

Encoder-based language models: BERT

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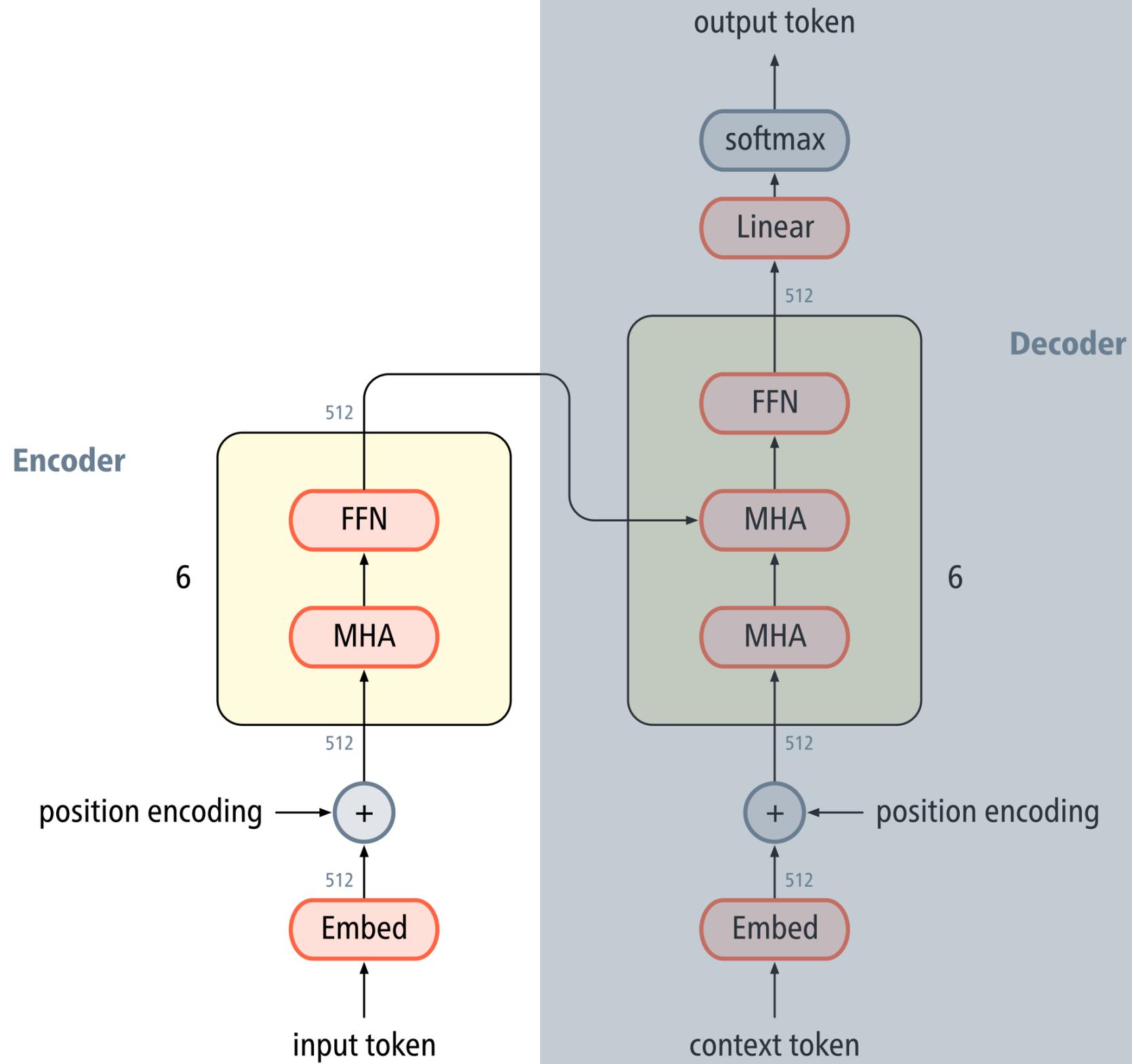
Department of Computer and Information Science

BERT

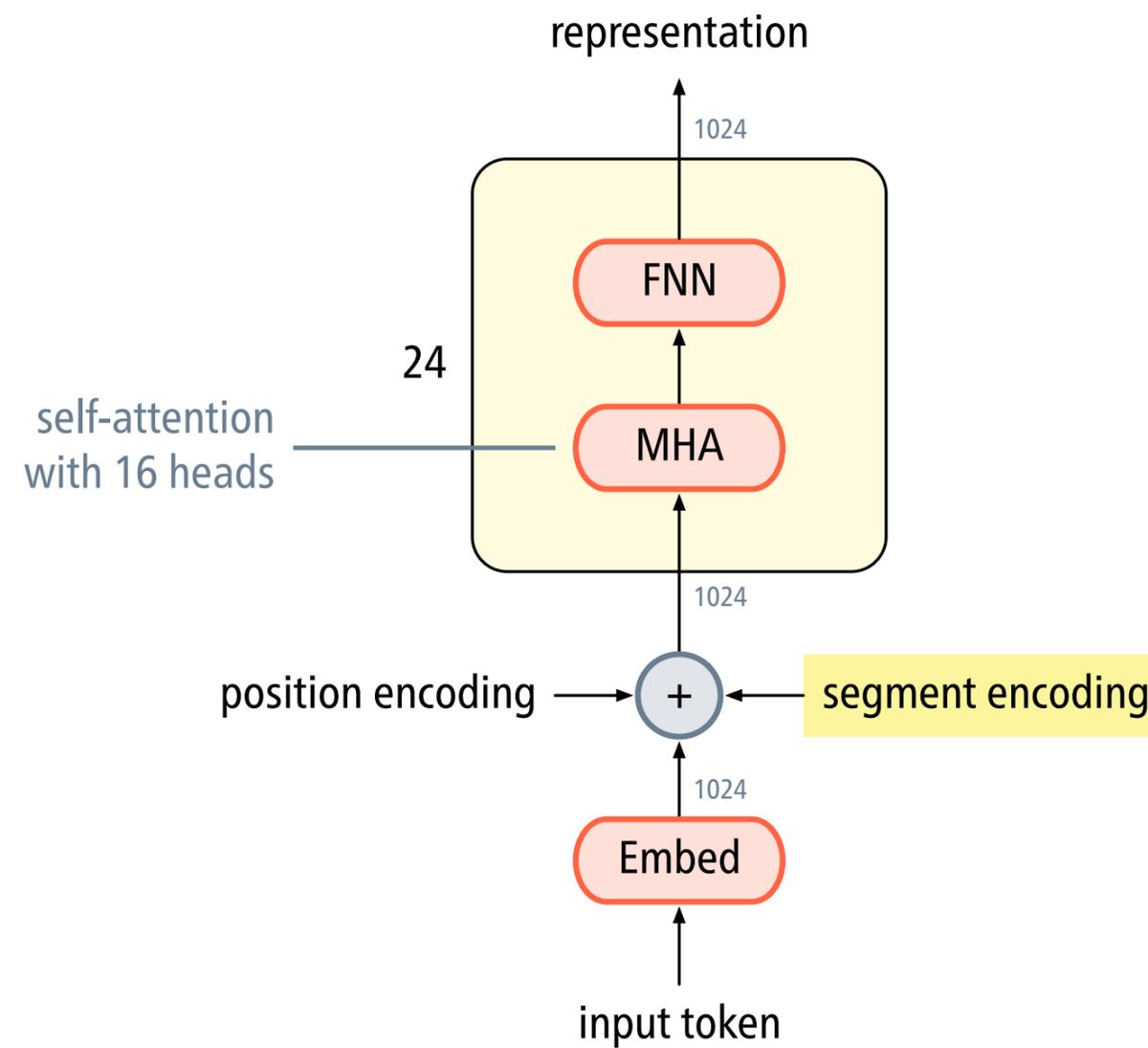
- The acronym **BERT** stands for “Bidirectional Encoder Representations from Transformers”.
- As an encoder, BERT can learn token representations that are conditioned on the complete input sequence.

non-directional





BERT (large model)



Model statistics

| | base | large |
|---------------------------|--------------|--------------|
| number of dimensions | 768 | 1024 |
| number of encoder blocks | 12 | 24 |
| number of attention heads | 12 | 16 |
| number of parameters | 110 M | 340 M |

[Devlin et al. \(2019\)](#)

Pre-training tasks

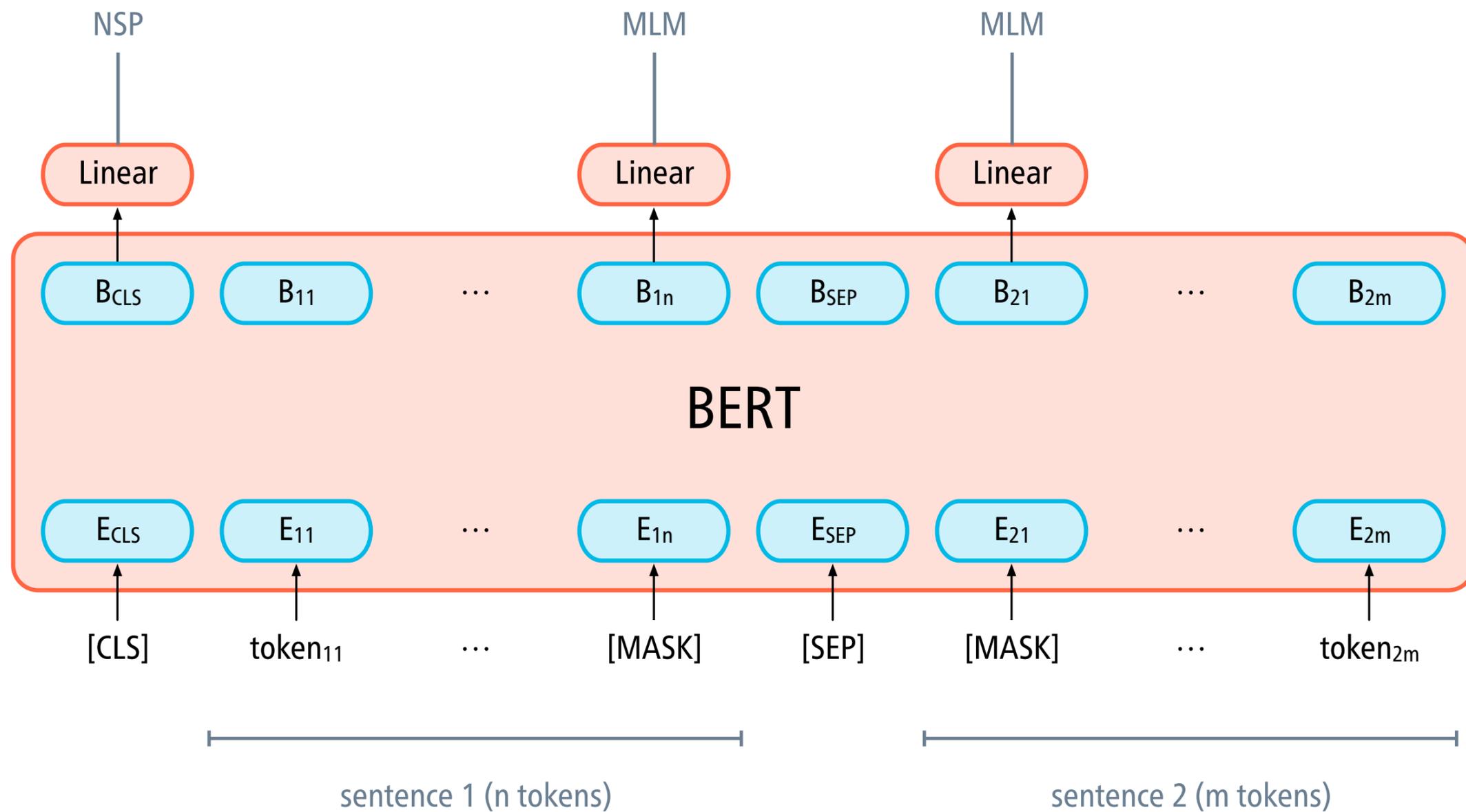
- **Masked Language Modelling (MLM)**

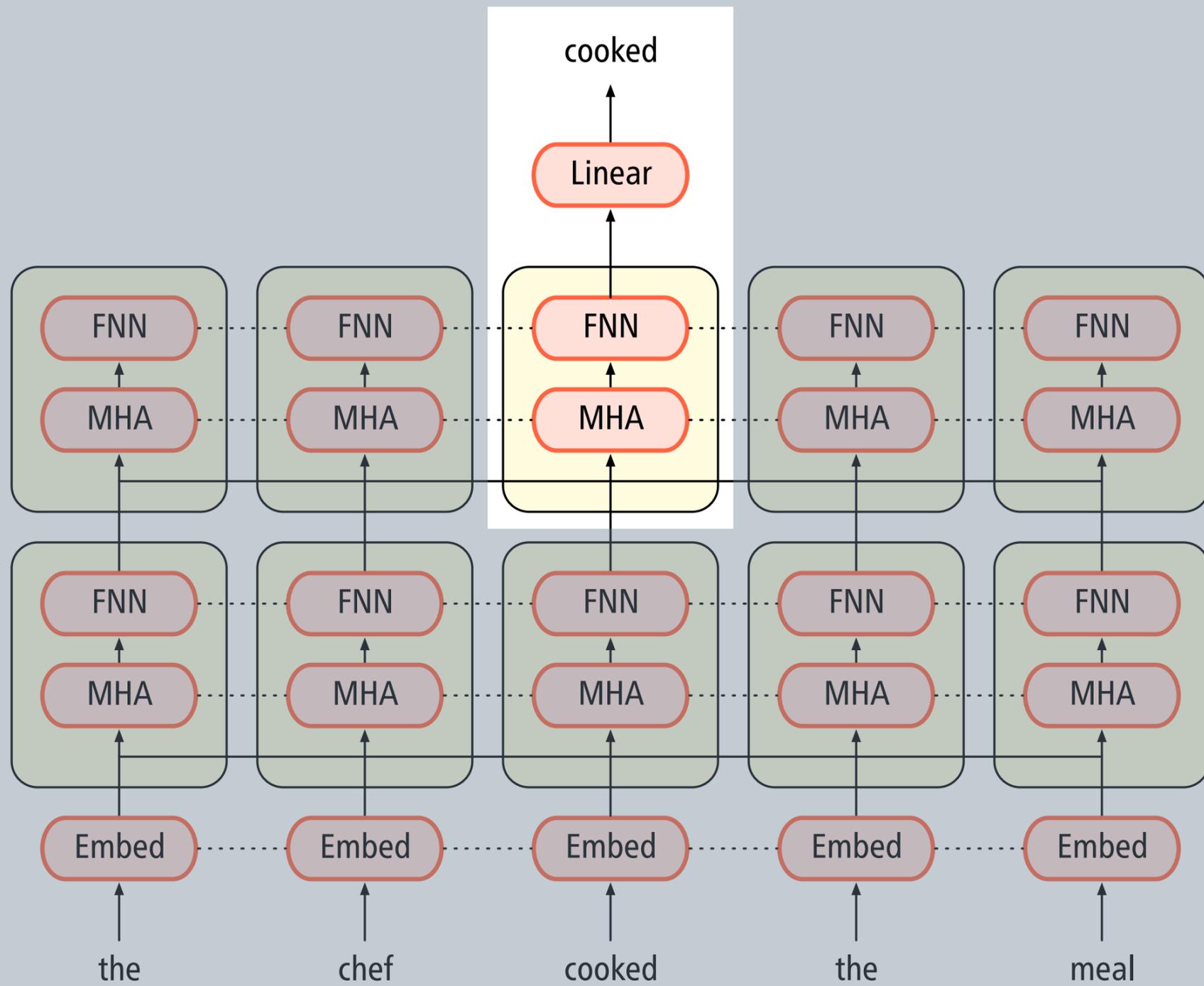
Tokens are masked out at random. The model is trained to predict the masked-out tokens.

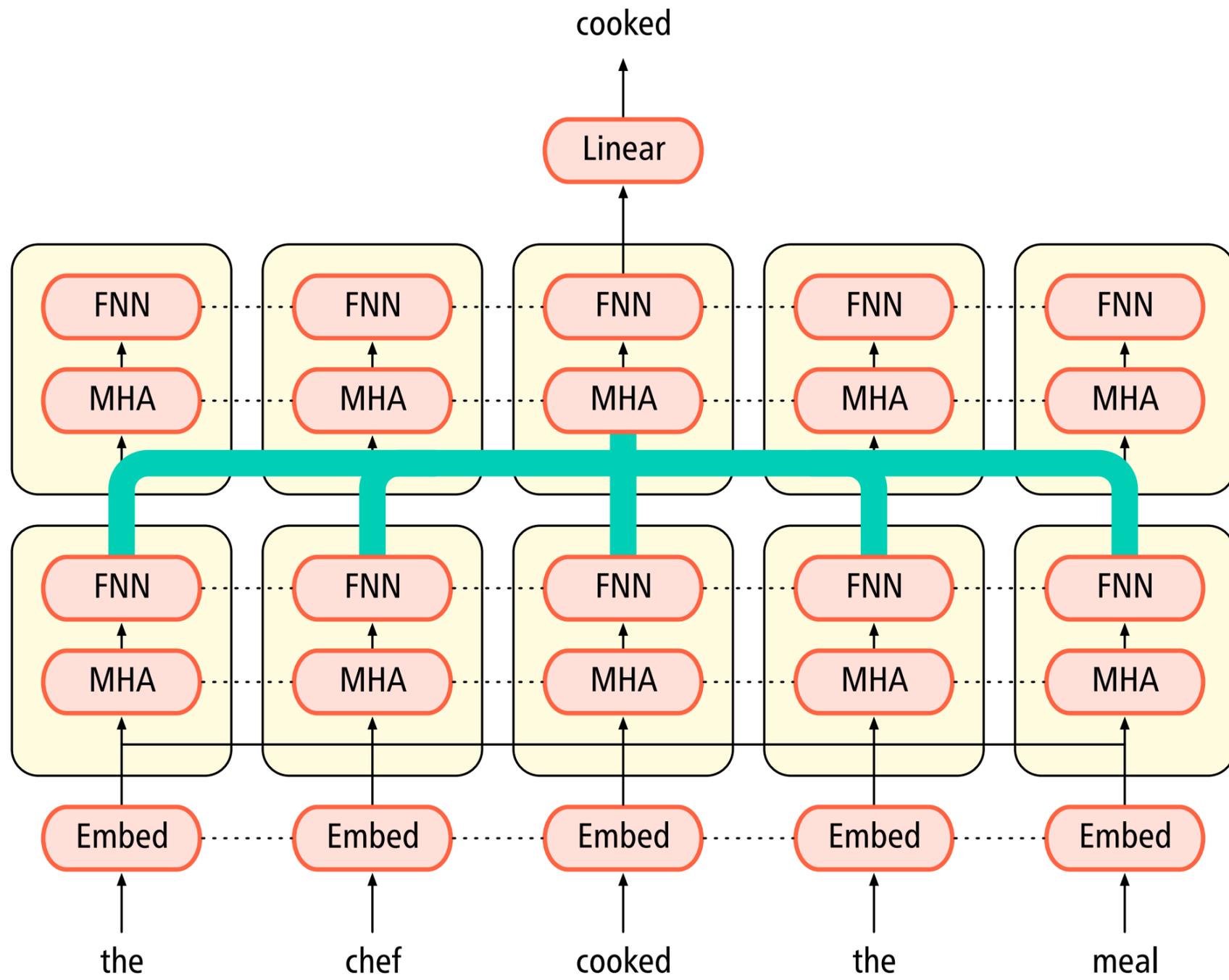
- **Next Sentence Prediction (NSP)**

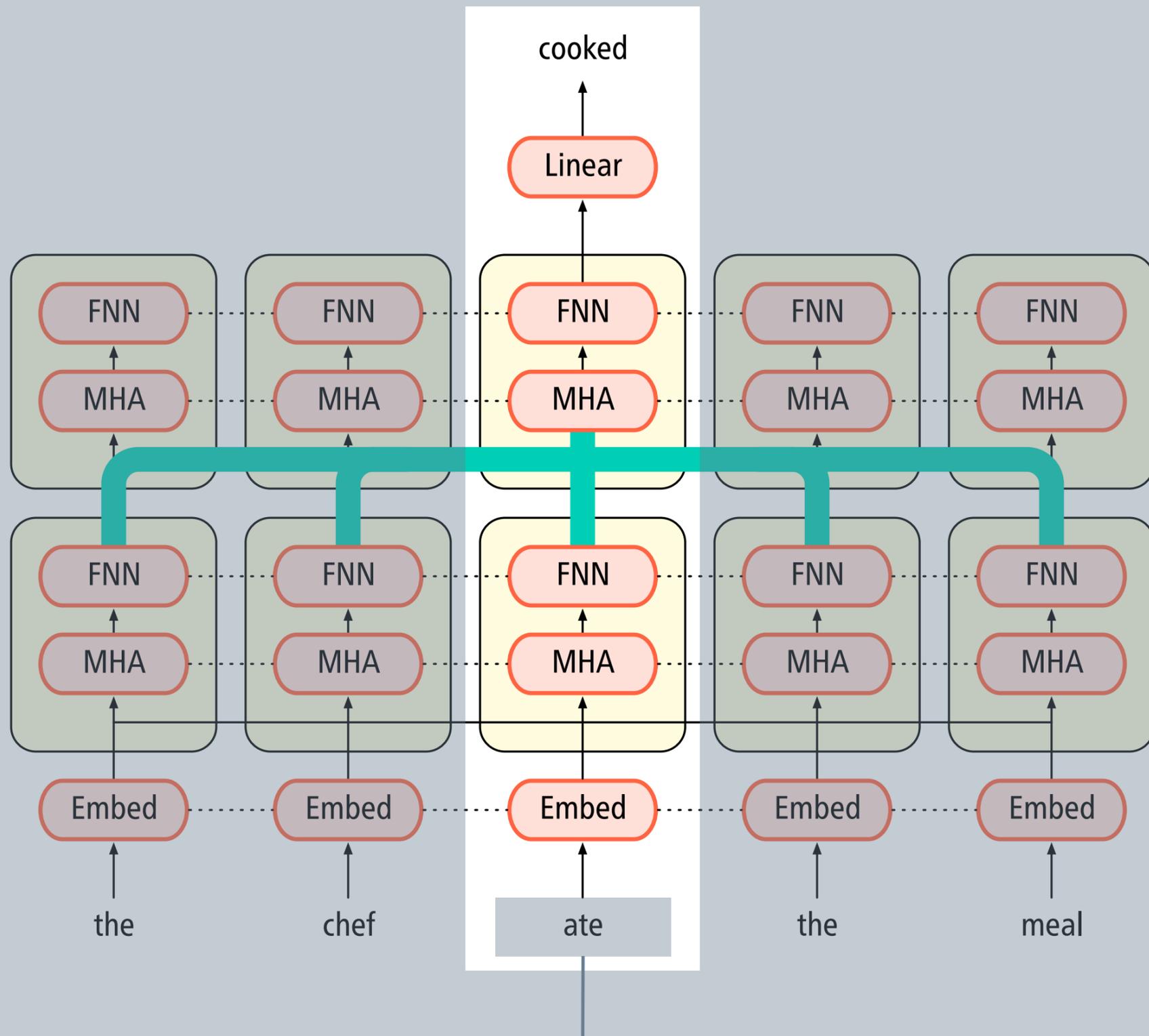
The model is trained to predict whether two randomly sampled sentences are adjacent in the training data.

Pre-training with MLM and NSP

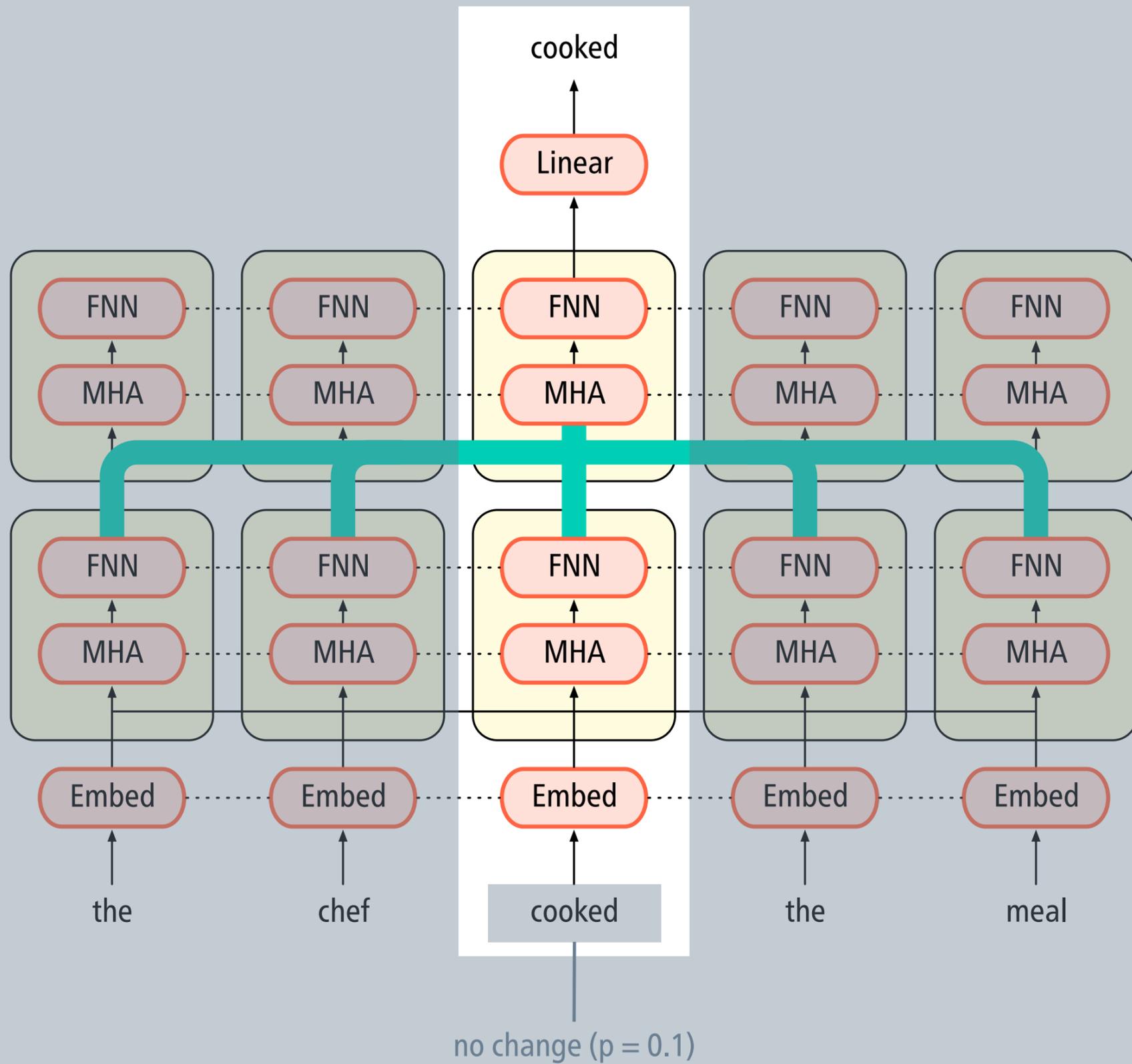




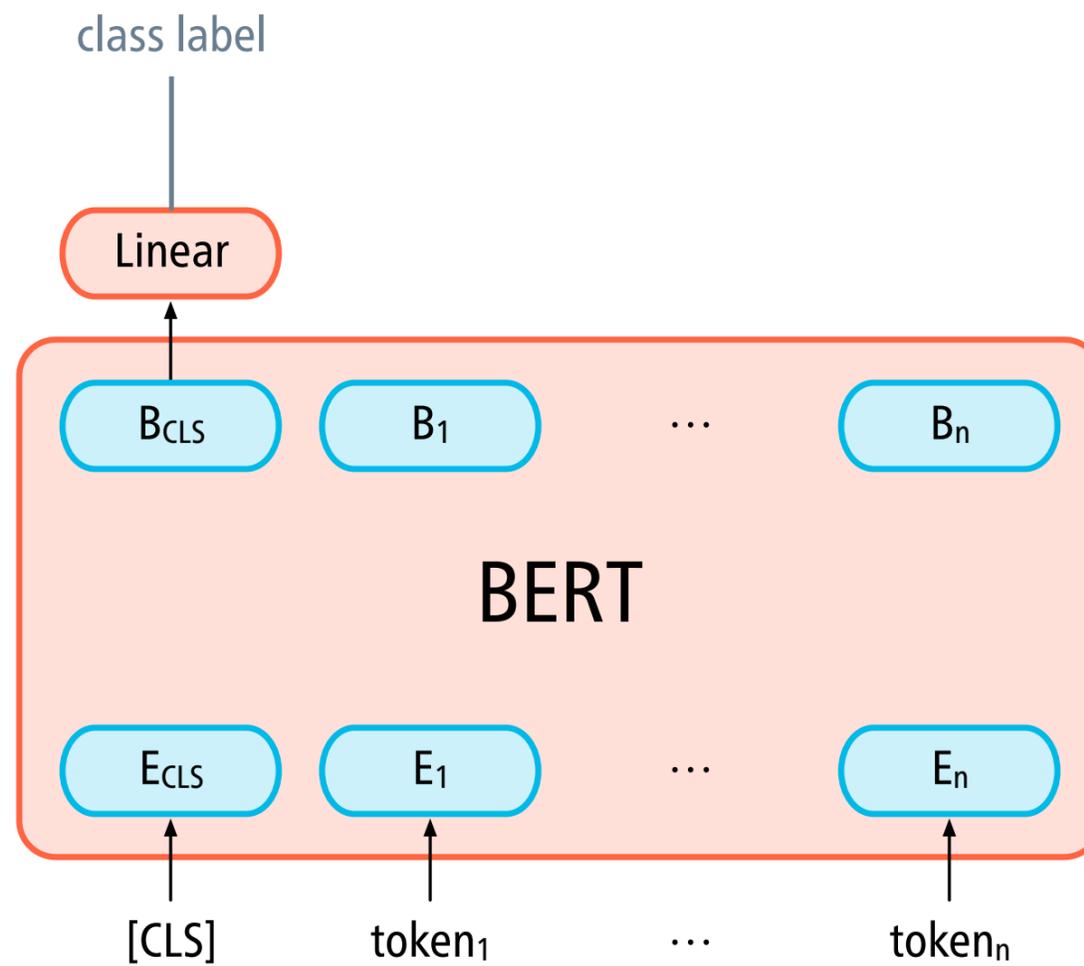




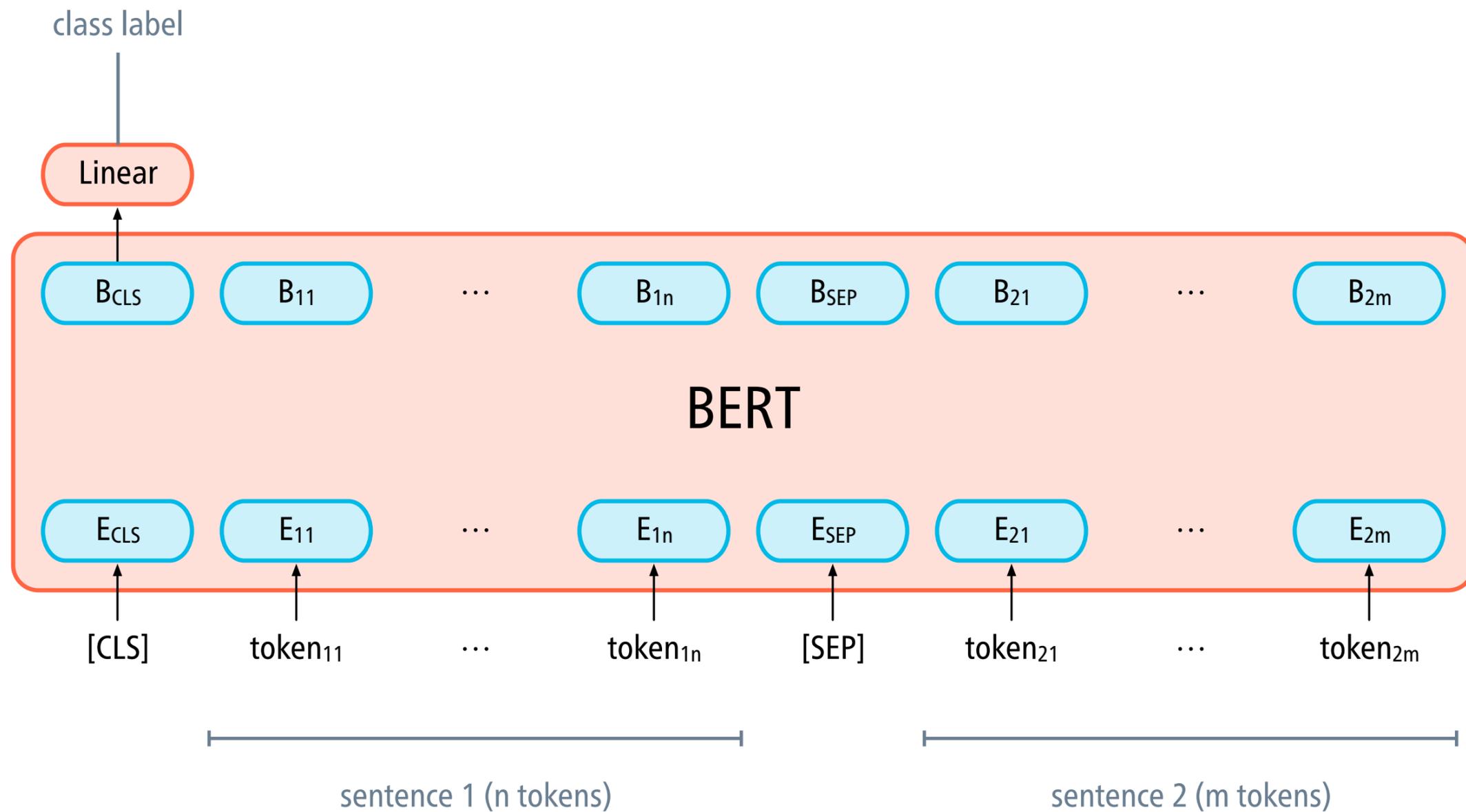
replaced with random word (p = 0.1)



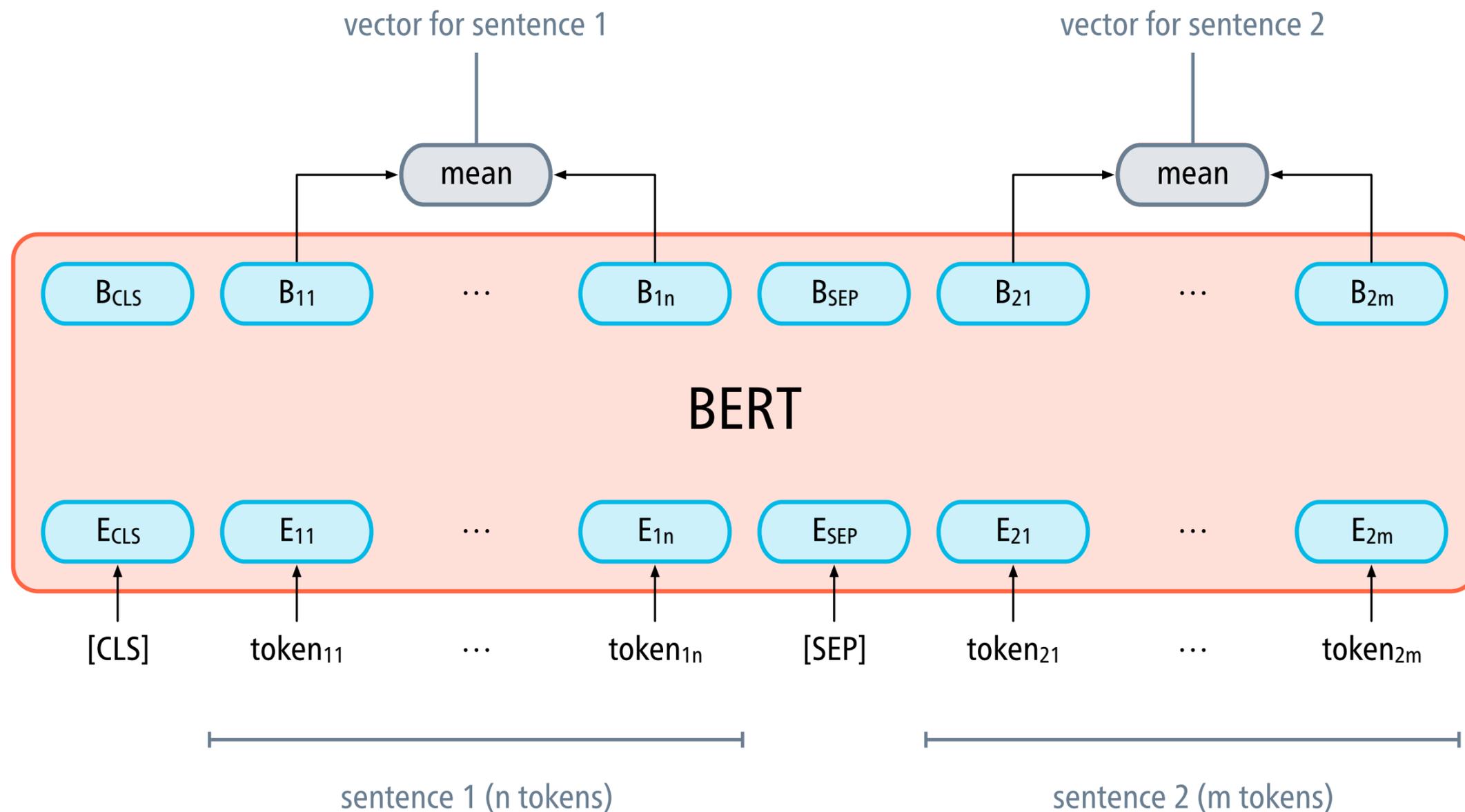
Fine-tuning on a single-sentence classification task



Fine-tuning on a sentence-pair classification task



Fine-tuning on a sentence-pair similarity task



Performance on the GLUE benchmark

| | GLUE |
|---------------------------|-------------|
| ELMo + Attention | 71.0 |
| Previous state-of-the-art | 74.0 |
| BERT (base) | 79.6 |
| BERT (large) | 82.1 |

GLUE test results, scored by the evaluation server | [Devlin et al. \(2019\)](#)

BERT-like models

- RoBERTa uses an improved recipe for pre-training and a significantly larger data set.

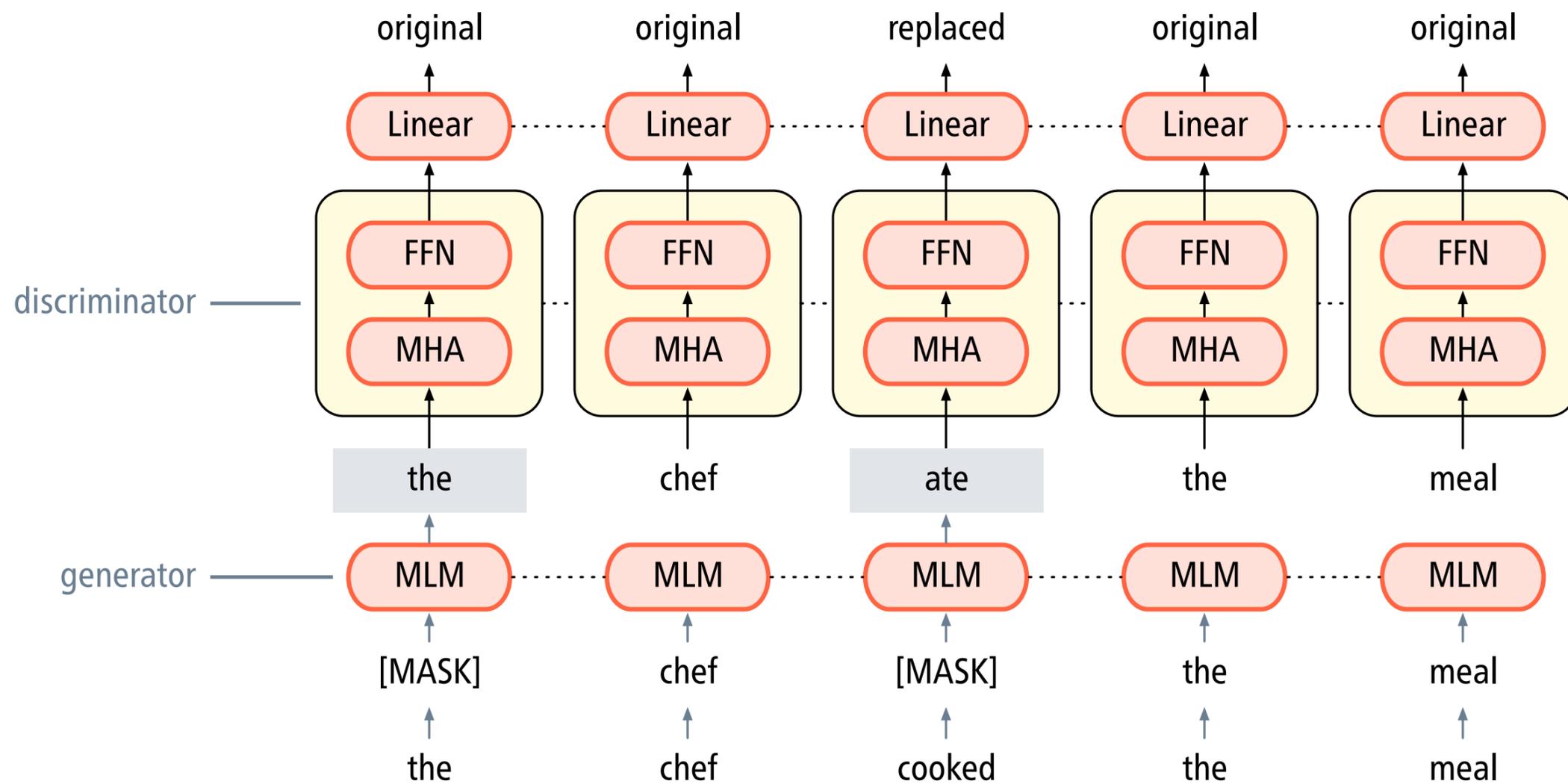
[Liu et al. \(2019\)](#)

- ALBERT and DistilBERT are models with reduced training time and model size, respectively.

[Lan et al. \(2019\)](#), [Sanh et al. \(2019\)](#)

- Many pre-trained BERT-like and other transformer models are available via [Hugging Face](#).

ELECTRA: Pre-training via replaced token detection



Effectiveness of replaced token detection

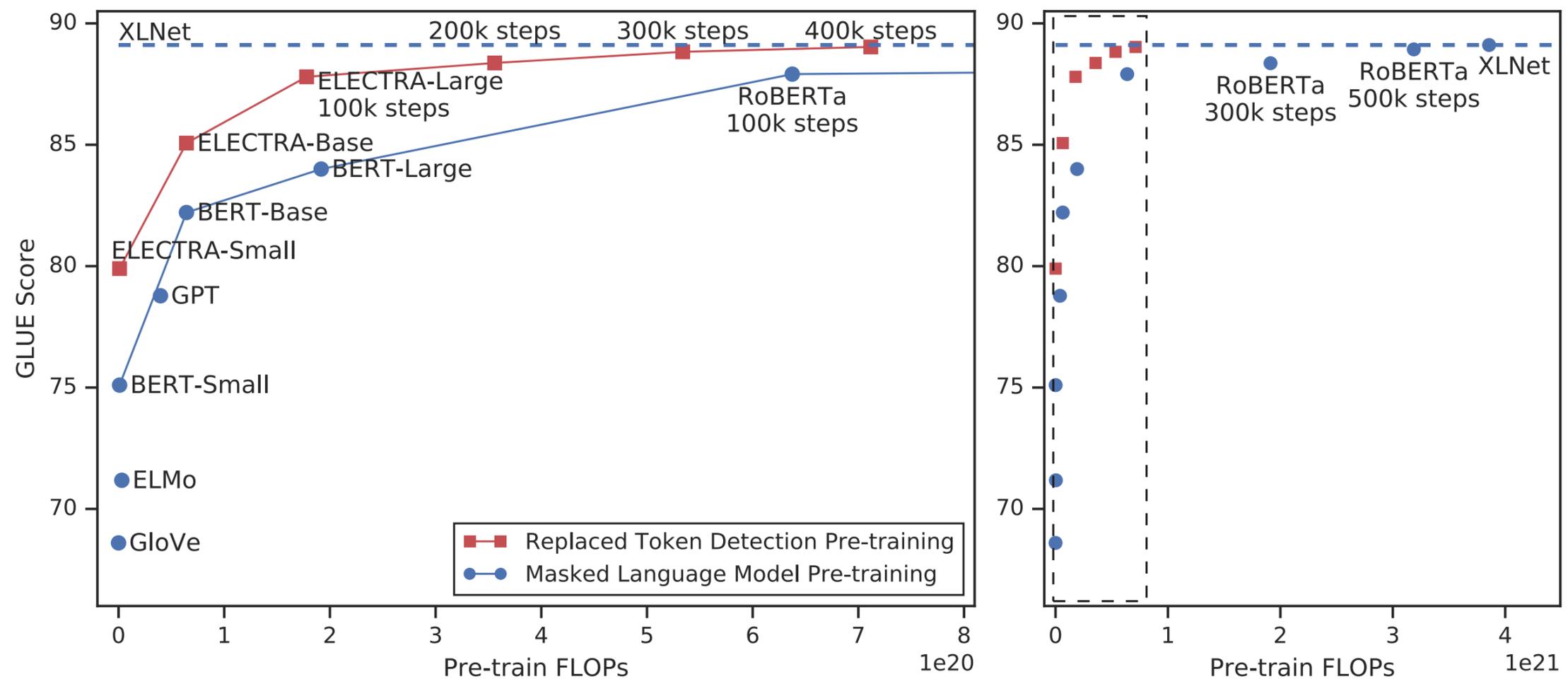


Figure 1: Replaced token detection pre-training consistently outperforms masked language model pre-training given the same compute budget. The left figure is a zoomed-in view of the dashed box.