Dialogue generation with transformers

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Abstract

- Try to create an AI agent giving meaningful responses on dialogues.
- Use transformer and self-attention architecture instead of RNN.
- Trained with OpenSubtitles dateset.

Dataset

OpenSubtitles is an open corpus with movies subtitles in different languages.

The dataset used to train the agent are the english subtitles from OpenSubtitles:

- $\sim 140 M \text{ utterances}$
- Dirty dataset. Metadata from the movie, comments not related to dialogues, ...
- Scripts from PolyAI used to clean the dataset
- Trained with just the previous sentence in the context.

```
{
"file_id": "lines-aaa",
"context": "They always do.",
"response": "Hmm, but they don't.",
"context/0": "He will.",
"context/1": "What then?",
...
}
```

Attention is all you need

- Traditional **Encoder-Decoder** architecture without RNN or convolutional networks
- Based on attention concept. Learning of Query (**Q**), Key (**K**) and Value(**V**) matrices
- Stacked transformers for encoder and decoder and multi head attention to attend to different information
- Positional encoding to give relative position information to tokens. In the paper a sin function added to embedding
- For decoder an extra **masked attention** layer is used to learn the language model. It cannot look forward so it masks attention for not generated tokens

Transformer Architecture

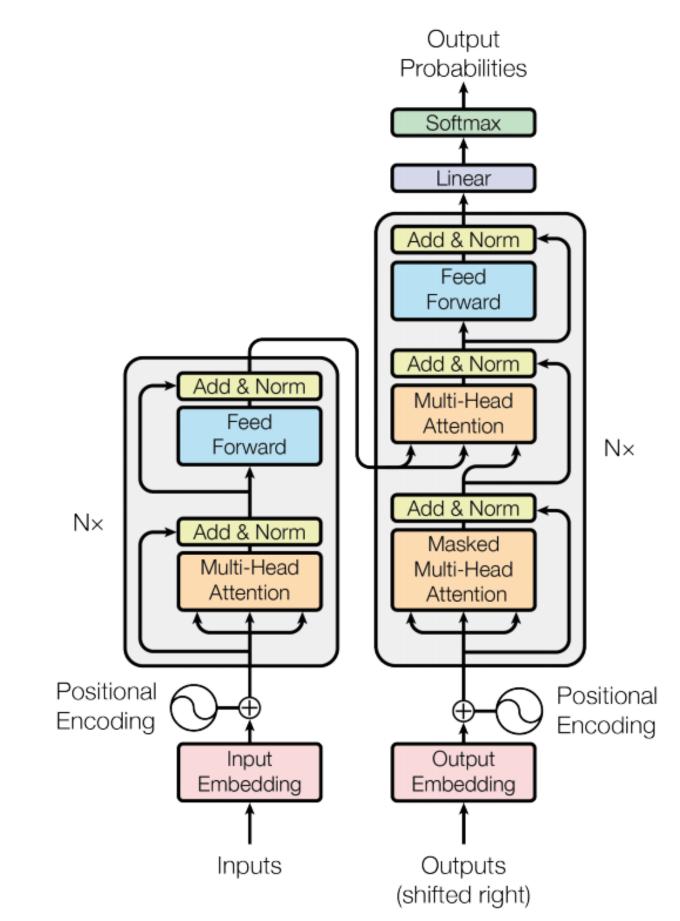


Figure: Transformer architecture

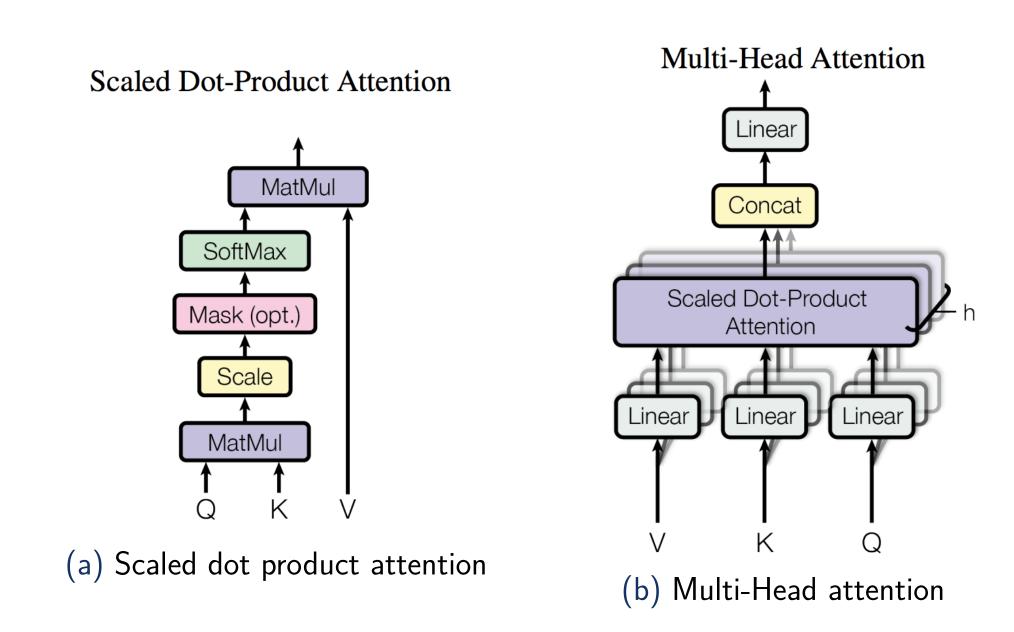


Figure: Attention

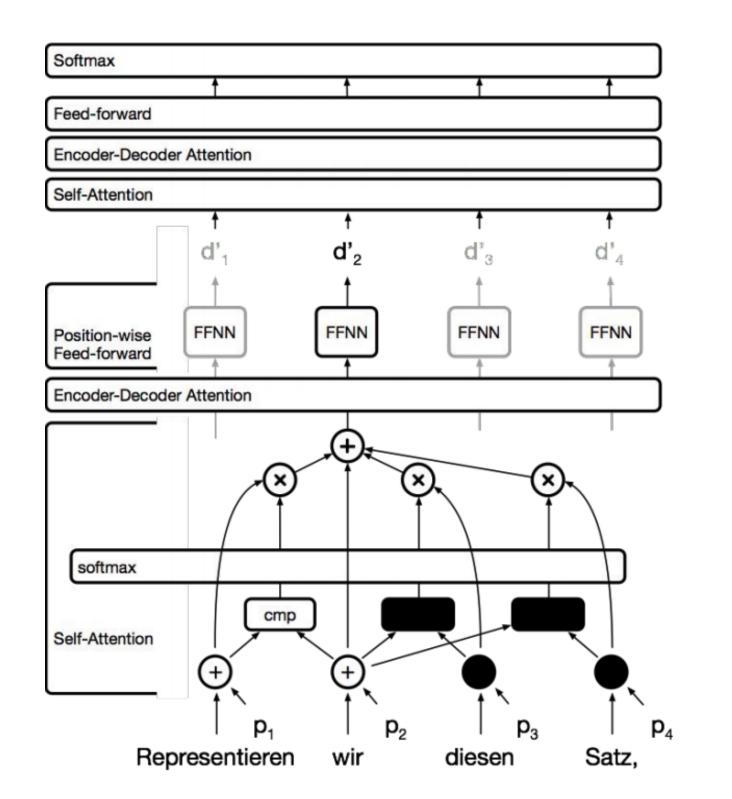


Figure: Decoder architecture

Transformer implementation and outcomes

Implementation based on *Transformer model for language understanding* from TensorFlow.

- Subword tokenizer from TensorFlow Dataset library
- Token embedding using Keras embedding layer
- 4 encoder layers
- 4 decoder layers
- Sublayers and embedding outputs of size 128
- Feed Forward inner layer of size 512
- 8 attention heads

Small train dataset

Trained with 78.000 utterances and 20 EPOCHS. 2 hr in a personal computer:

- Context: Hello I am John.
- Predicted response: I'm not gonna be a doctor.
- Context: Who are you?
- Predicted response: I am a bit much.
- Context: I want to eat a sandwich
- Predicted response: I want to be a doctor!

Big train dataset

Trained with \sim 2M utterances and 10 EPOCHS. 10 hr in the cloud and a GPU:

- Context: My name is John.
- Predicted response: I'm sorry...... Achi. Achi.
- Context: Good morning.
- Context: I am in my car
- Predicted response: What the fuck are you talking about??????????
- Context: I am going to the hospital
- Predicted response: What the fuck are you talking about??????????

Big train dataset

Trained with the recommended architecture from the paper. The same but 6 encoder and decoder layers, outputs of size 512 and FF inner layer of size 2048. The response is the same for all contexts.

Trained with $\sim 2M$ utterances and 3 EPOCHS. 9 hr in the cloud and a GPU:

- Context: My name is John.Context: Good morning.
- Context: I am in my car
- Context: I am going to the hospital
- Predicted response: I don't know what you're talking about.
- about.with.with.with.about.with.with.with.with.

Evaluation

The main way to evaluate the results in dialogue generation is with human judgement, comparing the result of the agent with other responses generated from other agents. I cold not evaluate my results this way.

Conclusion & Future Work

- It seems the results keep getting worst as it gains more information.
- Underfitting with less information?
- It seems the model converges to give dull responses, which could make sense from a dialogue prospective if a similar sentence to the context does not exist in the dataset
- The model learns a "language model" that creates responses grammatically correct fast but it does not give meaningful responses in the dialogue.
- Hard to deal with big datasets for time and space constraints.
- Training the agent with a big dataset with the recommended architecture properly.

References

- Vaswani, Attention is all you need
- Tensorflow Transformer model for language understanding https:

//www.tensorflow.org/tutorials/text/transformer

- OpenSubtitles corpus
- http://opus.nlpl.eu/OpenSubtitles-v2018.php
- Poly AI Conversational Datasets https:

//github.com/PolyAI-LDN/conversational-datasets