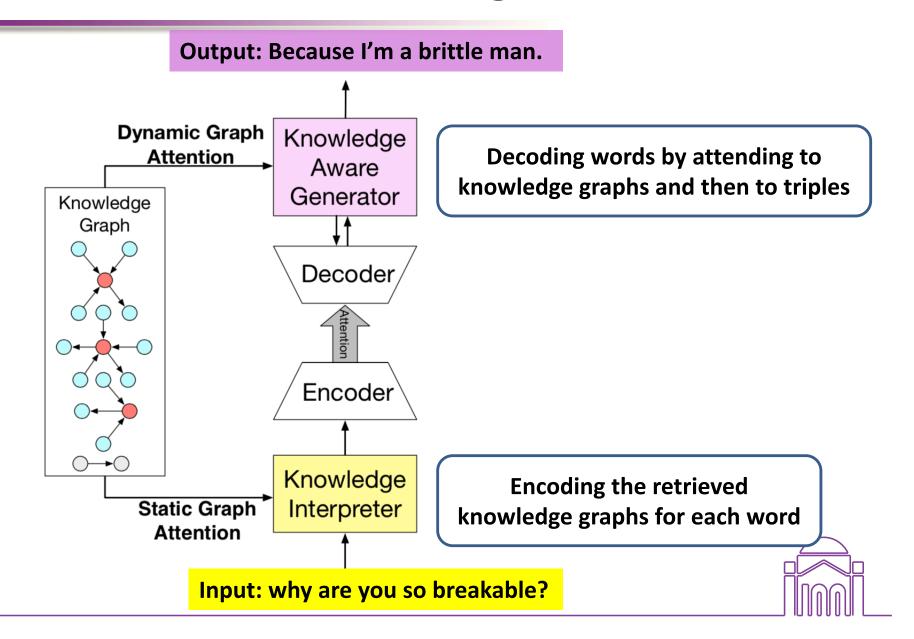




- Challenge I: applying large-scale, high-quality commonsense knowledge
- Challenge II: an elegant way of using commonsense knowledge in open-domain, open-topic conversation generation
- Our solution:
 - Static graph attention
 - Dynamic graph attention
 - Treating a knowledge graph as a whole



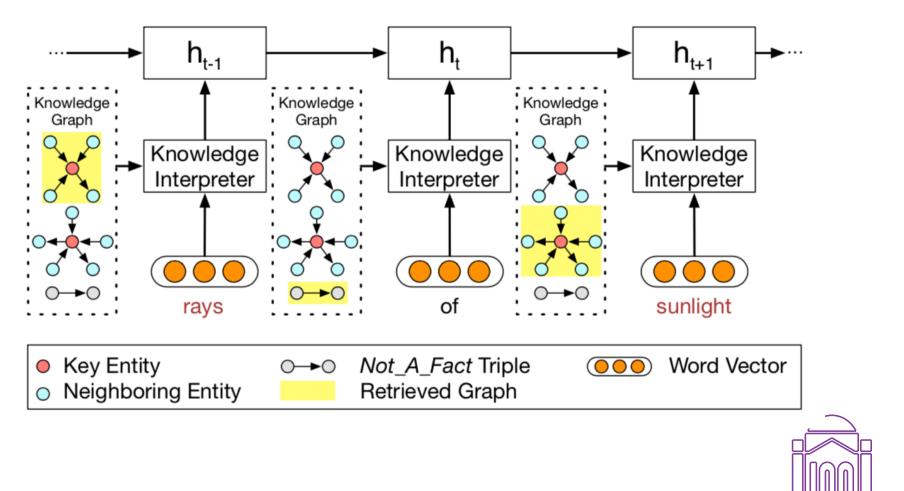




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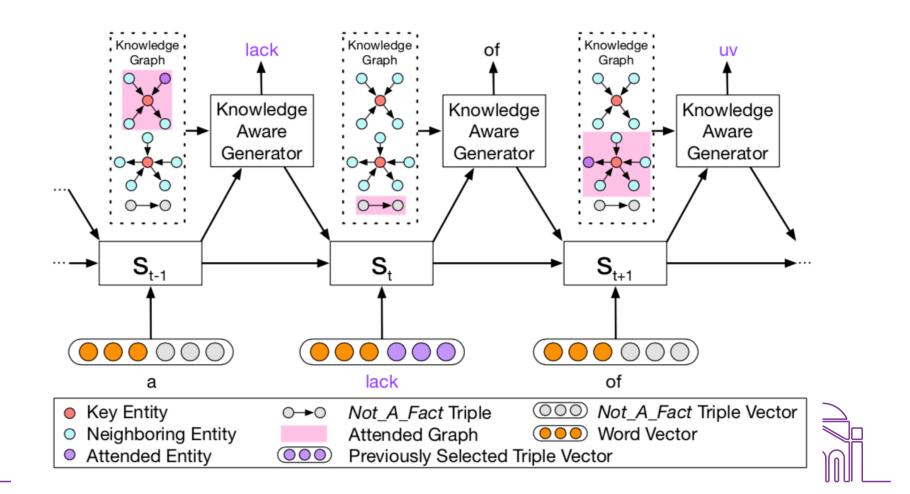


Static graph attention: encoding semantics in graph, Feeding knowledge-enhanced info. into the encoder





Dynamic graph attention: first attend a graph, then to a triple within that graph, finally generate with the words in a graph





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

Automatic evaluation

Manual evaluation

(Sign-test, p-value<0.005)

Model	Overall		High Freq.		Medium Freq.		Low Freq.		OOV	
WIOUCI	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.



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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**.



Generation Examples 2

<u>Post:</u> Totally <u>thought</u> it was going to be <u>doug's grave</u>. Knowledge: (thought, RelatedTo, thinking), (doug, HasProperty, dead), (dead, RelatedTo, grave), (grave, RelatedTo, death)

- Seq2Seq: I was thinking the same thing.
- MemNet: I thought it was a dead end.
- **CopyNet**: I think it's a bit of a stretch.
- **CCM**: I thought it was doug's **death**. I was **thinking** the same thing.



Asking Good Questions





- 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

Yansen Wang, Chenyi Liu, Minlie Huang, Liqiang Nie.





• 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 <u>fat</u> ... Machine: **How about** <u>climbing</u> this weekend?

Example 2: User: Last night, I stayed in <u>KTV</u> with friends. Machine: **Are** you happy with your <u>singing</u>?





• Asking **good** questions requires **scene understanding**

Scene: Dining at a restaurant

我昨天晚上去聚餐了

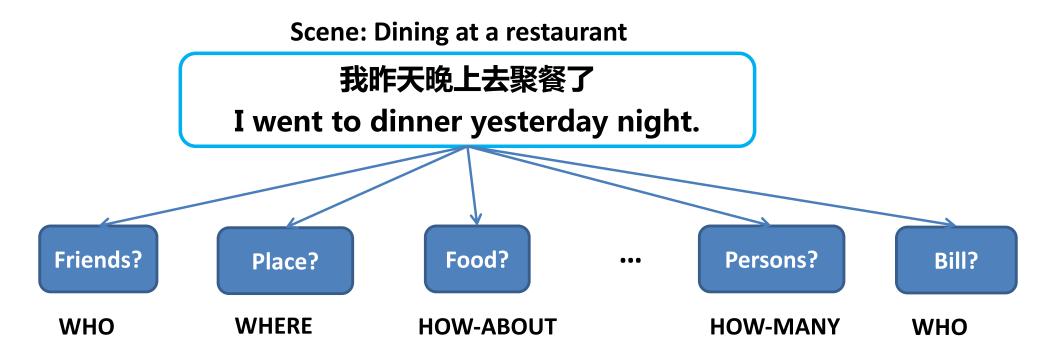
I went to dinner yesterday night.

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





• Asking **good** questions requires **scene understanding**

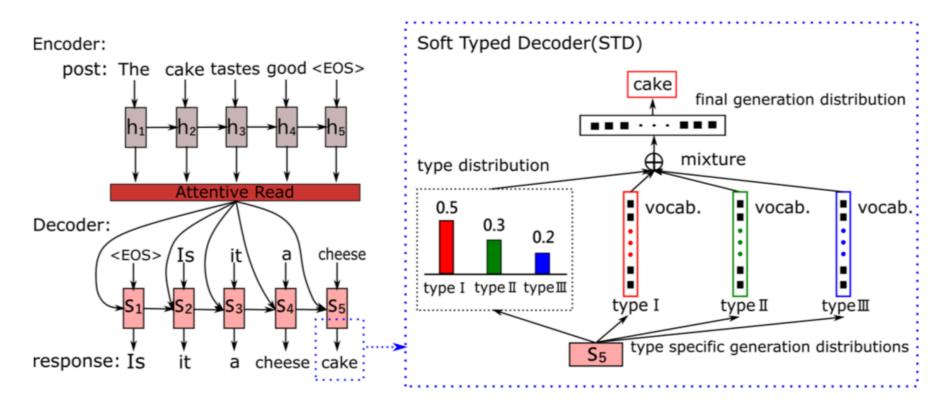


Yansen Wang, Chenyi Liu, Minlie Huang, Liqiang Nie.





• Typed decoders: soft typed decoder



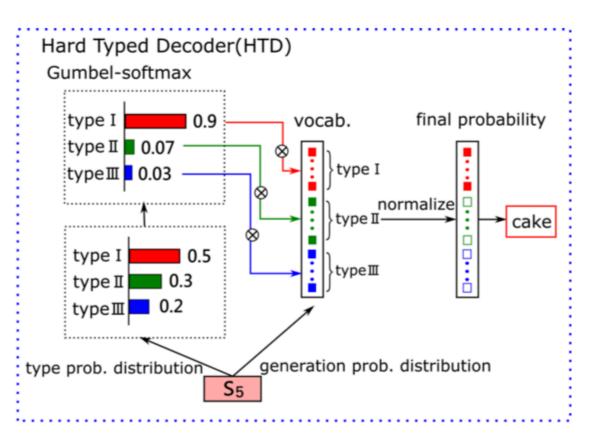
Yansen Wang, Chenyi Liu, Minlie Huang, Liqiang Nie. Learning to ask questions in open-domain conversation systems. **ACL 2018**.



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• Typed decoders: hard typed decoder



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





• Metrics: Appropriateness, richness, willingness

Models	Appropriateness			Richness			Willingness		
Widdels	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

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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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Post-3:	我在体育中心上班。
Human:	I <u>work</u> at the <u>sports center</u> . 你每天从 <u>家过去</u> 要 多久?
STD:	How long does it <u>take</u> you to <u>go</u> there from <u>home</u> ? 你是什么 <u>部门</u> 的啊?
HTD:	Which <u>department</u> are you from? 那你是 <u>体育学院的学生</u> 吗?
	Are you a student from the College of Physical Education?
Post-4:	今天的寿司真好吃。
Human:	The <u>sushi</u> today is so delicious! 我记得你不喜欢 <u>吃日式料理</u> 的不是 吗?
STD:	I remember you didn't like <u>eating Japanese food</u> , did you? 你 什么 时候来 <u>无锡</u> ,我请你。
HTD:	When will you come to <u>Wuxi</u> ? It is my <u>treat</u> . 是吗?改天 <u>尝尝</u> ! Really? I will try sometime!
	really. I will <u>up</u> sometime.

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• Type prediction at each decoding position

Post:	我喜欢小动物(I like little animals)							
Response:	你(you)	喜欢(like)	?	_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		





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 \rightarrow "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
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- <u>http://coai.cs.tsinghua.edu.cn/hml</u>
- Recruiting post-doctors and international PhDs!

