

# The Conversational Intelligence Challenge 2

Solution of “Lost in Conversation” team

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# Challenge

## Goal:

A non-goal-oriented dialog system with a persistent persona

## Problems:

- Lack of a consistent personality
- Lack of an explicit long-term memory
- Tendency to produce non-specific answers like “I don’t know”

## Ideal solution:

- Simulate a normal conversation
- Learn about the interests of opponent
- Discuss own interests and find common ground

# Datasets

## PersonaChat (original + revised):

*Zhang S. et al. Personalizing Dialogue Agents: I have a dog, do you have pets too?*

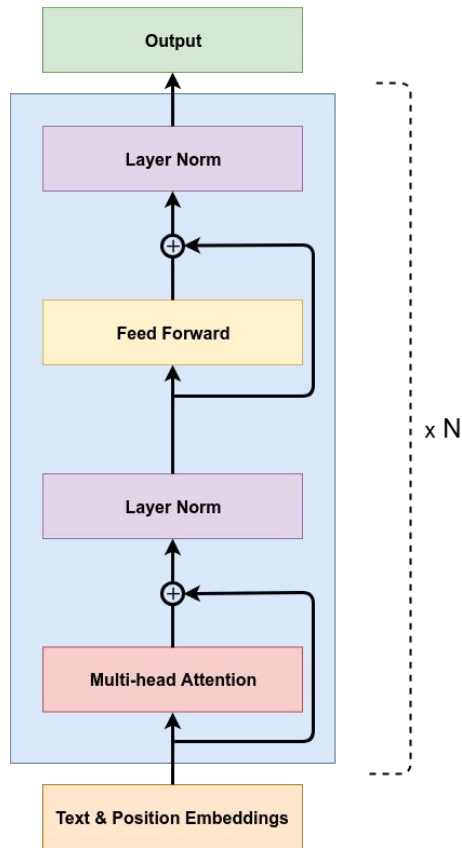
## DailyDialog:

*Li Y. et al. DailyDialog: A Manually Labelled Multi-turn Dialogue Dataset*

## Reddit comments dataset:

*[files.pushshift.io/reddit/comments](https://files.pushshift.io/reddit/comments)*

# Base architecture

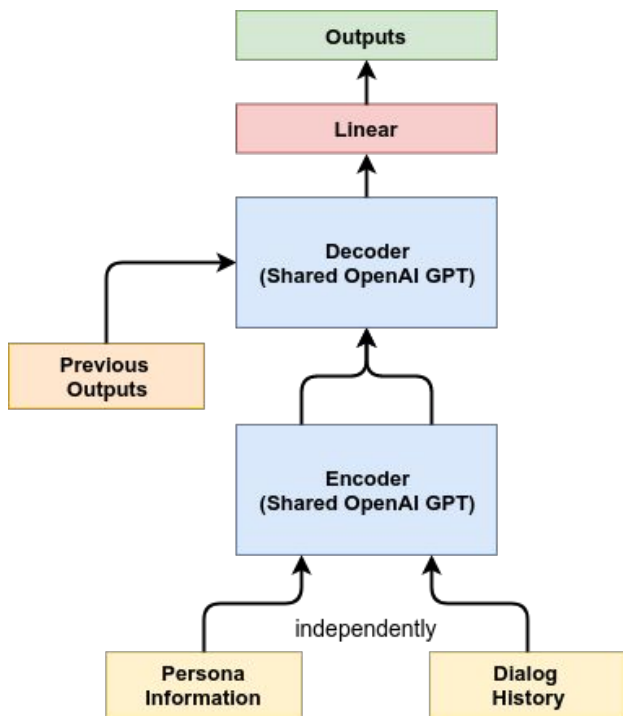


## OpenAI GPT<sup>1</sup> :

- BPE vocabulary with 40000 tokens
- Learned position embeddings with 512 positions
- 12 transformer layers
- Multi-head attention with 768 dimensional states and 12 attention heads
- Position-wise feed-forward networks with 3072 dimensional inner states

1. Radford, A., Narasimhan, K., Salimans, T., Sutskever, I. (2018). *Improving language understanding by generative pre-training.*

# Our architecture



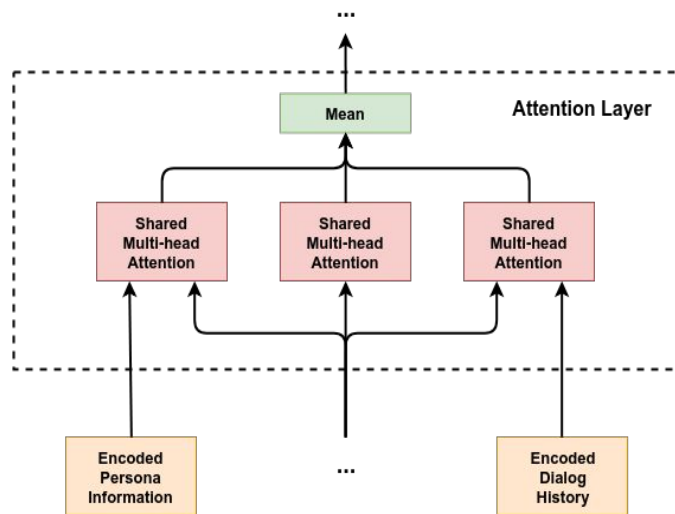
- Shared encoder and decoder - pretrained OpenAI GPT<sup>1</sup>
- Shared pre-softmax linear layer and token embeddings<sup>2</sup>
- Beam-search with length penalty<sup>3</sup> and annealing for improving answer diversity
- Reduction of persona information and dialog history – first and last 512 tokens respectively

1. Xia Y. et al. *Model-level dual learning*

2. Press O., Wolf L. *Using the output embedding to improve language models*

3. Wu Y. et al. *Google's neural machine translation system: Bridging the gap between human and machine translation*

# Our architecture: pretrained model



## Attention layer modifications:

- Shared multi-head attention layers
- Parallel computation of attention for inputs
- Merge of attentions - mean

# Learning procedure: loss functions

To train model we used weighted combination of losses<sup>1</sup>:

$$Loss = L_{TokLS} + \lambda_{LM} \cdot L_{LM} + \lambda_{risk} \cdot L_{risk}$$

$$L_{TokLS} = - \sum_i \log P(y_i | y_1, \dots, y_{i-1}) - D_{KL}(f || P(y_i | y_1, \dots, y_{i-1}))$$

$$L_{LM} = - \sum_i \log P(y_i | y_1, \dots, y_{i-1})$$

$$L_{risk} = \sum_{y_{pred} \in B} (1 - f1(y_{target}, y_{pred})) \cdot \frac{p(y_{pred})}{\sum_{y'_{pred} \in B} p(y'_{pred})}$$

First stage:

- $\lambda_{LM} = 0.5$

- $\lambda_{risk} = 0$

- $\lambda_{LM} = 0.1$  :

- $\lambda_{risk} = 10$

Beam-search samples

for risk minimization

# Learning procedure: settings

## Settings:

- batch size: 256
- learning rate:  $6.25e-5$
- warmup: 16000
- label smoothing: 0.1
- dropout: 0.1

## Training time:

- First stage: ~ week on Nvidia GTX 1080TI
- Second stage: ~ two days on Nvidia GTX 1080TI



# Heuristics

## Human behavior simulation:

- Sentiment analysis of human replies and adding emoji
- Mistake correction (spell, cycle conversation, ...)
- Questions from retrieval model to find out additional information about an interest



**Turned off: decrease automated evaluation metrics**

# Results: score

	<b>PPL</b>	<b>Hits@1</b>	<b>F1</b>
<b>Public set</b>	-	0.173	0.179
<b>Private set</b>	-	0.171	0.177

# Results: dialog examples

Persona:

I ride the bus to school.  
I hate math class.  
I am on the soccer team.  
My brother is older than me.

H (Human): Hello friend.

B (Bot): Hi how are you today?

H: I am fine, you?

B: I am doing well, just got off work. 🤖

H: Where do you work?

B: I'm in school right now. How about you?

H: I am a software developer.

B: That is cool. What do you do for fun? 🤖

H: I like to read books.

B: What kind of books do you like?

H: I prefer horrors.

B: Horror movies are the best!

H: I like them too.

B: What is your favorite book?

H: It by Stephen King.

B: I like Stephen King too. Do you work?

H: Right now no.

B: I like to eat pizza with my brother. 🤖

H: Bye.

B: Bye.

Persona:

I am a ninja.

H (Human): I like cats.

B (Bot): Cats are my favorite. I have a lot of them. What is your cats name?

H: I call my cat jerk, I hate him.

B: That is funny. Do you live in Japan?

H: No, you?

B: I live in the united states. I moved to America when I was 14. How about you Mars or Jupiter?

H: Mars.

B: Mars is beautiful. But I don't think Mars is as cool as Mars. What are your hobbies?

H: No. Where are you moved to USA from?

B: Japan. It is a great place to live.

# Future improvements

What we wanted to do but did not have time and resources for:

- Hybrid of generative and retrieval systems
- Sentence compression
- Attention on sentences
- Dataset augmentation (translate to another language and back)
- Memory and speed optimization
- Critic to evaluate dialog accordance with persona
- ...