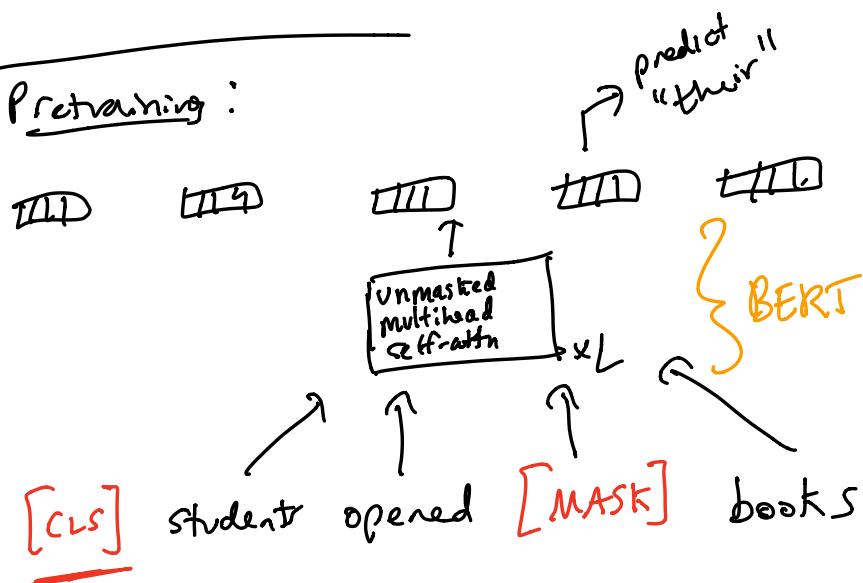


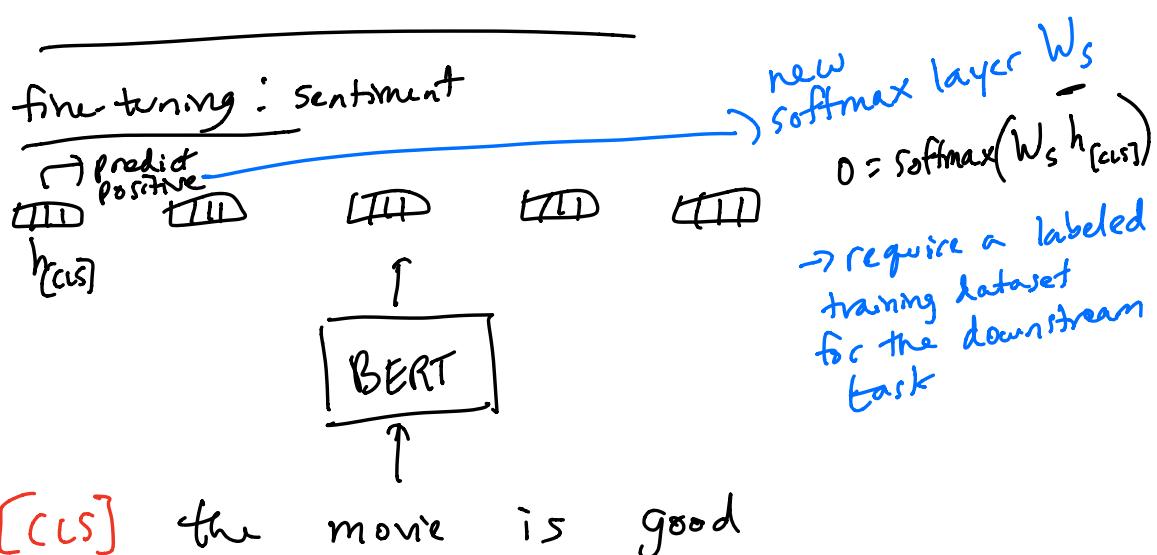
## BERT:

- ↳ example of the encoder-only paradigm
- ↳ Pretraining: train w/ self-supervised obj called "masked LM"
- ↳ Fine-tuning: process of adapting the pretrained model to a particular downstream task

### Pretraining:



### Fine-tuning: sentiment



## Extractive QA

↳ inputs: + a question  
+ a paragraph that contains the answer

↳ output: a span of the paragraph  
that answers the question

↳ datasets: SQuAD v1,2 ; QuAC /CoQA,  
HotpotQA, ...

Q: Who starred in the Matrix as Neo?

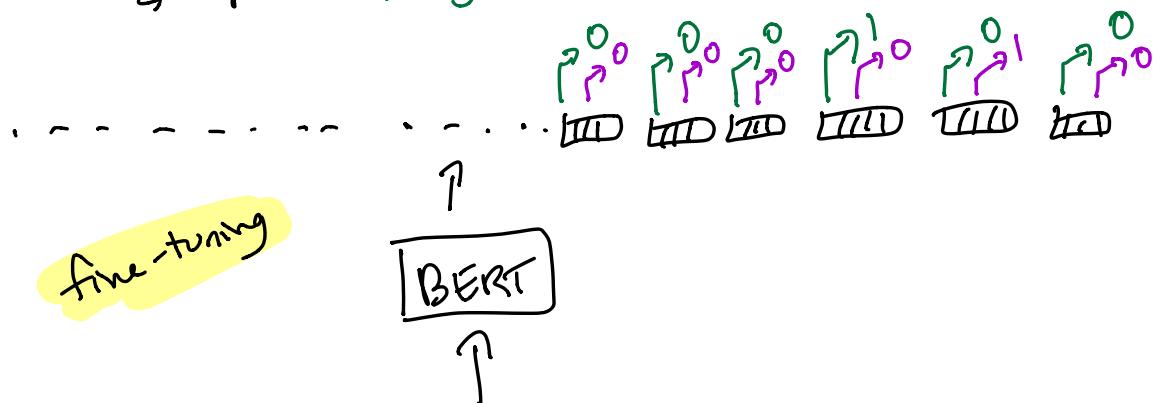
P:  $w_1, w_2, w_3, \dots$ . Neo was played by  
actor  $i$  Keanu Reeves  $j, \dots, \dots$ .

$$A := (i, j)$$

How do we use BERT for extractive QA?

↳ softmax layer on each token in passage

→ softmax layers or linear layers  
↳ predict beginning, end index of answer spans



[CLS] Who starred in the Matrix [SEP] w<sub>1</sub> w<sub>2</sub> w<sub>3</sub> ... Keanu Reeves ...

how do we select an answer span at test time?

→ find the span  $w_{i \dots j}$  that maximizes

$$P_{\text{START}}(i) \cdot P_{\text{END}}(j)$$

↳ exclude spans where  $j < i$

↳ exclude spans longer than a threshold

advanced variants of BERT:

↳ pretraining improvements  $\Rightarrow$  RoBERTa  
↳ more data

↳ longer sequences during pretraining

- BERT  $\Rightarrow$  512 tokens max

- XLNet  $\Rightarrow$  900 tokens

↳ more pretraining objectives

↳ ELECTRA

↳ "students opened [MASK] books"

↳ "students opened monster books"

real? corrupted? real corrupted real

↳ smaller models

↳ tinyBERT, distilBERT, ALBERT

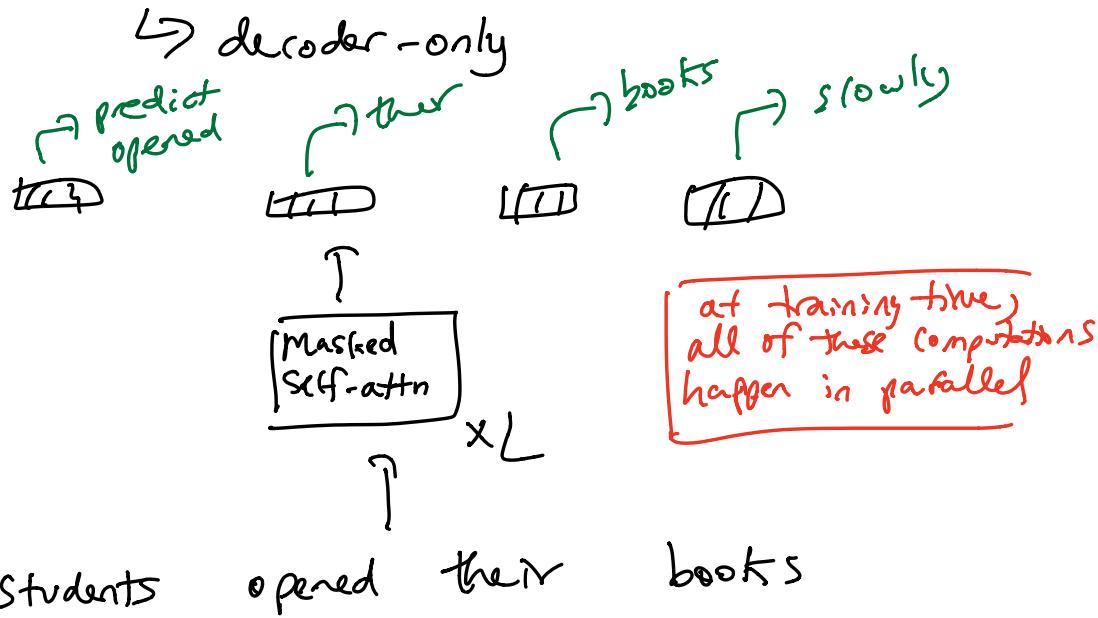
↳ distillation

↳ pruning

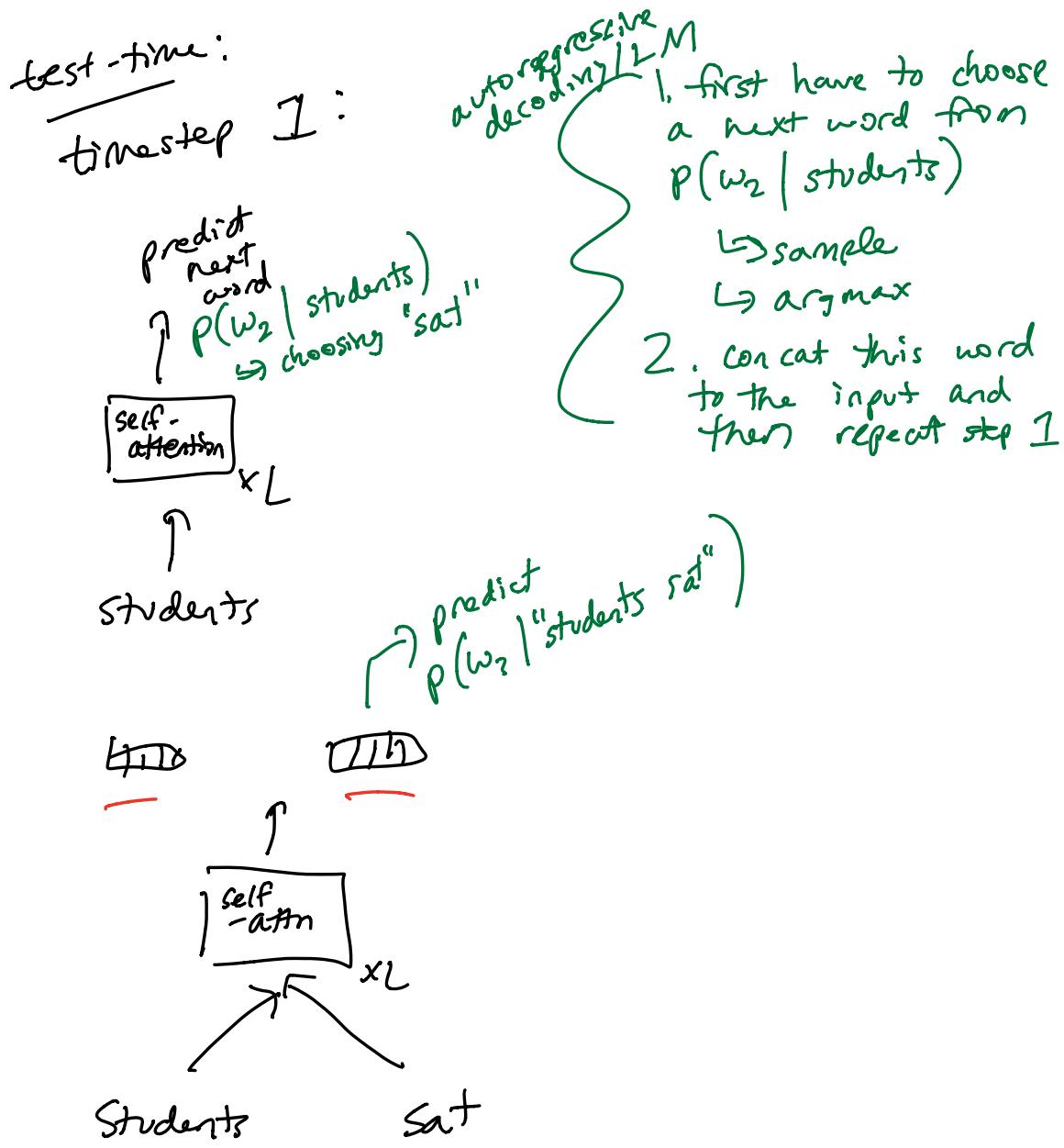
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Transformer Ms at test-time:

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⇒ these four predictions can happen simultaneously b/c we already know the identity of the gold next tokens  
ground-truth



at test-time, I have to decode the output word-by-word, because I don't have access to the gold next tokens

↳ decoding algorithms