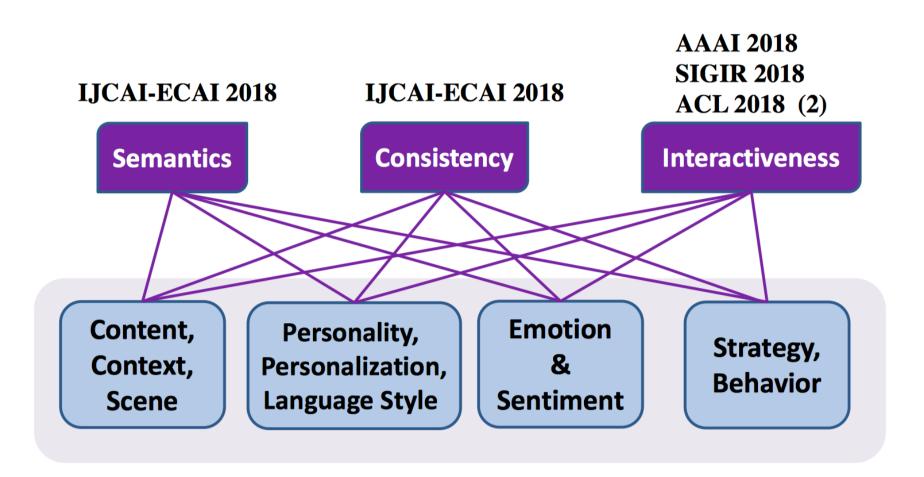
# Emotional Chatting Machine: Emotional Conversation Generation with Internal and External Memory

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## Challenges in Chatting Machines



## Challenges in Chatting Machines

- Behaving more <u>interactively</u>:
  - Perceiving and Expressing Emotions (AAAI 2018)
  - ◆ Proactive Behavior by Asking Good Questions (ACL 2018)
  - Controlling sentence function (ACL 2018)
  - ◆ Topic change (SIGIR 2018)
- Behaving more consistently:
  - **♦ Explicit Personality Assignment (IJCAI-ECAI 2018)**
- Behaving more <u>intelligently</u> with <u>semantics</u>:
  - ◆ Better Understanding and Generation Using Commonsense Knowledge (IJCAI-ECAI 2018 distinguished paper)

#### **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**.
- 6 Chat more: deepening and widening the chatting topic via a deep model. SIGIR 2018.

# **Emotional Chatting Machine**

#### **Example:**

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

Sad: 我想知道,谁来帮我干活?

Angry: 这是要干什么?领导?

### Challenge

- Emotion-labeled data are difficult to obtain in a large-scale corpus.
- Balance grammaticality and expressions of emotions.
- Existing methods cannot produce desirable emotional responses but just hard-to-perceive general expressions.

## **Importance**

- 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!

## Task Definition

**Encoder-Decoder Framework:** 

$$y_t \sim \mathbf{o}_t = P(y_t \mid y_1, y_2, \cdots, y_{t-1}, \mathbf{c}_t),$$
  
= softmax( $\mathbf{W}_{\mathbf{o}} \mathbf{s}_t$ ).

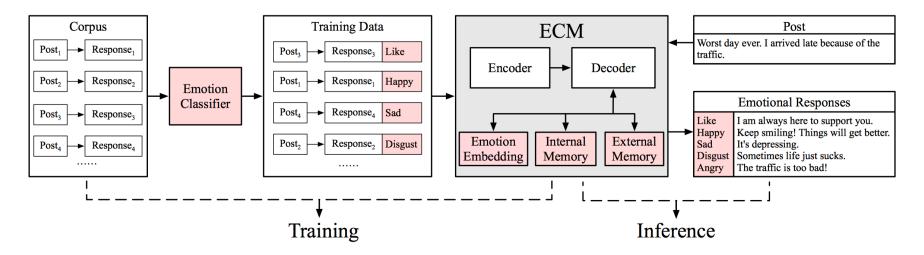
Emotional Chatting Machine(ECM):

$$P(\mathbf{Y}|\mathbf{X},e) = \prod_{t=1}^{m} P(y_t|y_{< t},\mathbf{X},e)$$

emotion category

{Angry, Disgust, Happy, Like, Sad, Other}

## Overview



- First, since the emotion category is a high-level abstraction of an emotion expression, ECM embeds the emotion category and feeds the emotion category embedding to the decoder.
- Second, we assume that during decoding, there is an internal emotion state, and
  in order to capture the implicit change of the state and to balance the weights
  between the grammar state and the emotion state dynamically, ECM adopts an
  internal memory module.
- 3. Third, an explicit expression of an emotion is modeled through an explicit selection of a generic (non-emotion) or emotion word by an external memory module.

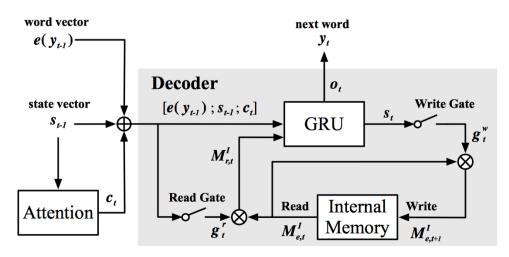
# **Emotion Category Embedding**

- Each emotion category is represented by a real-valued, low dimensional vector
- 2. For each emotion category e, we **randomly initialize** the vector of an emotion category ve
- 3. learn the vectors of the emotion category through training
- 4. Note that: static the emotion category embedding will not change during the generation process which may sacrifice grammatical correctness of sentences

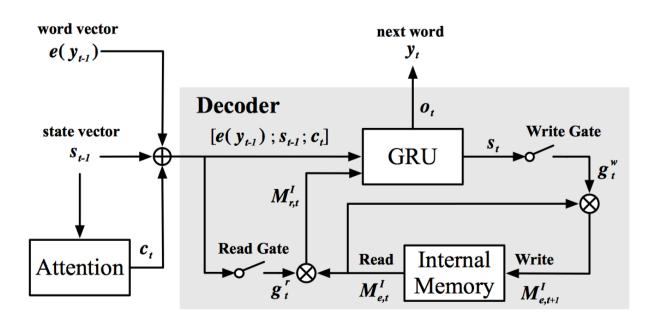
$$s_t = \mathbf{GRU}(s_{t-1}, [c_t; e(y_{t-1}); \underline{v_e}]).$$

# Internal Memory

- Psychological (emotional responses)
  - Relatively short lived
  - Involve changes
  - Dynamic emotion situation
- Goal: capture the emotion dynamics during decoding
- Process:
  - Init internal emotion state for each category
  - At each step the emotion state decays by a certain amount
  - The emotion state should decay to zero indicating the emotion is completely expressed when decoding finished



# Internal Memory



$$egin{array}{lll} oldsymbol{g}^{r} &=& \operatorname{sigmoid}(\mathbf{W}_{\mathbf{g}}^{\mathbf{r}}[oldsymbol{e}(y_{t-1});oldsymbol{s}_{t-1};oldsymbol{c}_{t}]), \, \operatorname{Read} \, \operatorname{gate} \ oldsymbol{g}^{w}_{t} &=& \operatorname{sigmoid}(\mathbf{W}_{\mathbf{g}}^{\mathbf{w}}oldsymbol{s}_{t}). \, \, \, \operatorname{Write} \, \operatorname{gate} \ \end{array}$$

read from and write into the internal memory, respectively

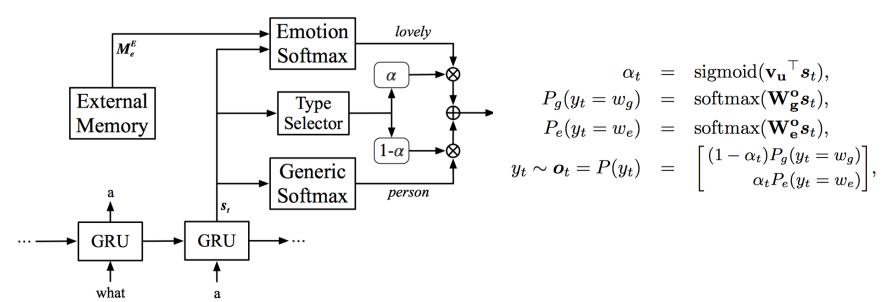
$$m{M}_{r,t}^I = m{g}_t^r \otimes m{M}_{e,t}^I, \;\; ext{Init at first step} \ m{M}_{e,t+1}^I = m{g}_t^w \otimes m{M}_{e,t}^I,$$

Information of internal state

$$m{s}_t = \mathbf{GRU}(m{s}_{t-1}, [m{c}_t; m{e}(y_{t-1}); m{M}_{r,t}^I]).$$

# **External Memory**

- Motivation
  - Emotion expressions are quite distinct with emotion words
- Goal
  - Model emotion expressions explicitly by assigning different generation probabilities to emotion words and generic words
- Choose to generate words from an emotion vocabulary a generic vocabulary.
- final distribution P(yt) is a concatenation of two distributions.



## Conclusion

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

#### 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 Sad: 我想知道, 谁来帮我干活? **Social Interaction Data** Angry: 这是要干什么?领导? **Emotion** Response **Post Tagged** data **Emotional Chatting Emotion** Response **Post Machine** Classifier Response **Post** Training Data **ECM** Post<sub>i</sub> --- Response Post, - Response, Like Encoder Post<sub>4</sub> → Response, Post₁ → Response₂ Disgust Response **Post** Training Inference

今天领导前来视察工作!

# Thanks!