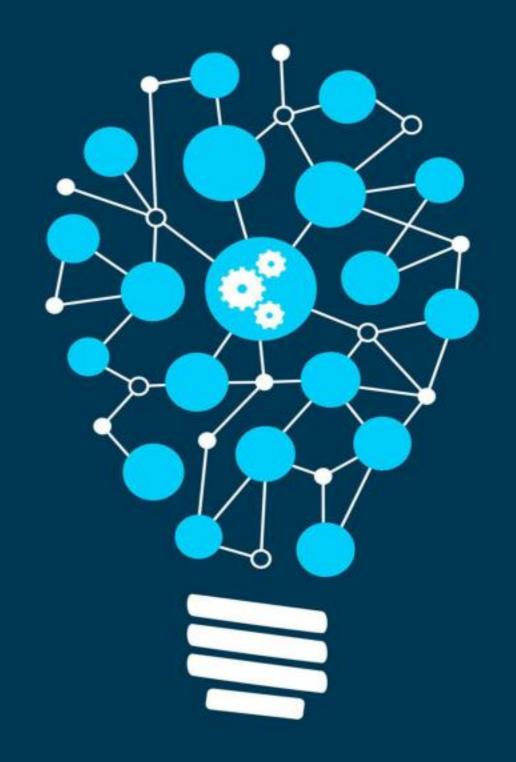


人工智能技术及应用

Artificial Intelligence and Application

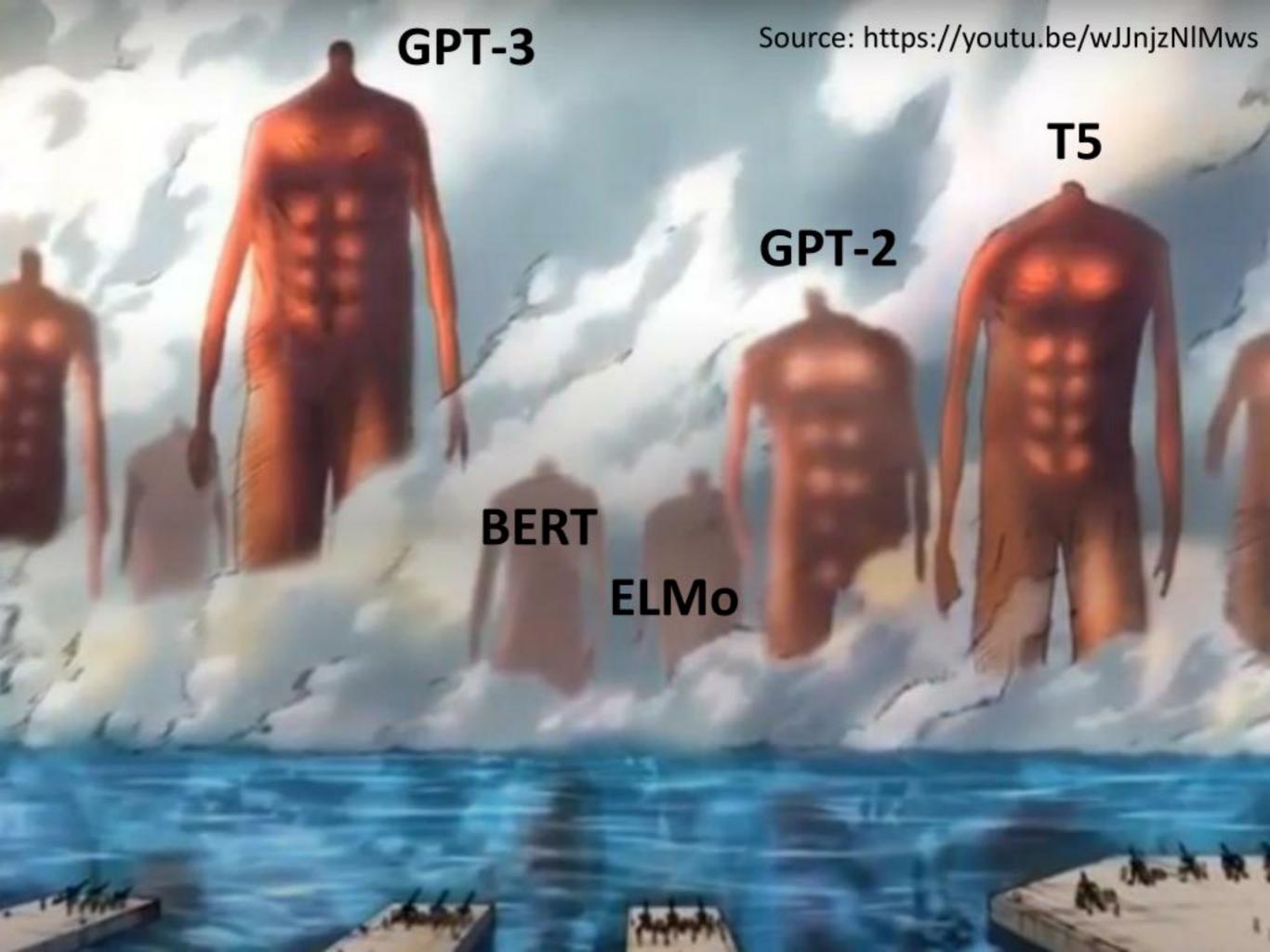
Self-Supervised Learning

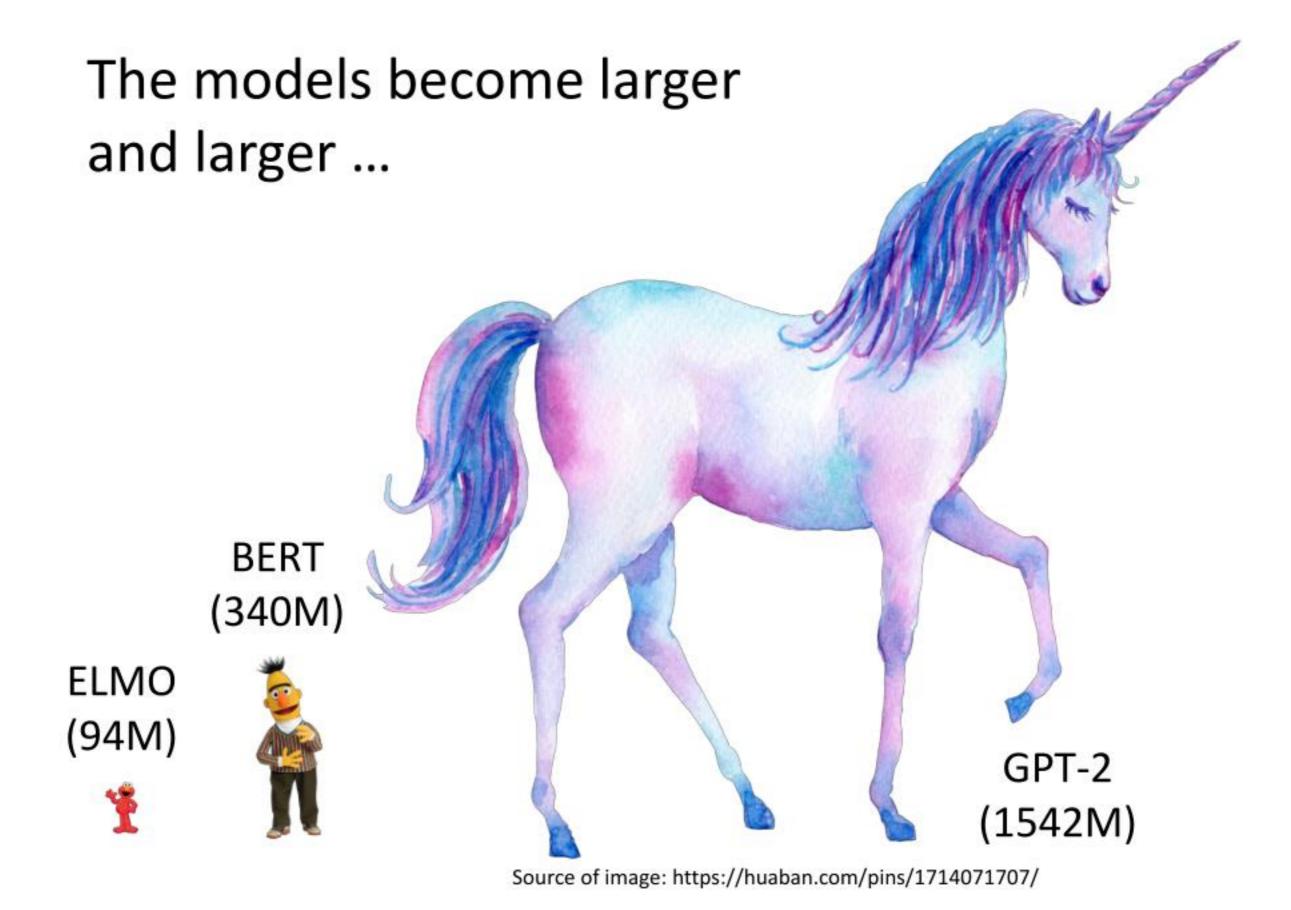


ERNIE (Enhanced Representation through Knowledge Integration)









The models become larger and larger ...

Turing NLG (17B)

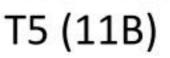
GPT-3 is 10 times larger than

Turing NLG. (175B)









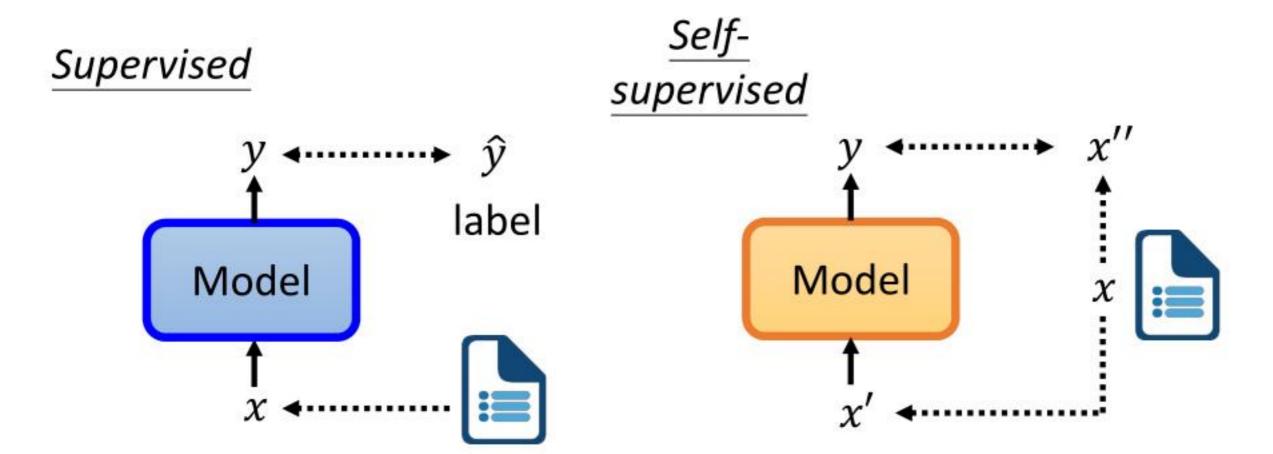


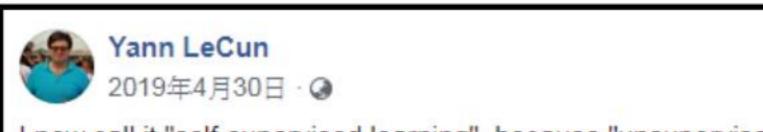
Outline





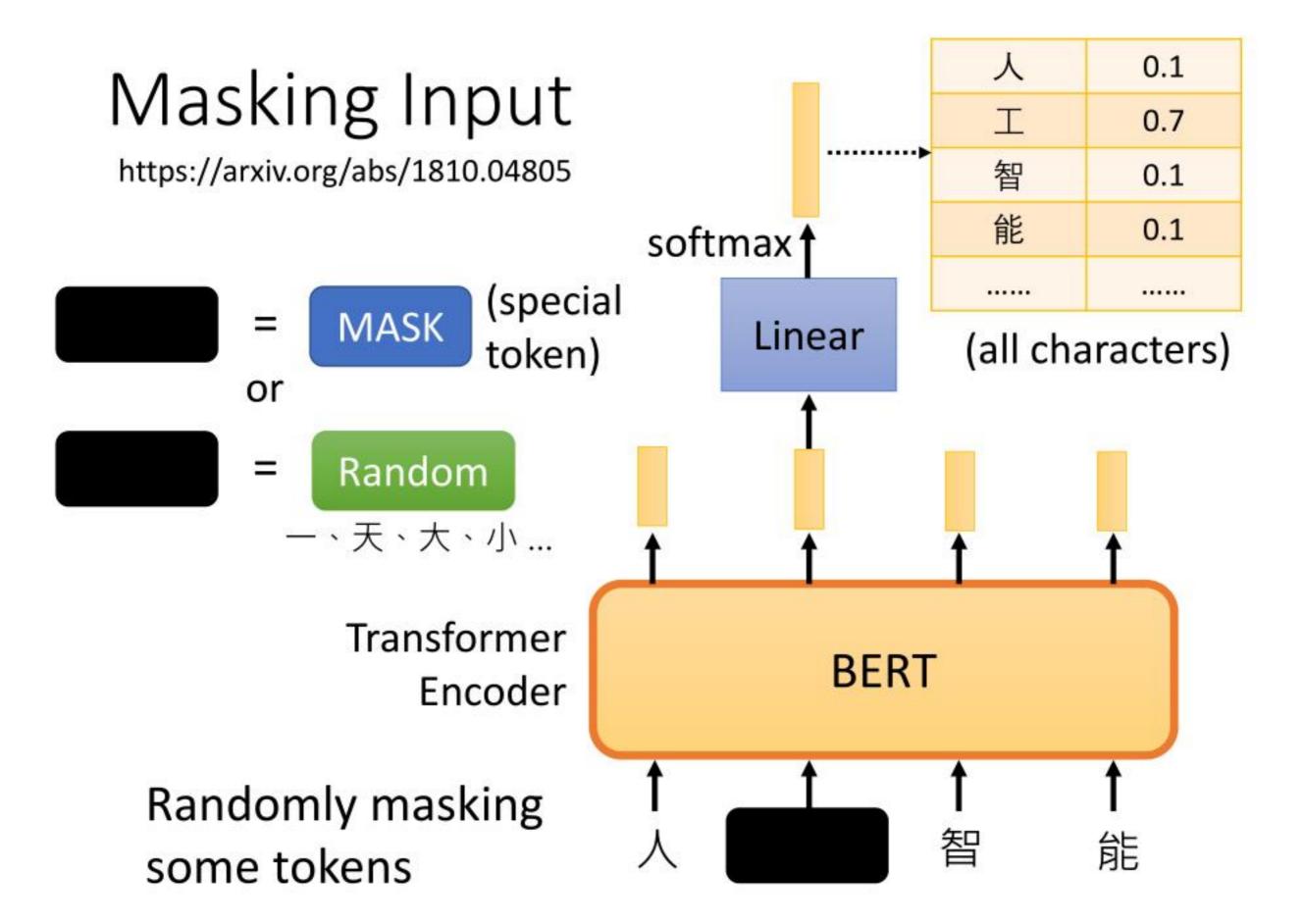
Self-supervised Learning





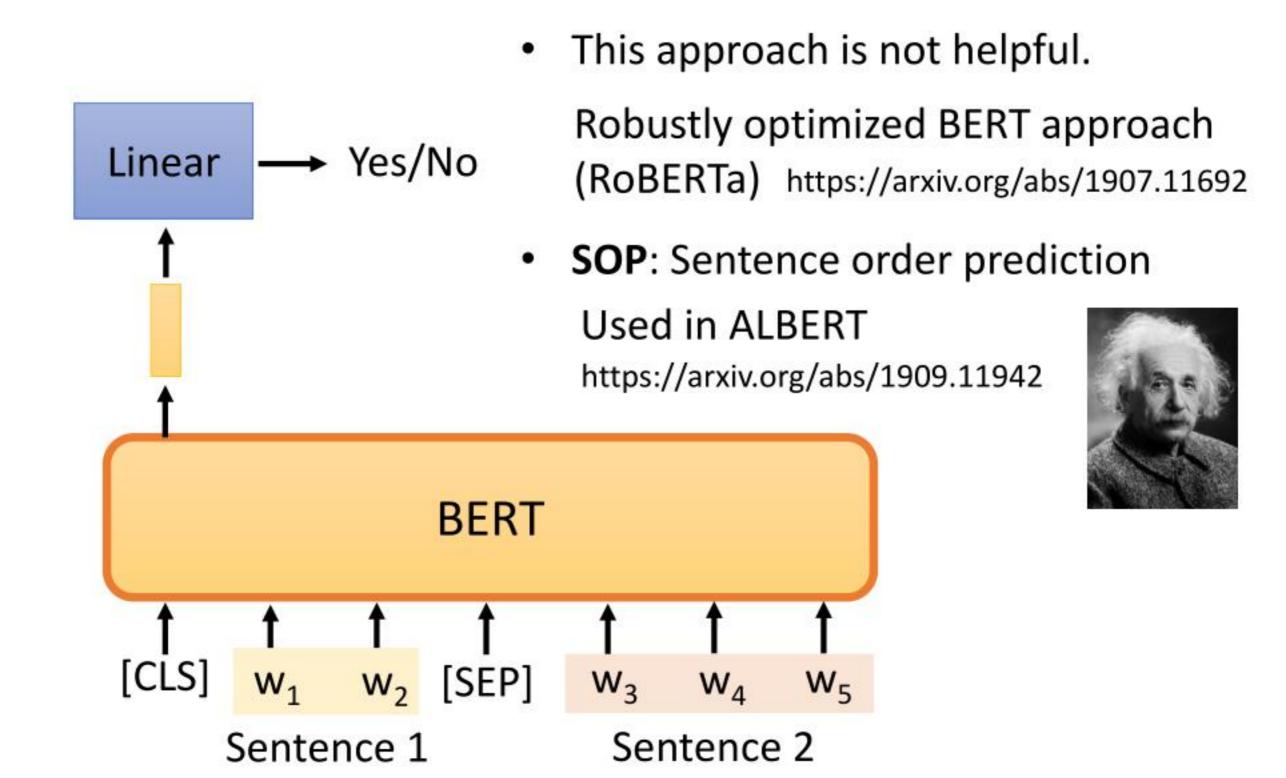
I now call it "self-supervised learning", because "unsupervised" is both a loaded and confusing term.

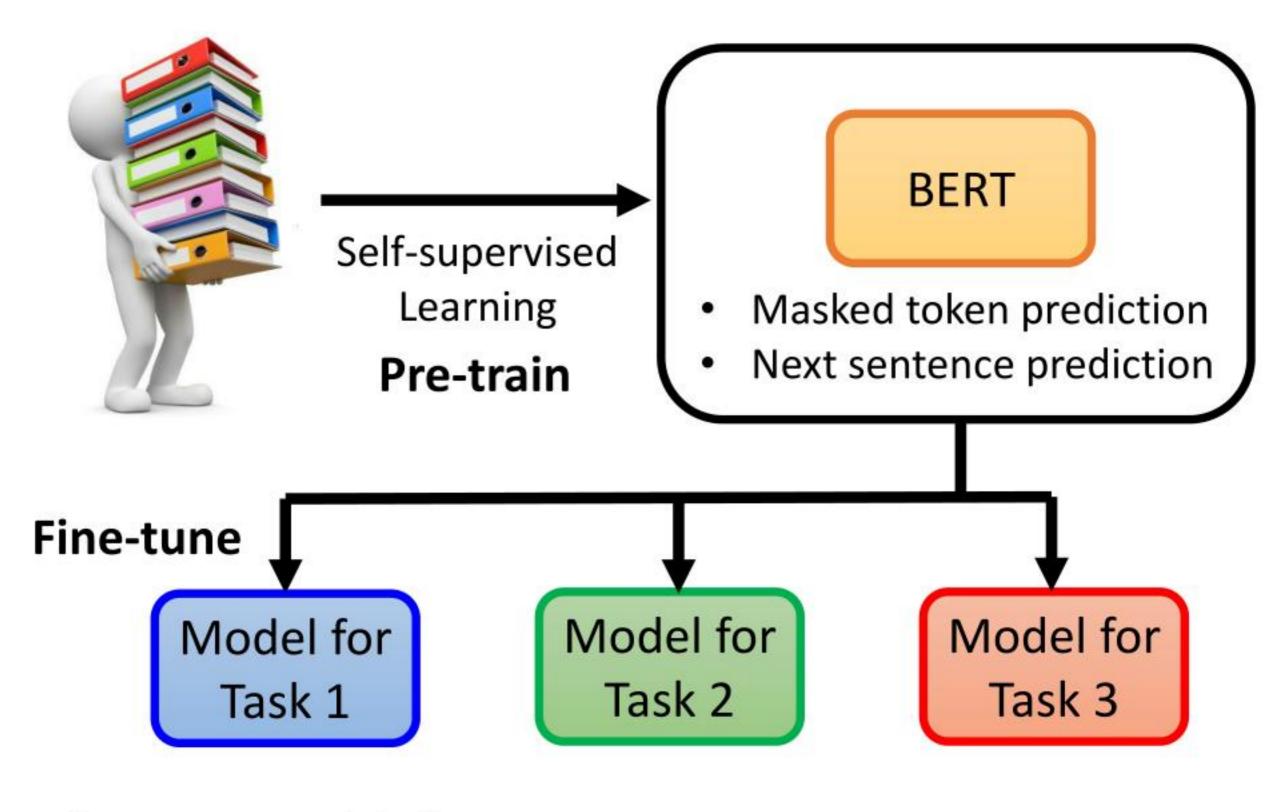
In self-supervised learning, the system learns to predict part of its input from other parts of it input. In other words a portion of the input is used as a supervisory signal to a predictor fed with the remaining portion of the input.



Ground truth Masking Input https://arxiv.org/abs/1810.04805 minimize cross softmax entropy (special MASK Linear token) or Random 一、天、大、小 ... Transformer **BERT** Encoder Randomly masking some tokens

Next Sentence Prediction





Downstream Tasks

- The tasks we care
- We have a little bit labeled data.

GLUE

General Language Understanding Evaluation (GLUE)

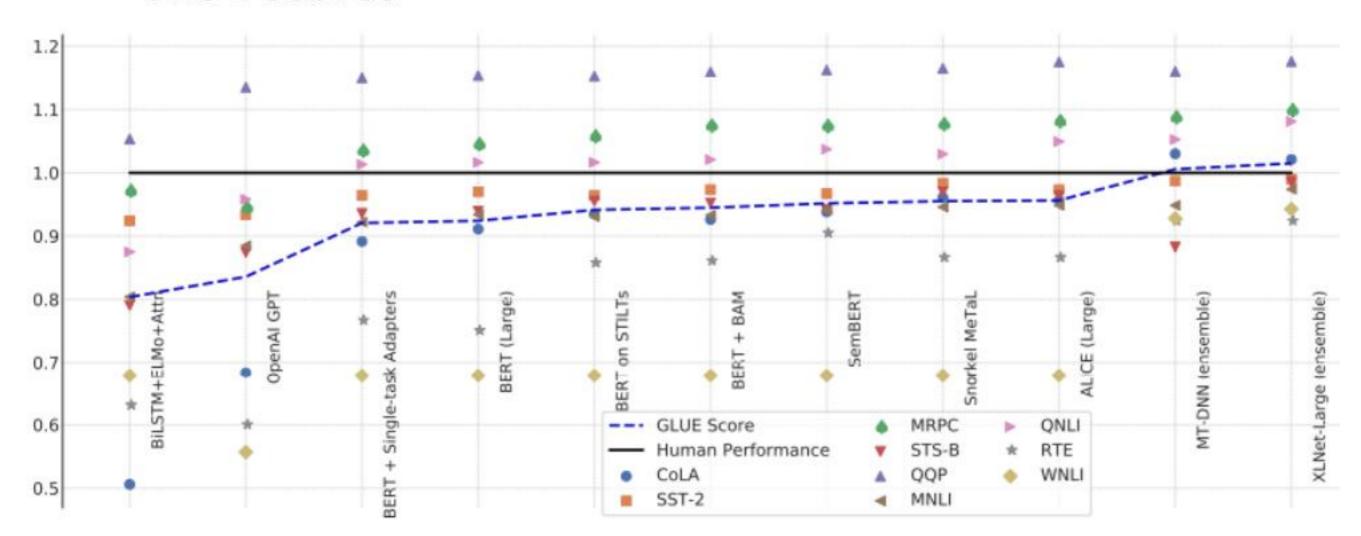
https://gluebenchmark.com/

- Corpus of Linguistic Acceptability (CoLA)
- Stanford Sentiment Treebank (SST-2)
- Microsoft Research Paraphrase Corpus (MRPC)
- Quora Question Pairs (QQP)
- Semantic Textual Similarity Benchmark (STS-B)
- Multi-Genre Natural Language Inference (MNLI)
- Question-answering NLI (QNLI)
- Recognizing Textual Entailment (RTE)
- Winograd NLI (WNLI)

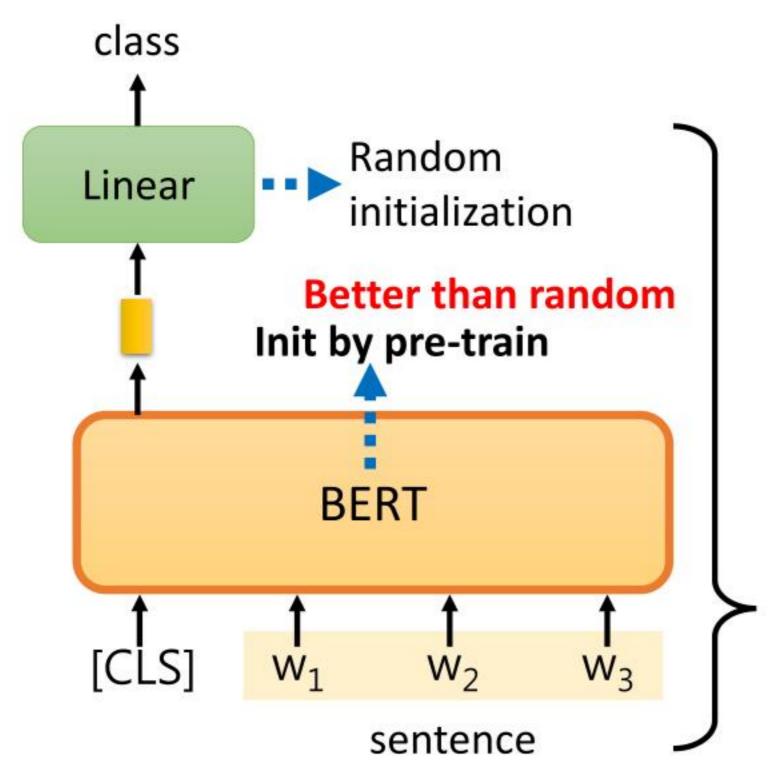
GLUE also has Chinese version (https://www.cluebenchmarks.com/)

BERT and its Family

GLUE scores



Source of image: https://arxiv.org/abs/1905.00537

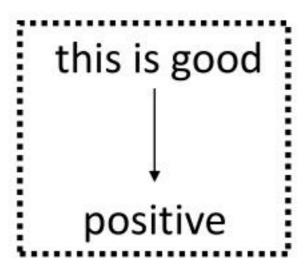


Input: sequence

output: class

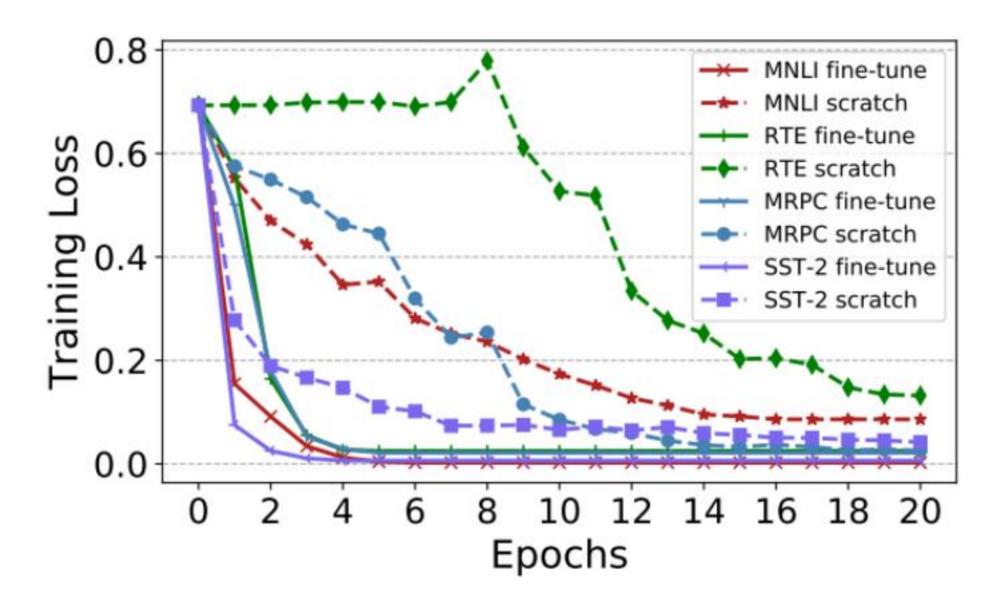
Example:

Sentiment analysis

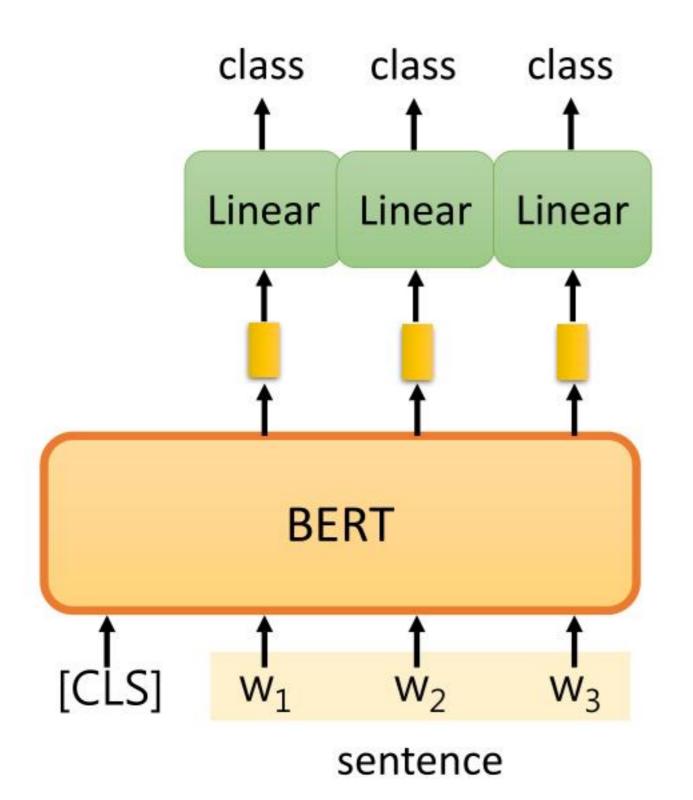


This is the model to be learned.

Pre-train v.s. Random Initialization (scratch)



Source of image: https://arxiv.org/abs/1908.05620

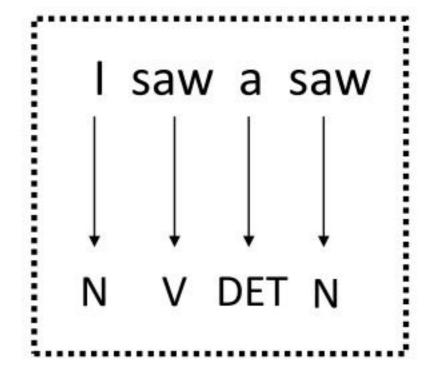


Input: sequence

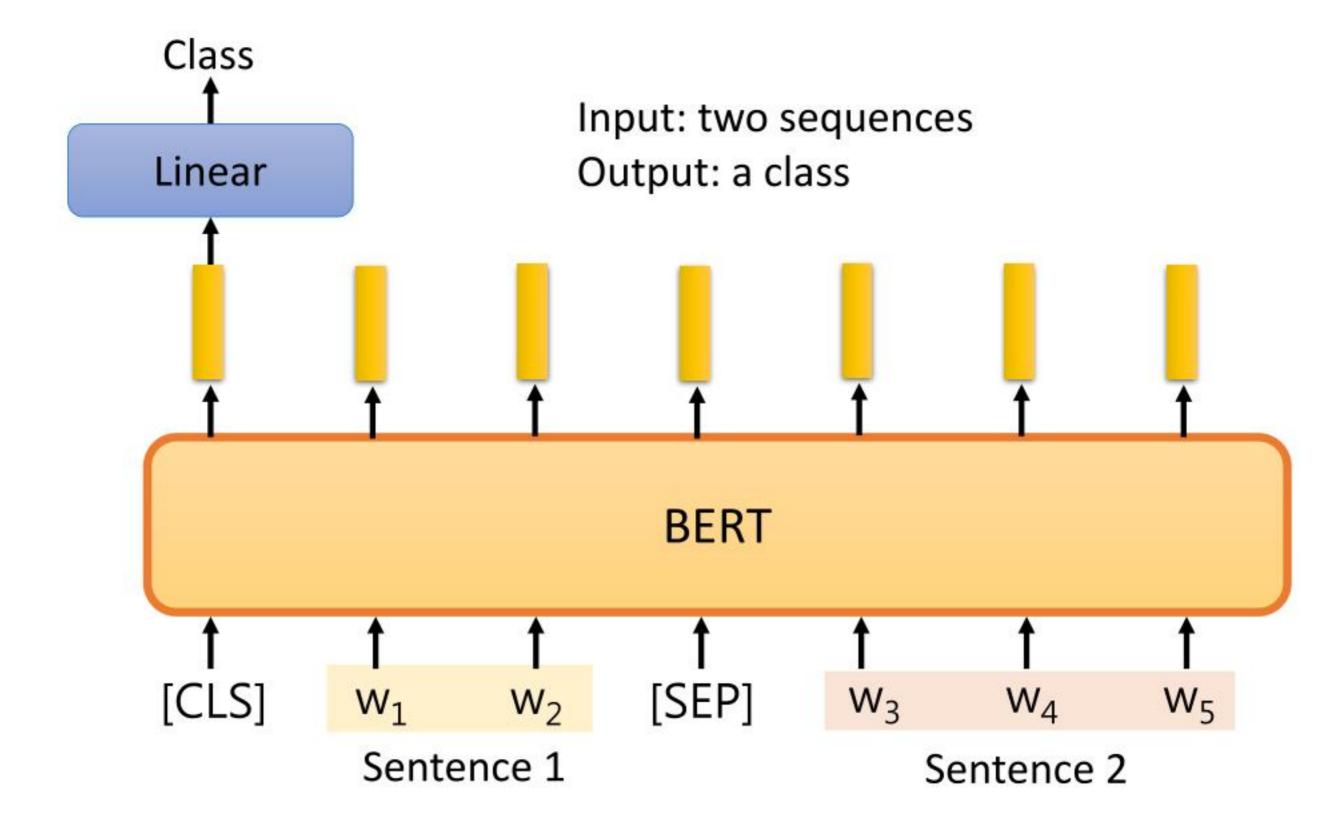
output: same as input

Example:

POS tagging



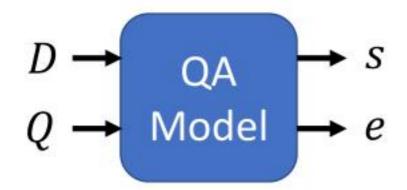
Input: two sequences Output: a class Example: contradiction Natural Language Inferencee (NLI) entailment neutral premise: A person on a horse Model jumps over a broken down airplane contradiction hypothesis: A person is at a diner.



 Extraction-based Question Answering (QA)

Document:
$$D = \{d_1, d_2, \dots, d_N\}$$

Query:
$$Q = \{q_1, q_2, \dots, q_M\}$$



output: two integers (s, e)

Answer:
$$A = \{d_s, \dots, d_e\}$$

In meteorology, precipitation is any product of the condensation of 17 spheric water vapor that falls under gravity. The main forms of precipitation include drizzle, rain, sleet, snow, graupel and hail... Precipitation forms as smaller droplets coalesce via collision with other rain drops or ice crystals within a cloud. Short, intense periods of rain 77 atte 79 cations are called "showers".

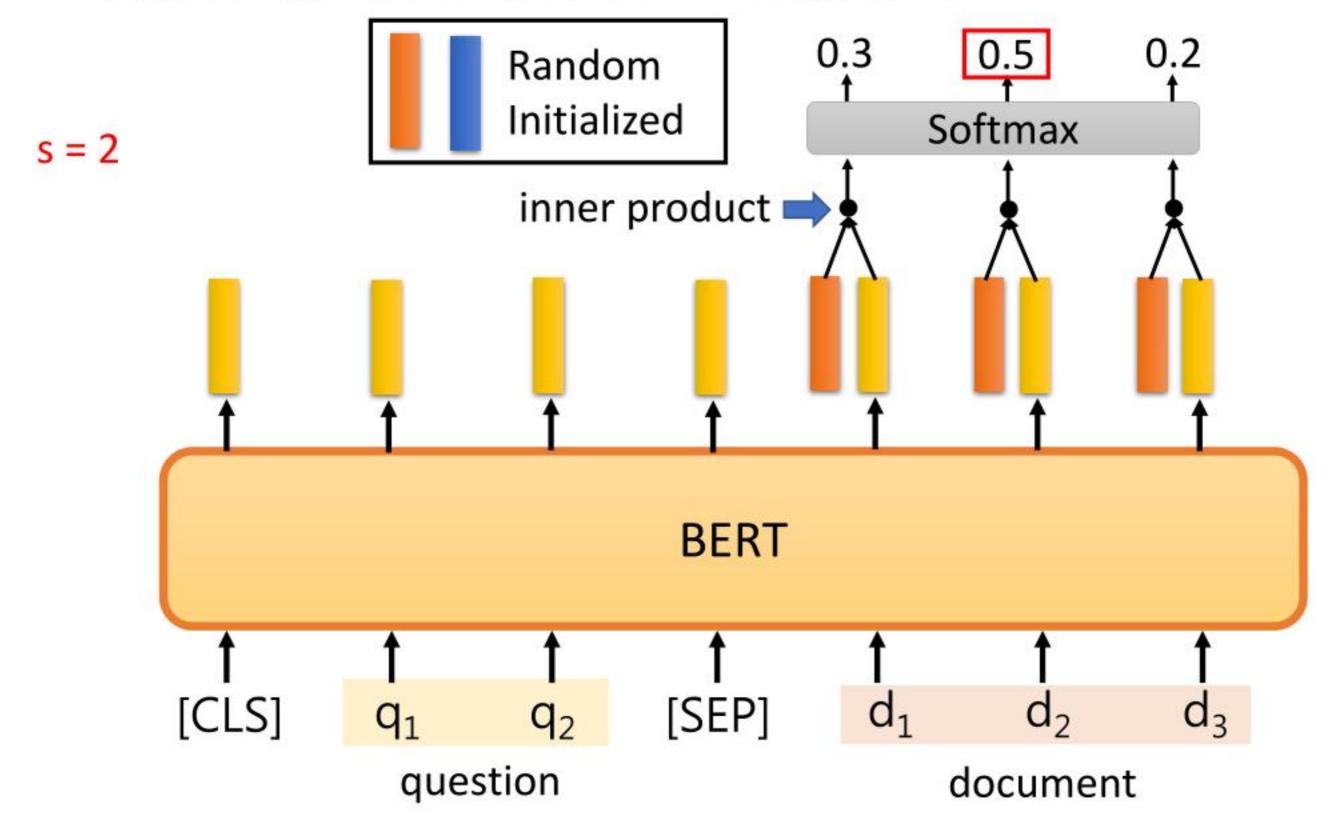
What causes precipitation to fall?

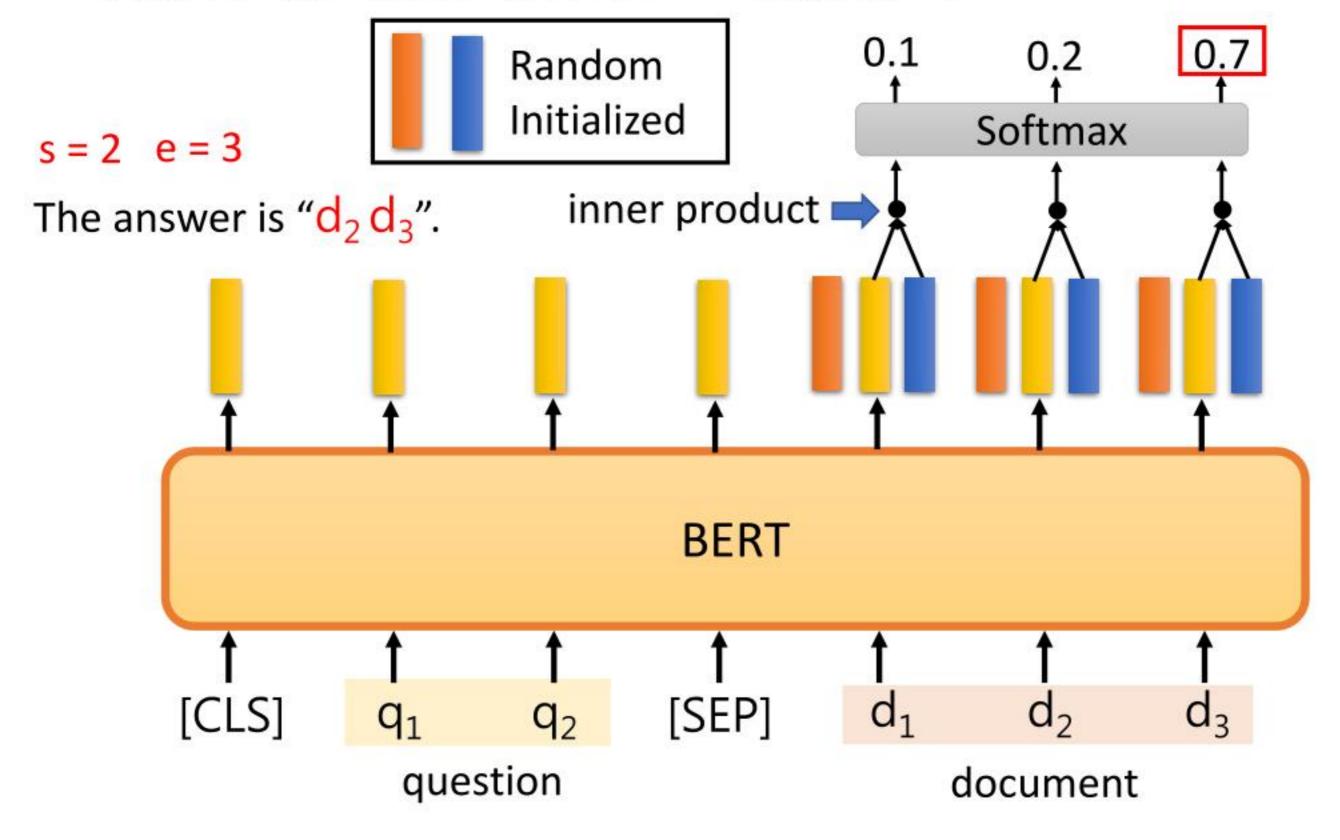
$$s = 17, e = 17$$

What is another main form of precipitation besides drizzle, rain, snow, sleet and hail? graupel

Where do water droplets collide with ice crystals to form precipitation?

$$s = 77, e = 79$$



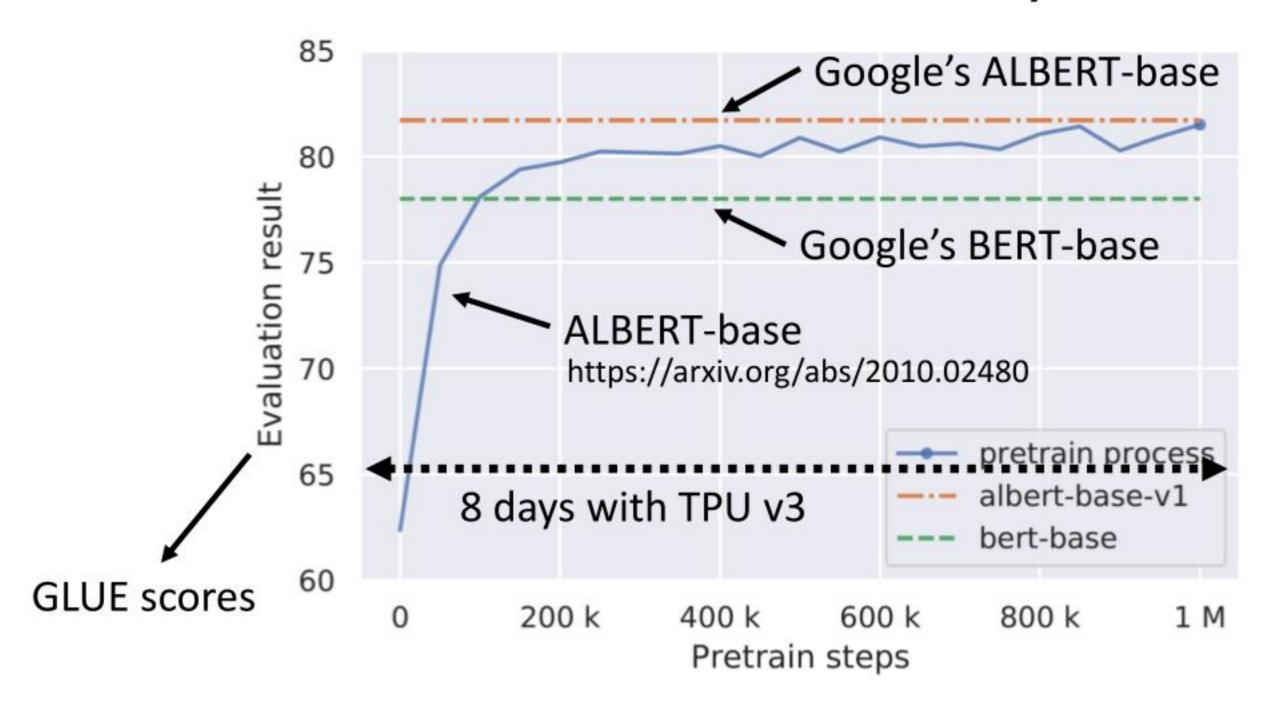




Training BERT is challenging!

Training data has more than 3 billions of words.

3000 times of Harry Potter series



BERT Embryology (胚胎学)

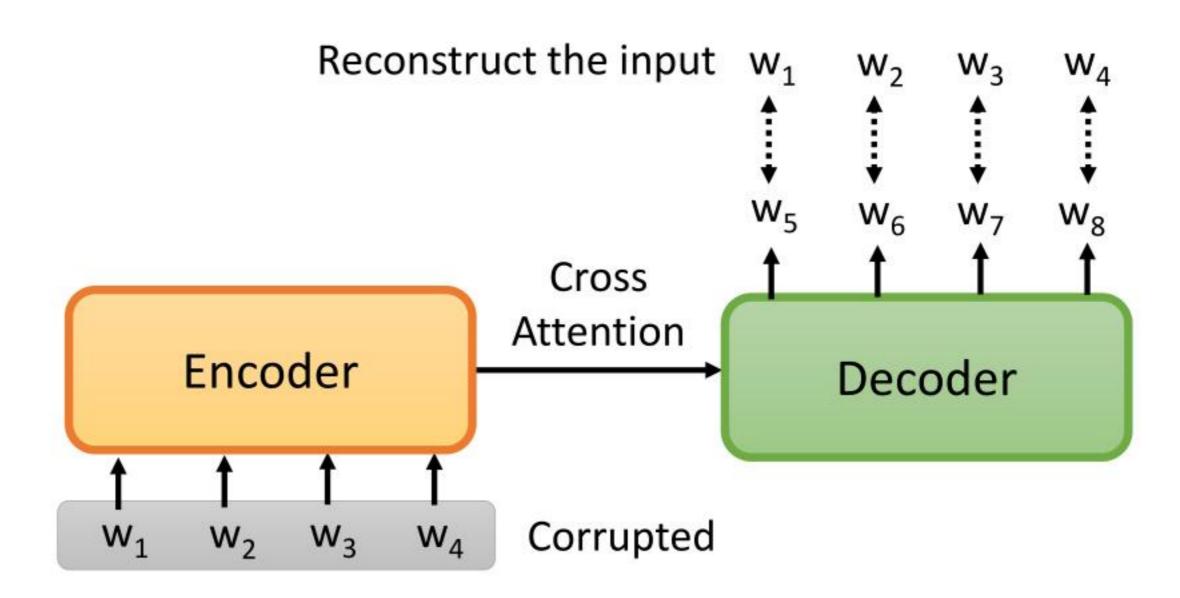
https://arxiv.org/abs/2010.02480



When does BERT know POS tagging, syntactic parsing, semantics?

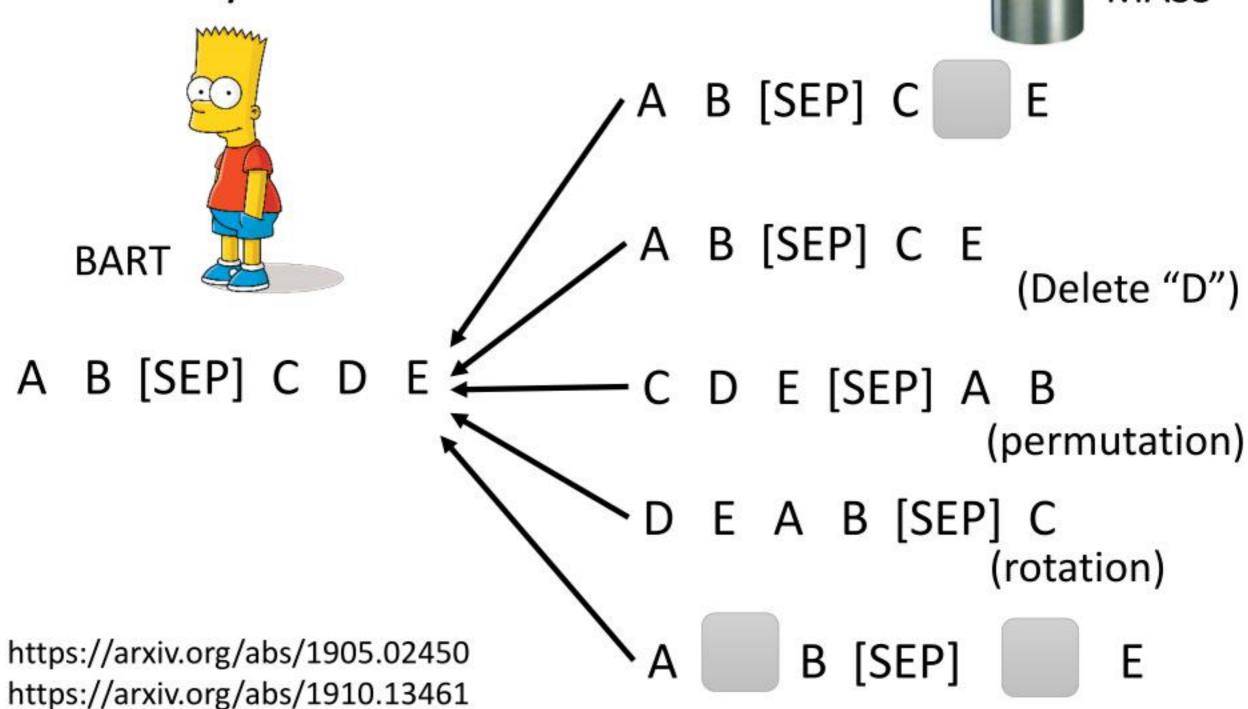
The answer is counterintuitive!

Pre-training a seq2seq model



MASS / BART





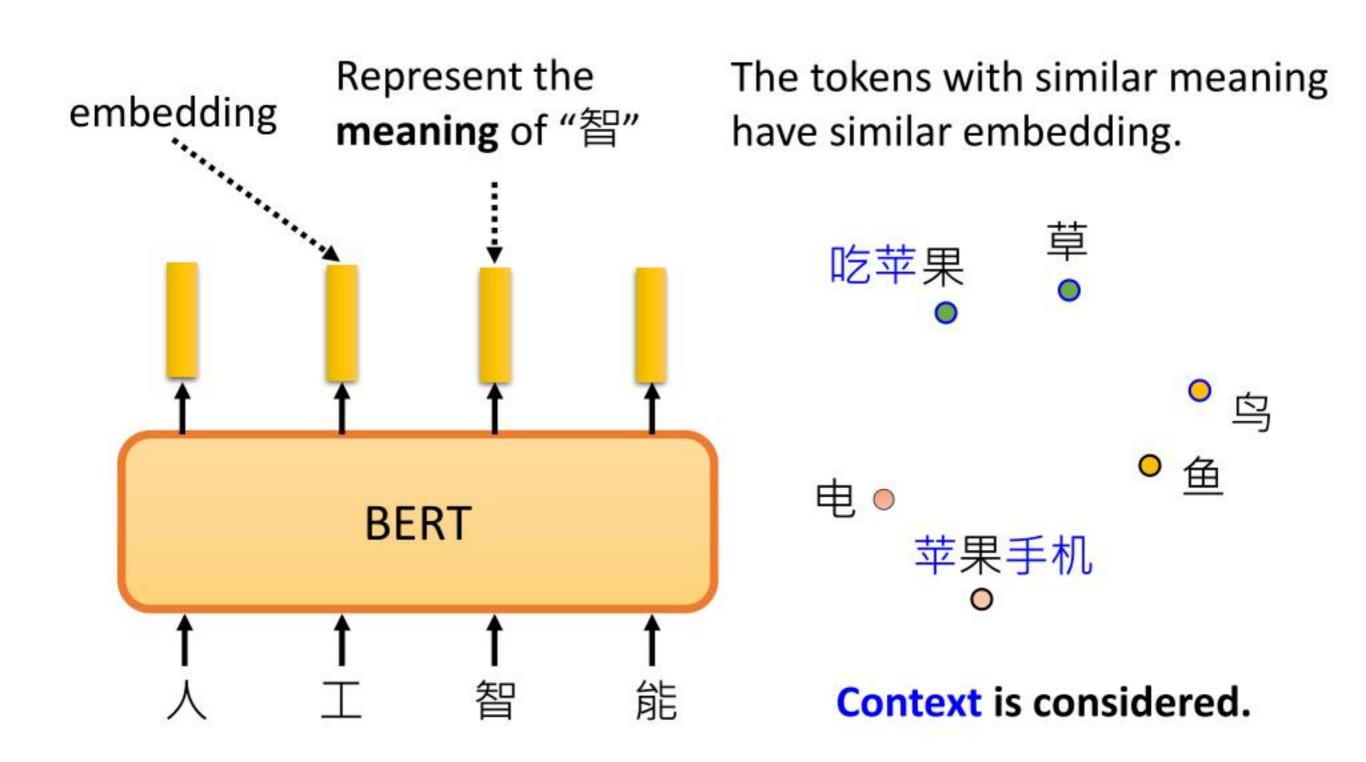
Text Infilling

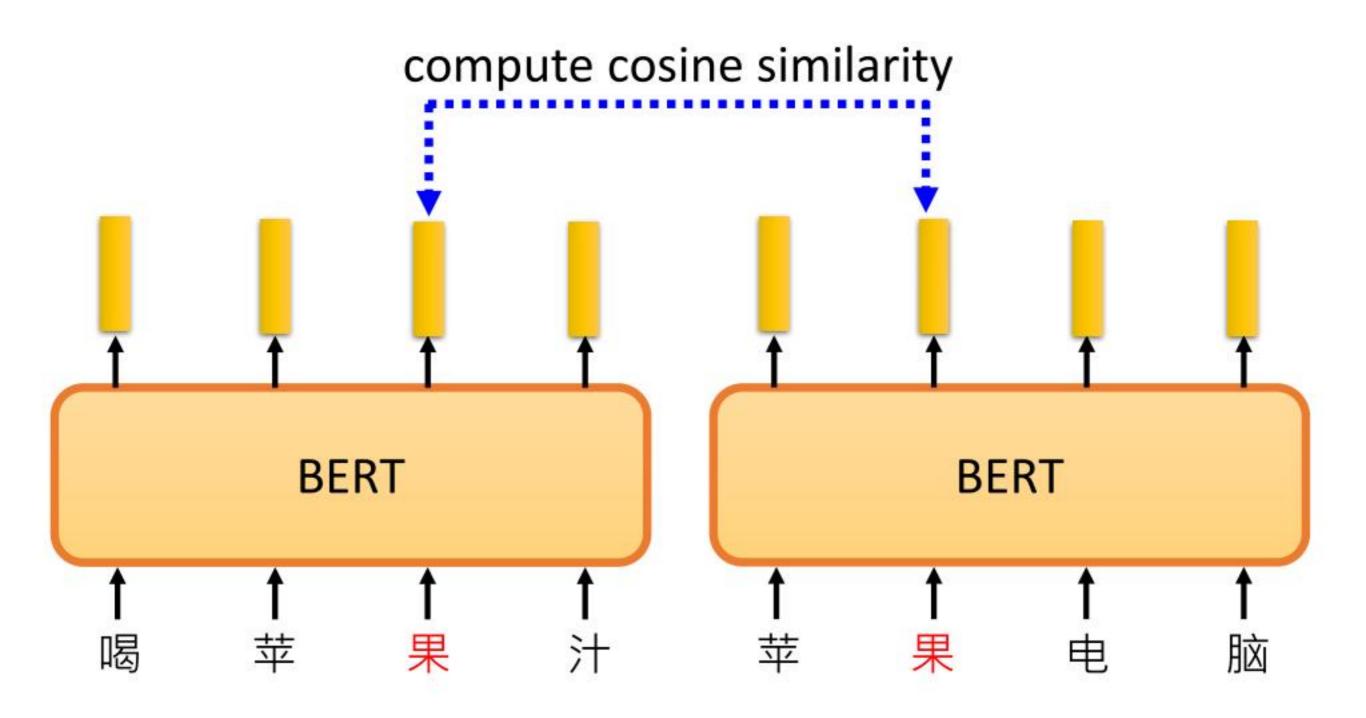
T5 – Comparison

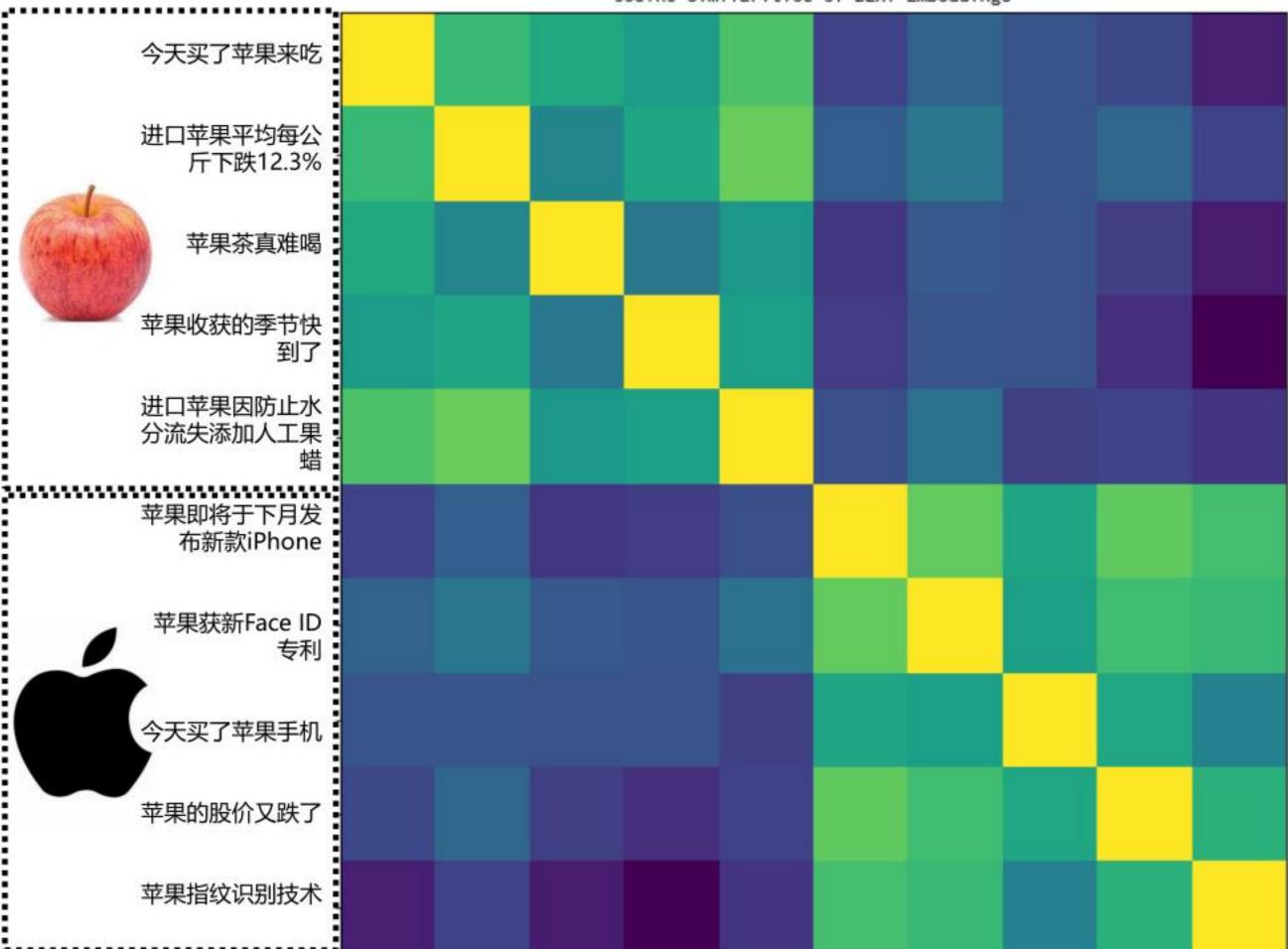
- Transfer Text-to-Text Transformer (T5)
- Colossal Clean Crawled Corpus (C4)



Objective	Inputs	Targets
Prefix language modeling BERT-style Deshuffling Li.d. noise, mask tokens Li.d. noise, replace spans Li.d. noise, drop tokens Random spans	Thank you for inviting Thank you <m> <m> me to yo party me for your to . las Thank you <m> <m> me to Thank you <x> me to you Thank you me to your pa Thank you <x> to <y> we</y></x></x></m></m></m></m>	me to your party last week . (original text) Corruption rate Span length approaches Language modeling Mask Mask Mask
		BERT-style Spans 25% 5 Deshuffling Drop 50% 10

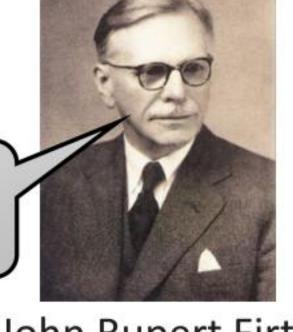




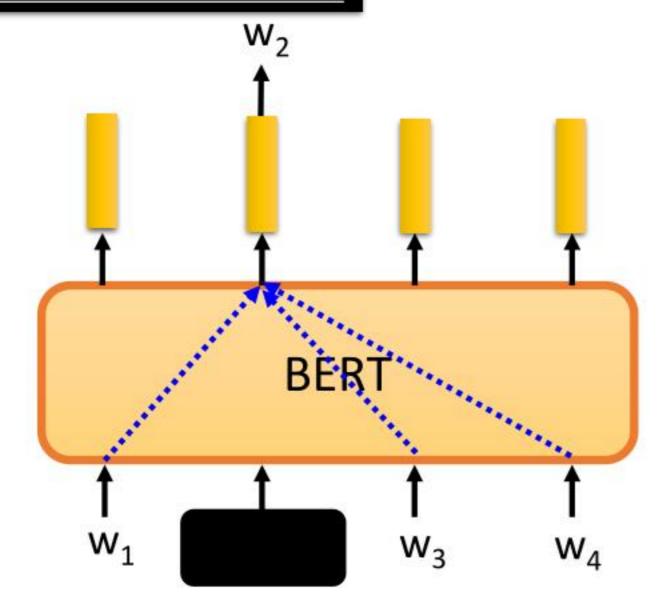


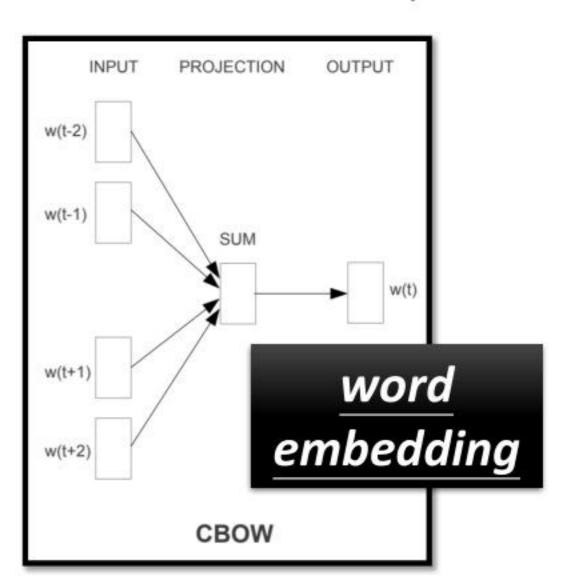
Contextualized word embedding

You shall know a word by the company it keeps

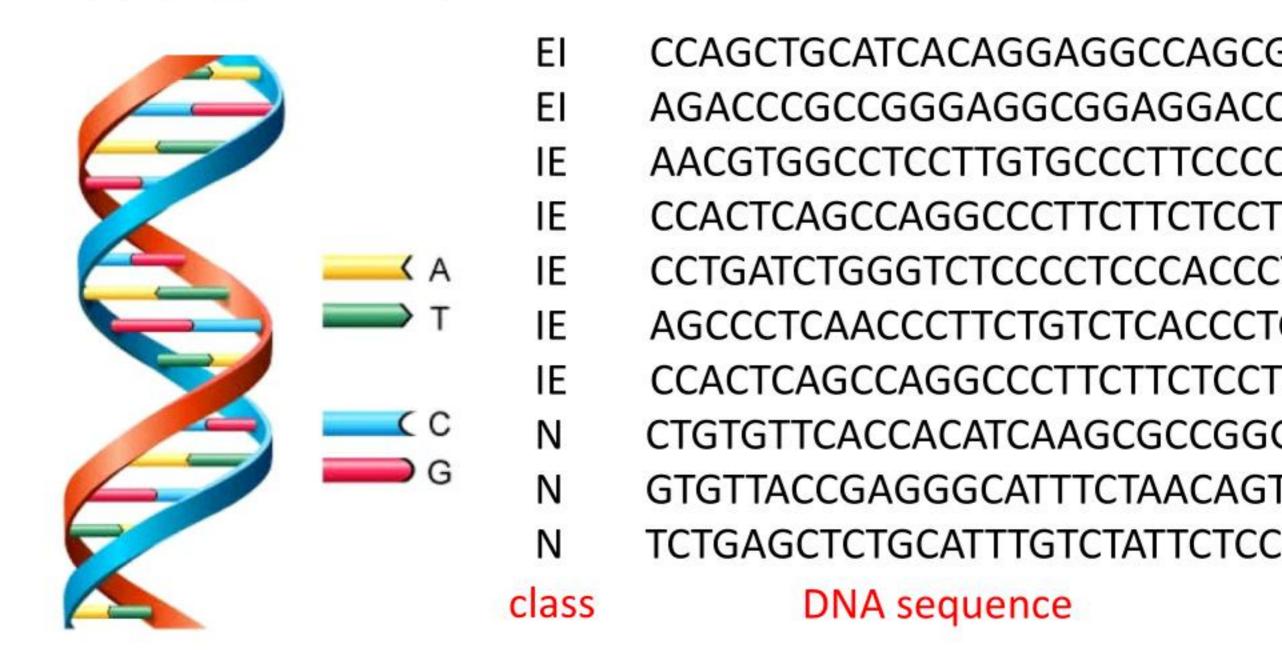


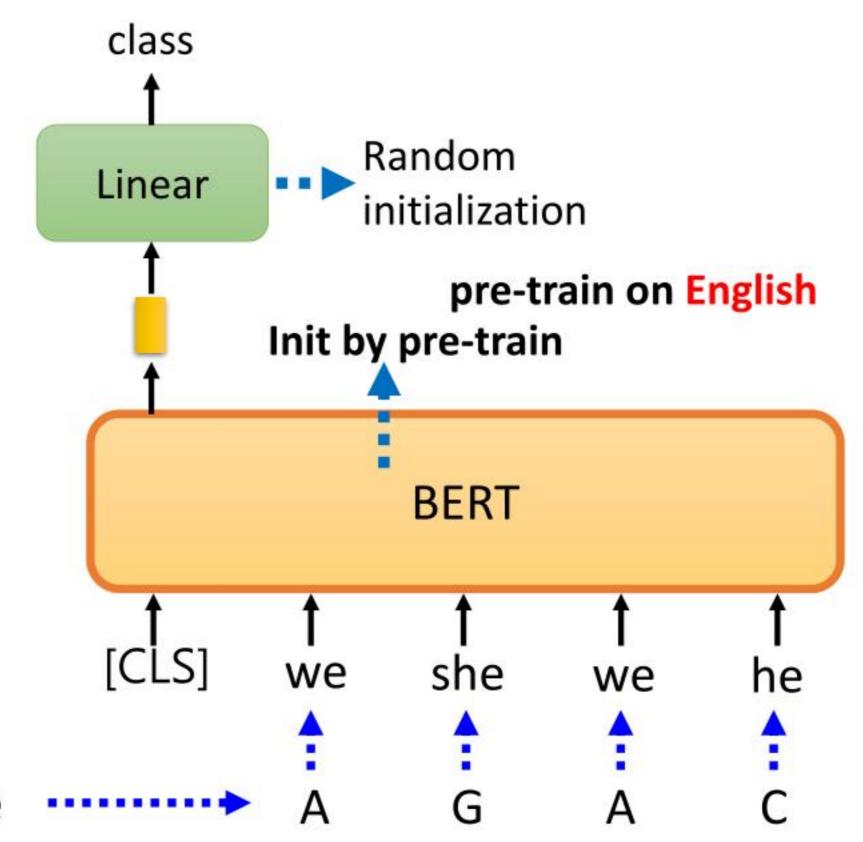
John Rupert Firth





Applying BERT to protein, DNA, music classification





Α	we		
Т	you		
С	he		
G	she		

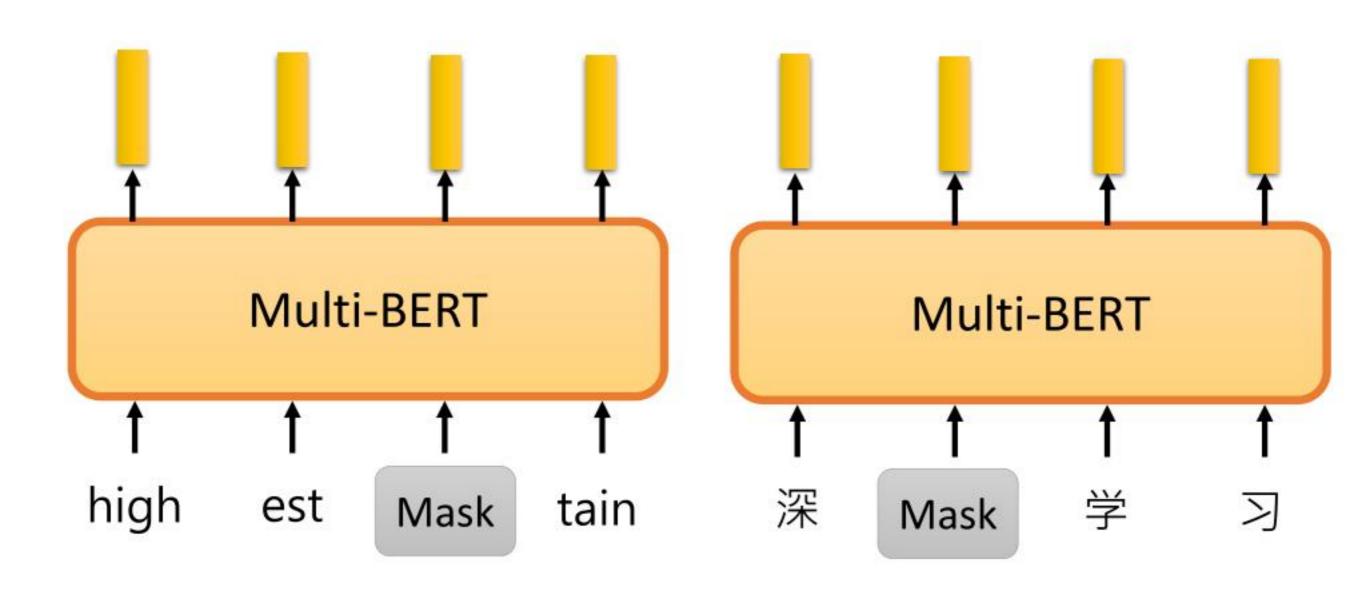
DNA sequence

Applying BERT to protein, DNA, music classification

	Protein			DNA				Music
	localization	stability	fluorescence	Н3	H4	H3K9ac	Splice	composer
specific	69.0	76.0	63.0	87.3	87.3	79.1	94.1	-
BERT	64.8	74.5	63.7	83.0	86.2	78.3	97.5	55.2
re-emb	63.3	75.4	37.3	78.5	83.7	76.3	95.6	55.2
rand	58.6	65.8	27.5	75.6	66.5	72.8	95	36



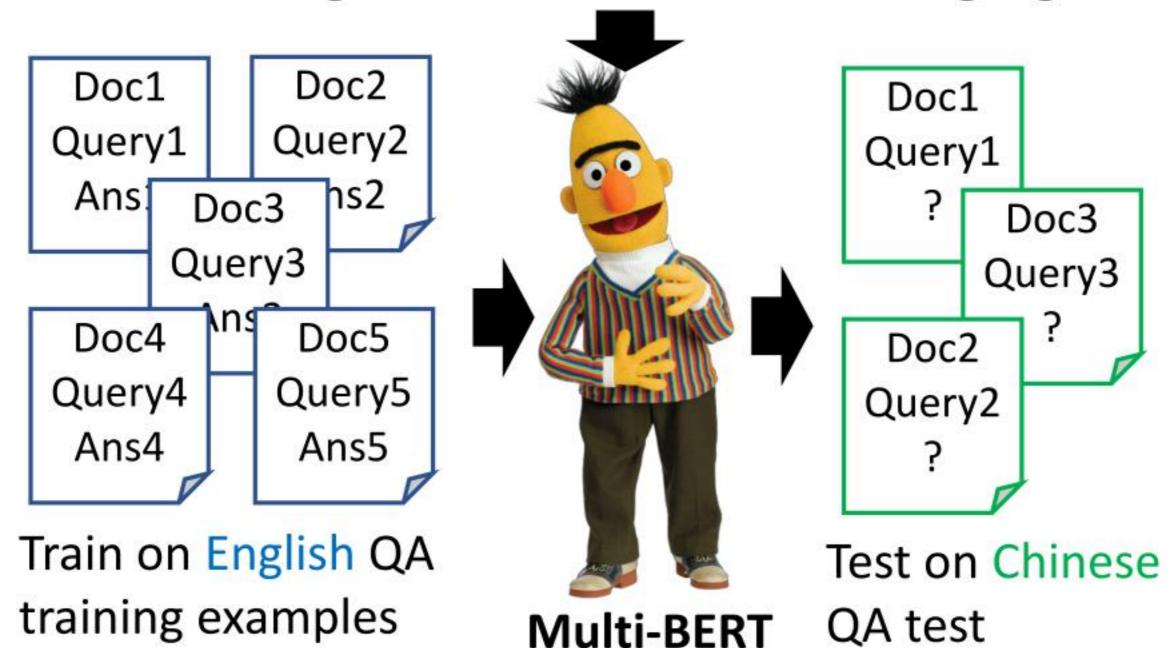
Multi-lingual BERT



Training a BERT model by many different languages.

Zero-shot Reading Comprehension

Training on the sentences of 104 languages



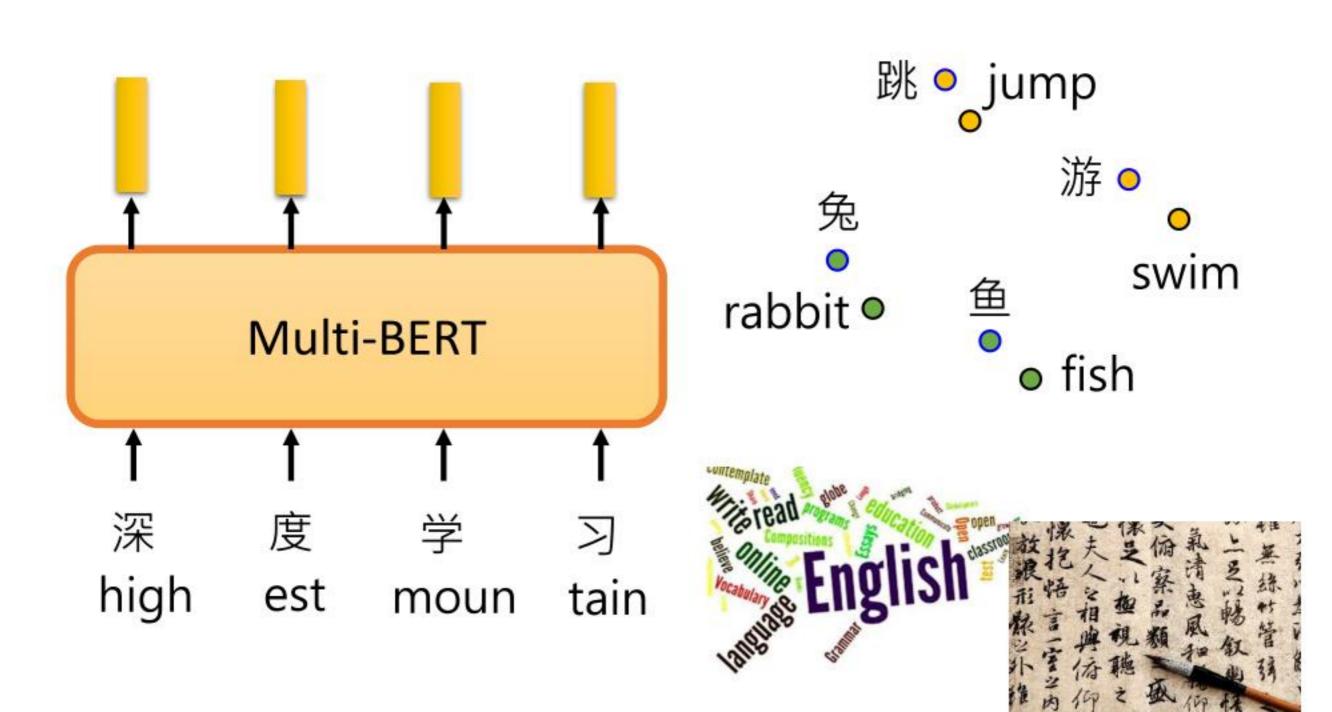
Zero-shot Reading Comprehension

English: SQuAD, Chinese: DRCD

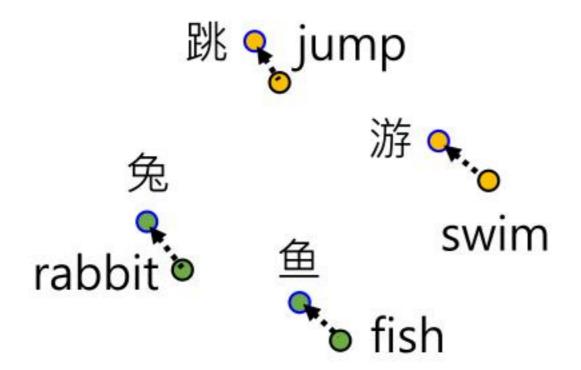
Model	Pre-train	Fine-tune	Test	EM	F1
QANet	none	Chinese		66.1	78.1
BERT	Chinese	Chinese		82.0	89.1
	104 languages	Chinese	Chinese	81.2	88.7
		English		63.3	78.8
		Chinese + English		82.6	90.1

F1 score of Human performance is 93.30%

Cross-lingual Alignment?



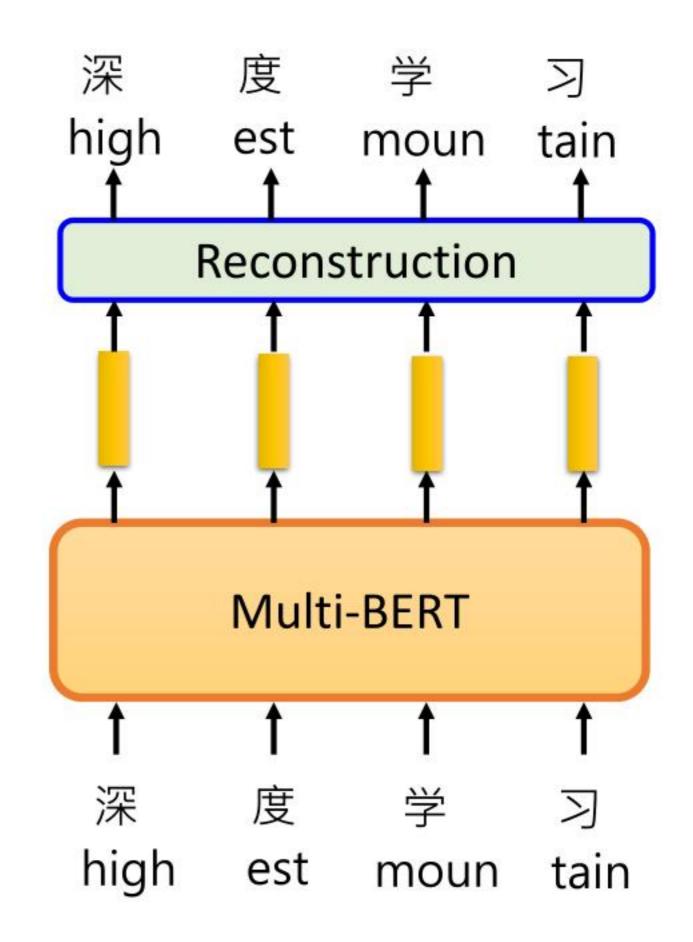
Weird???



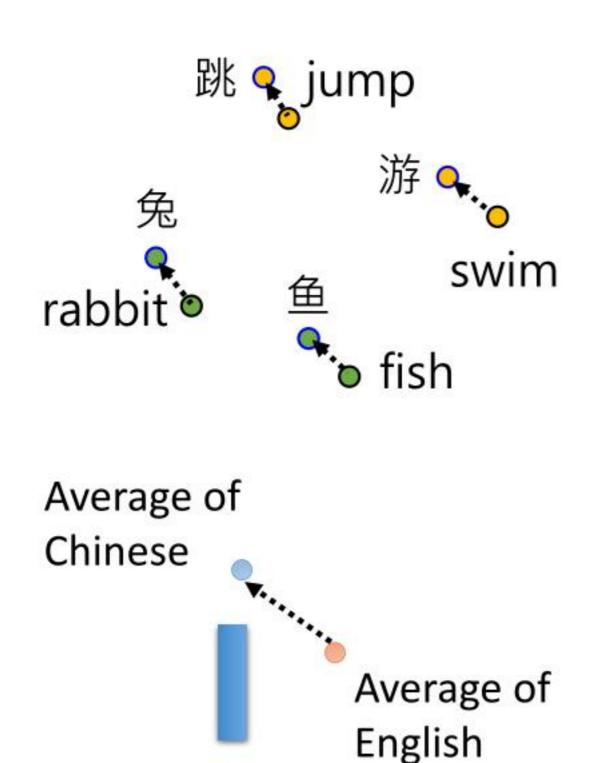
If the embedding is language independent ...

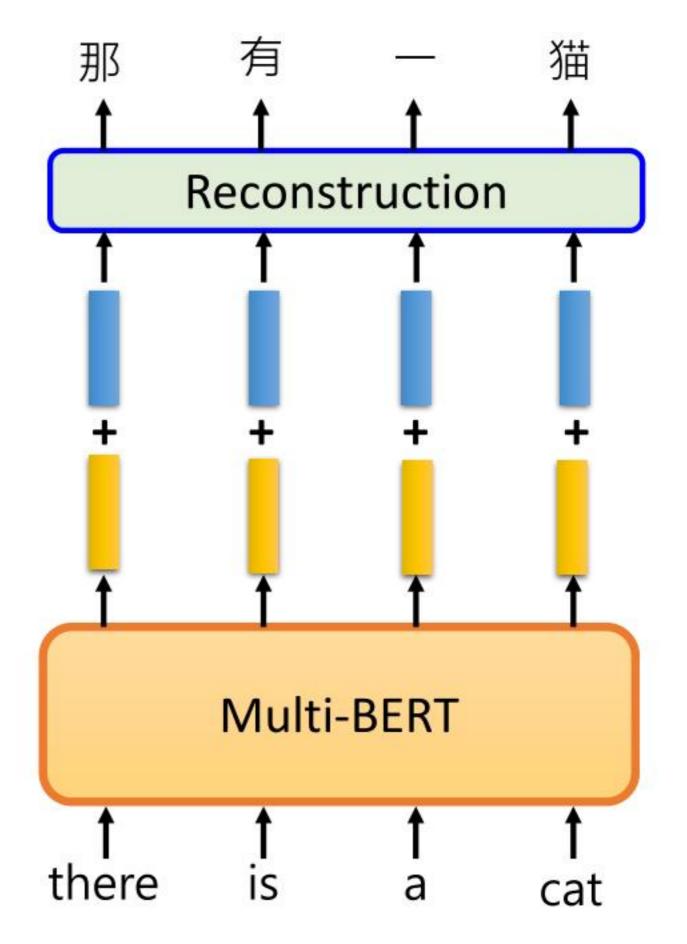
How to correctly reconstruct?

There must be language information.

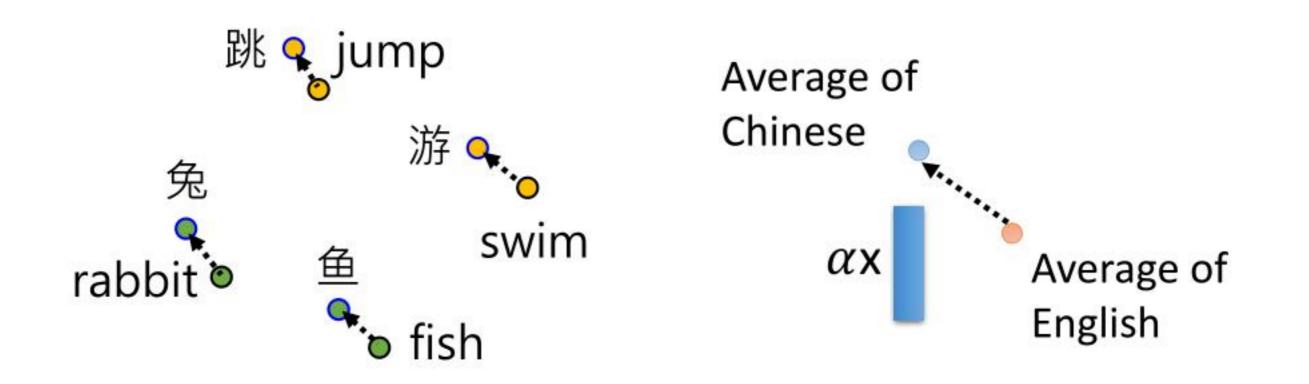


Where is Language?





If this is true ...



Input (en) The girl that can help me is all the way across town. There is no one who can help me.

Ground Truth (zh) 能帮助我的女孩在小镇的另一边。没有人能帮助我。。 en+zh, $\alpha=1$. 孩,can 来我是all the way across 市。。There 是无人人can help 我。 en+zh, $\alpha=2$. 孩的的家我是这个人的市。。他是他人人的到我。 en+zh, $\alpha=3$ 。,的的的他是的个的的,。:他是他人,的。他。

Unsupervised token-level translation ©