MSiA490 SEC20/28 Text Analytics Lab 3 - Word2Vec and BERT

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Overview

- Discussions on Word2Vec and BERT
- Using BERT
- Quiz

Question1

How is a Skip-gram model trained? What is its training objective? How is BERT trained? What are its training objectives?

Question 1

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Notes 1

- The training objective of the Skip-gram model is to maximize the average log probability of the context words.
- The training of BERT has two stages: pretraining and fine-tuning. The pretraining stage involves guessing the correct masked out word and predicting if the second sentence follows the first one.

Question 2

What makes BERT different from Word2Vec models?

Question 2

What are some differences between BERT and Word2Vec?

Notes 2

- Word2Vec vector provides a vector for each token/word that encodes the meaning of that token/word.
- BERT provides contextual word representations that encode different meanings under different context.

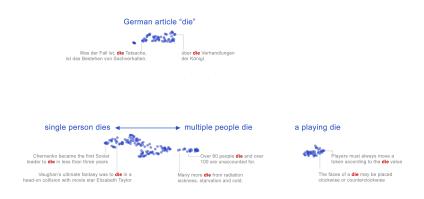


Figure: Embeddings for the word "die" in different contexts.¹

 $^{^{1}}$ "Visualizing and Measuring the Geometry of BERT". Coenen et al.

Question 3

What are the advantages of pre-training for BERT?

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Notes

Pretraining BERT provides us with a generalized **language model** that can be used later for a variety of natural language processing tasks. The amount of training data required for those down streaming tasks does not need to be huge as the pre-trained model contains significant information on the language itself.

Question 4

What are some down streaming tasks we can perform by fine-tuning BERT?

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Notes

- Single-sentence classification tasks
- Single-sentence tagging tasks
- Question answering tasks

Using Bert

Hugging Face's Transformers

Install

pip install transformers

Example usage

- > from transformers import BertTokenizer, BertModel
- > tokenizer = BertTokenizer.from_pretrained("bert-baseuncased")
- > model = BertModel.from_pretrained("bert-base-uncased")
- > inputs = tokenizer("Hello world!", return_tensors="pt
 ")
- > outputs = model(**inputs)

Using Bert

Other options/resources

- An alternative bert-as-service: https://github.com/hanxiao/bert-as-service
- A potentially outdated but still good tutorial on Hugging Face's Transformers:

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https://mccormickml.com/2019/05/14/BERT-word-embeddings-tutorial/
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Quiz

Which of the following can be somehow realized with BERT but **not** with Word2Vec?

- A Computing vectorized sentence representations
- B Computing vectorized word representations
- C Computing contextualized word representations
- D All of the above

Quiz

Which of the following is part of the pretraining stage for BERT?

- A Single-sentence tagging
- B Masked-out word prediction
- C Next sentence prediction
- D A and B
- E B and C
- F A, B and C