Question Answering

Video: Week 3 Overview 6 min

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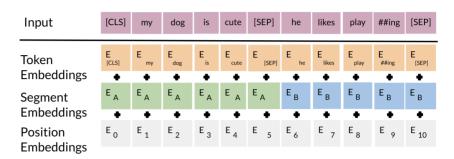
- Reading: Week 3 Overview 10 min
- Video: Transfer Learning in NLP
 7 min
- Reading: Transfer Learning in NLP
- Video: ELMo, GPT, BERT, T5
- Reading: ELMo, GPT, BERT, T5
 10 min
- Video: Bidirectional Encoder Representations from Transformers (BERT)
 4 min
- Reading: Bidirectional Encoder Representations from Transformers (BERT)
- Video: BERT Objective
- Reading: BERT Objective
 10 min
- Video: Fine tuning BERT 2 min
- Reading: Fine tuning BERT 10 min
- Video: Transformer: T5 3 min
- Reading: Transformer T5 10 min
- Video: Multi-Task Training Strategy 5 min
- Reading: Multi-Task
 Training Strategy
 10 min
- Video: GLUE Benchmark 2 min
- Reading: GLUE Benchmark
 10 min
- Video: Question Answering 2 min
- Reading: Question
 Answering
 10 min
- Lab: SentencePiece and BPE
- Reading: Content Resource

Assignment

Programming Assignment:
Question Answering
3h

BERT Objective

We will first start by visualizing the input.

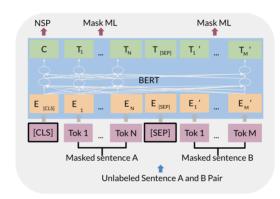


The input embeddings are the sum of the token embeddings, the segmentation embeddings and the position embeddings.

The input embeddings: you have a CLS token to indicate the beginning of the sentence and a sep to indicate the end of the sentence

The segment embeddings: allows you to indicate whether it is sentence a or b.

Positional embeddings: allows you to indicate the word's position in the sentence.



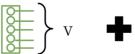
- [CLS]: a special classification symbol added in front of every input
- [SEP]: a special separator token

The C token in the image above could be used for classification purposes. The unlabeled sentence A/B pair will depend on what you are trying to predict, it could range from question answering to sentiment. (in which case the second sentence could be just empty). The BERT objective is defined as follows:

Objective 1: Multi-Mask LM Objective 2: Next Sentence Prediction

Loss: Binary Loss

Loss: Cross Entropy Loss



You just combine the losses!