

# BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding

Devlin et al.

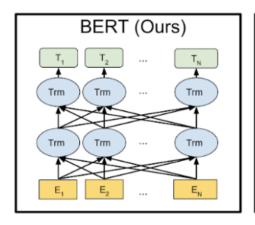
Jungsoo Park

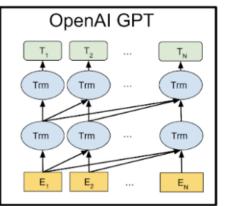
Data Mining & Information Systems Lab.

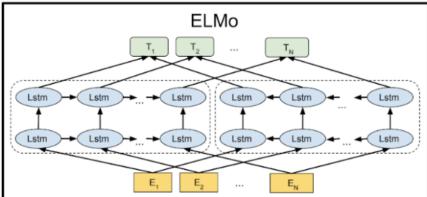
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### Natural Language Processing's Imagenet



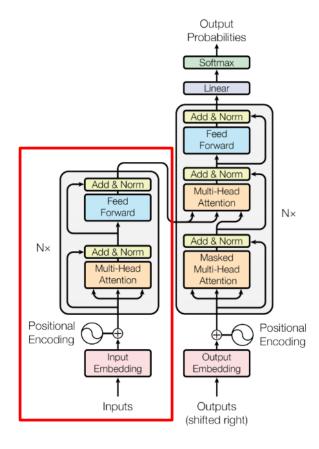








#### **Transformer Model**





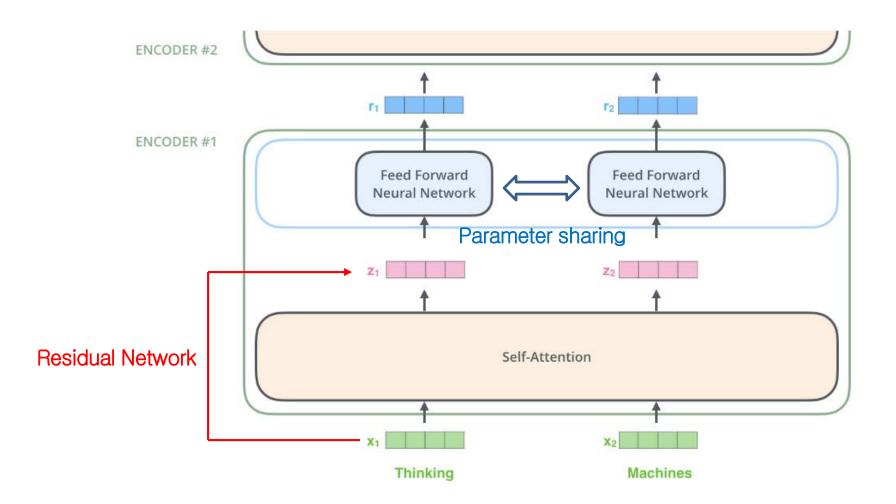


Figure from <a href="http://jalammar.github.io/illustrated-transformer/">http://jalammar.github.io/illustrated-transformer/</a>

## 2 Recap



Input	Thinking	Machines	
Embedding	X <sub>1</sub>	X <sub>2</sub>	
Queries	<b>q</b> <sub>1</sub>	<b>q</b> <sub>2</sub>	Wa
Keys	<b>k</b> <sub>1</sub>	k <sub>2</sub>	Wĸ
Values	V <sub>1</sub>	V <sub>2</sub>	Wv

Figure from <a href="http://jalammar.github.io/illustrated-transformer/">http://jalammar.github.io/illustrated-transformer/</a>

# 2 Recap



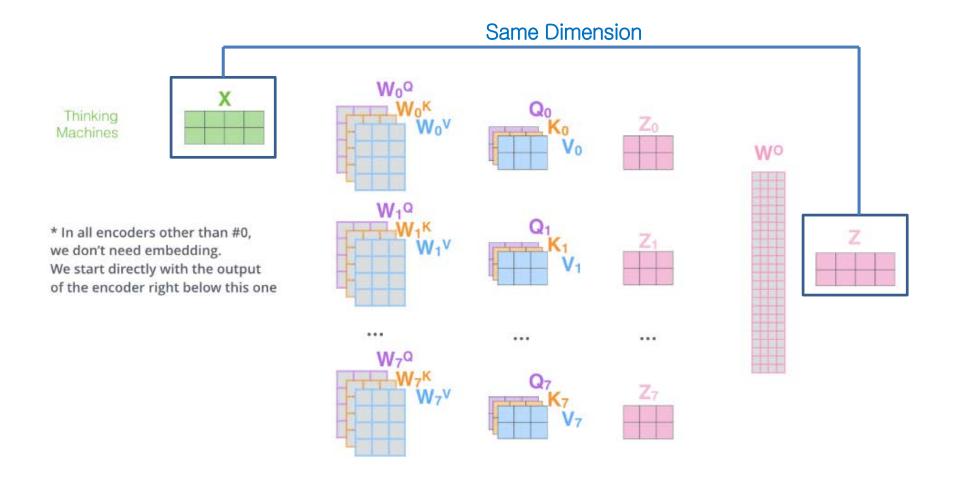
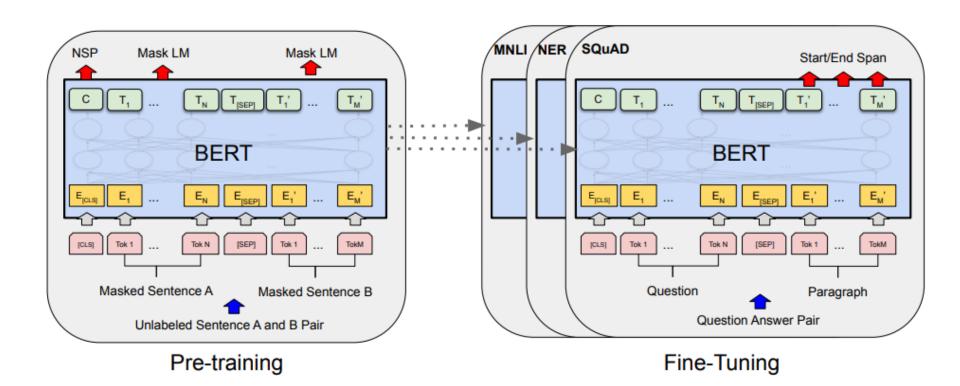


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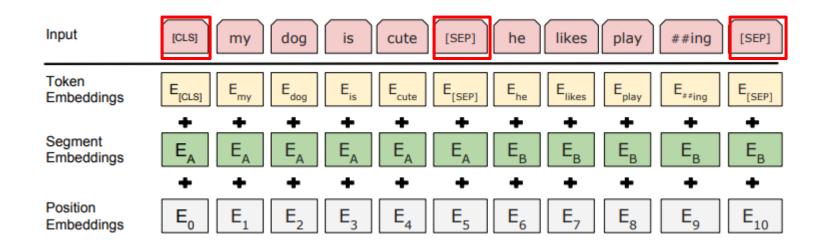


#### General Architecture





### Input Representation



Token Embedding : WordPiece Embedding

Segment Embedding: Corresponding Sentence Embedding

Position Embedding: Learned Positional Embedding(up to 512 tokens)



### Pre-training Task #1: Masked Language Modeling

Masking some input tokens, and predict only those masked ones(Cloze)

- Masking 15% of WordPiece tokens in each sequence
- Rather than always replacing the token with [MASK], sometimes replace the given word with another word, and also sometimes keep the word unchanged
- This'll close the gap between pre-training phase and fine-tuning phase
- May converge slowly



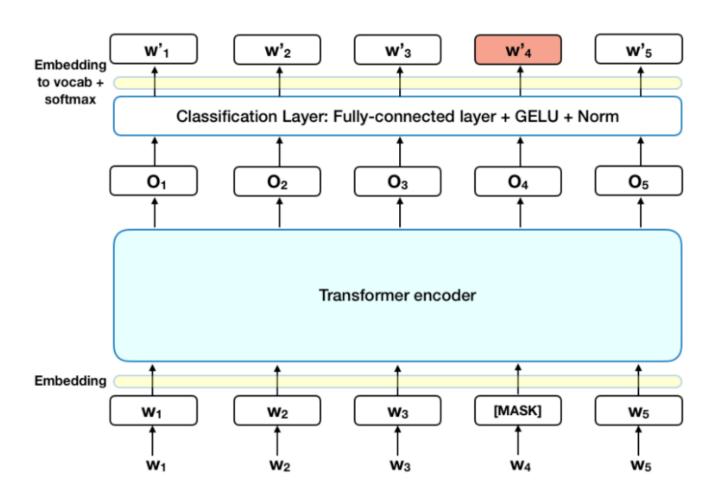


Figure from <a href="https://www.lyrn.ai/2018/11/07/explained-bert-state-of-the-art-language-model-for-nlp/">https://www.lyrn.ai/2018/11/07/explained-bert-state-of-the-art-language-model-for-nlp/</a>



### Pre-training Task #2 :Next Sentence Prediction

#### Understanding the relationship between two senteces

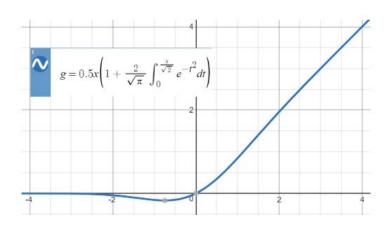
- Binarized next sentence prediction task
- In monolingual text, 50% of the time, next sentence is the actual sentence which follows the previous one, whereas for the 50% of the time, the random sentence from corpus is given.



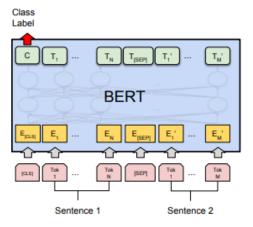
### Pre-training Procedure

#### Understanding the relationship between two senteces

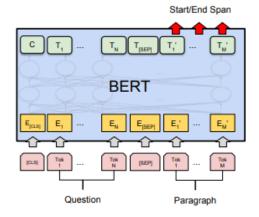
- Document Level Corpus was used for training(Book Corpus, Wikipedia)
- Sentences are generally longer than the usual sentences that we know
- Each sequence(combined of two sentences)'s length is smaller than 512
- GELU activation was used



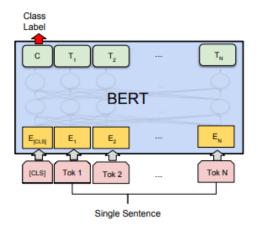




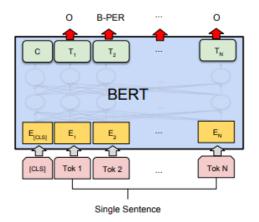
(a) Sentence Pair Classification Tasks: MNLI, QQP, QNLI, STS-B, MRPC, RTE, SWAG



(c) Question Answering Tasks: SQuAD v1.1



(b) Single Sentence Classification Tasks: SST-2, CoLA



(d) Single Sentence Tagging Tasks: CoNLL-2003 NER



# What Does BERT Look At? An Analysis of BERT's Attention

Kevin Clark et al.

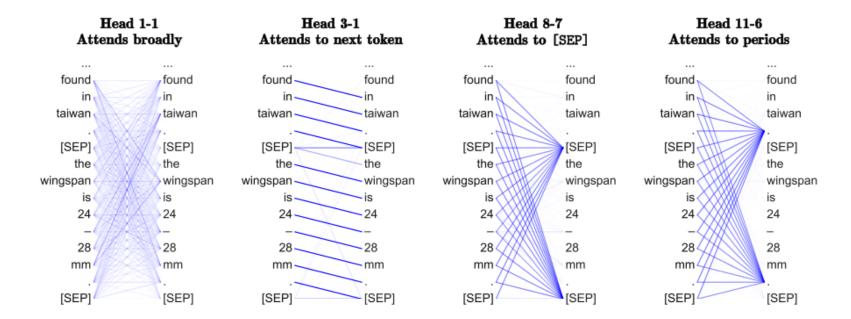
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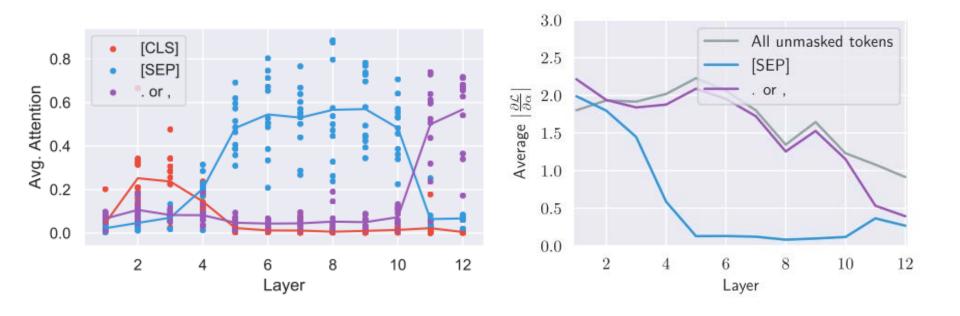




- In earlier layers, attention heads tend to put their attention to previous, or next token of given token's position.
- In the higher layers, entropies of attention distribution gets smaller.
- Substantial amount of BERT's attention focuses on a few tokens ([SEP], [CLS], ",", ".")







- Rather than being an artifact of stochastic training, high attention would have been caused by systematic reason. (Because, special tokens don't get masked away)
- Another possible explanation is that [SEP] is used to aggregate segment level information, however doubtful. (Right Figure) => Sort of "no-op"