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# Reinforcement Learning for Chatbot with Emotions

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## 1 Problem statement

Dialogues generation with similar emotions:

Given a user's sentence/utterance as an initial input, generate dialogues with similar or proper emotions and correct word context.

## 2 Motivation

In daily conversations, like online chatting, users send messages with emotional interactions instead of simple Q&A pattern. A natural, fluent one-to-one conversation, generally, contains dialogues in which both users share similar emotion levels, such as happy, sadness. Our goal is to model a natural dialogue and the model can figure out users' emotions and response words with proper emotions.

## 3 Proposed approach

Our project contains three parts:

### 3.1 Emotion extraction and feature embedding

In order to understand how emotions are presented in sentence and how can they be embedded into a metric space, we plan to train a classification model (sentence based RNN model) to do feature extraction and embedding. These features will be used later by our model to interpret user emotions. Also, this classification model will be used for adding possible emotino tags to our movie-subtile database.

### 3.2 Generative model for sentence generation

To generate a response from an input sentence, a generate model is necessary to model the sentence generation. Bascailly, we plan to use Seq2Seq model with word embedding method like word2vec to achieve word embedding and language modeling.

### 3.3 Reinforcement Learning Framework of dialogue generations

To take context information into consideration, we plan to fit our model into reinforcement learning framework. Specifically, we try to make our model predict a user's current emotion with optimized forward rewards (similarity of possible emotions) and backward rewards.

## 4 Datasets

We would collect 2.7 million sentences of subtitles from open subtitle database (<http://opus.lingfil.uu.se/OpenSubtitles.php/>). The subtitles of movies and TV programs provide sufficient conversation contents for training. Nevertheless, subtitles are slightly different from daily conversation due to theatrical and dramatic contents. We would filter long segment of words which is impractical in daily conversation, which could be monologues in movies.

To tag the sentences in OpenSubtitles dataset with emotion tags, we need to train a model to predict each sentence's emotion. We would use a dataset of tweets with emotion tags from LiveJournal to train the prediction model.

## References

- [1] Li, J., Monroe, W., Ritter, A., Galley, M., Gao, J., & Jurafsky, D. (2016). Deep reinforcement learning for dialogue generation. *arXiv preprint arXiv:1606.01541*.
- [2] Zhou, H., Huang, M., Zhang, T., Zhu, X., & Liu, B. (2017). Emotional Chatting Machine: Emotional Conversation Generation with Internal and External Memory. *arXiv preprint arXiv:1704.01074*.
- [3] Sutskever, I., Vinyals, O., & Le, Q. V. (2014). Sequence to sequence learning with neural networks. *In Advances in neural information processing systems* (pp. 3104-3112).