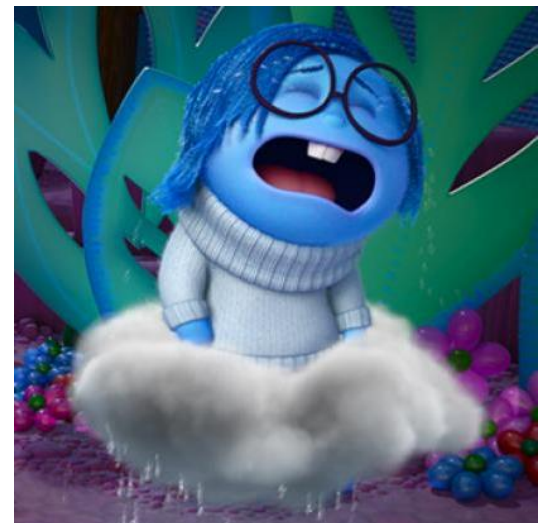


# BERT

Bidirectional Encoder Representations from Transformers



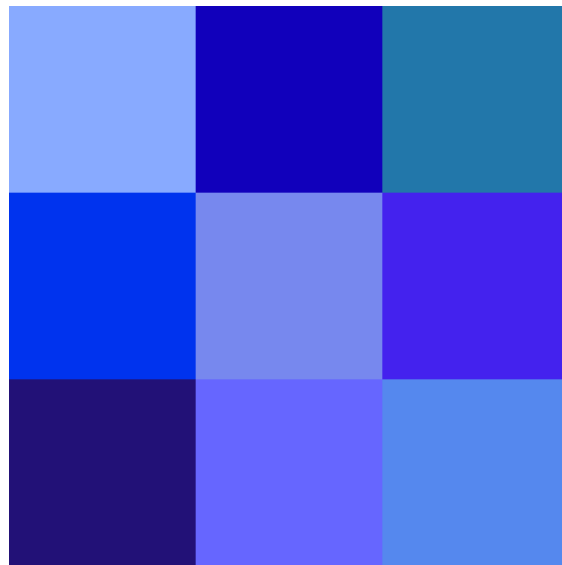
# Pre-Training :



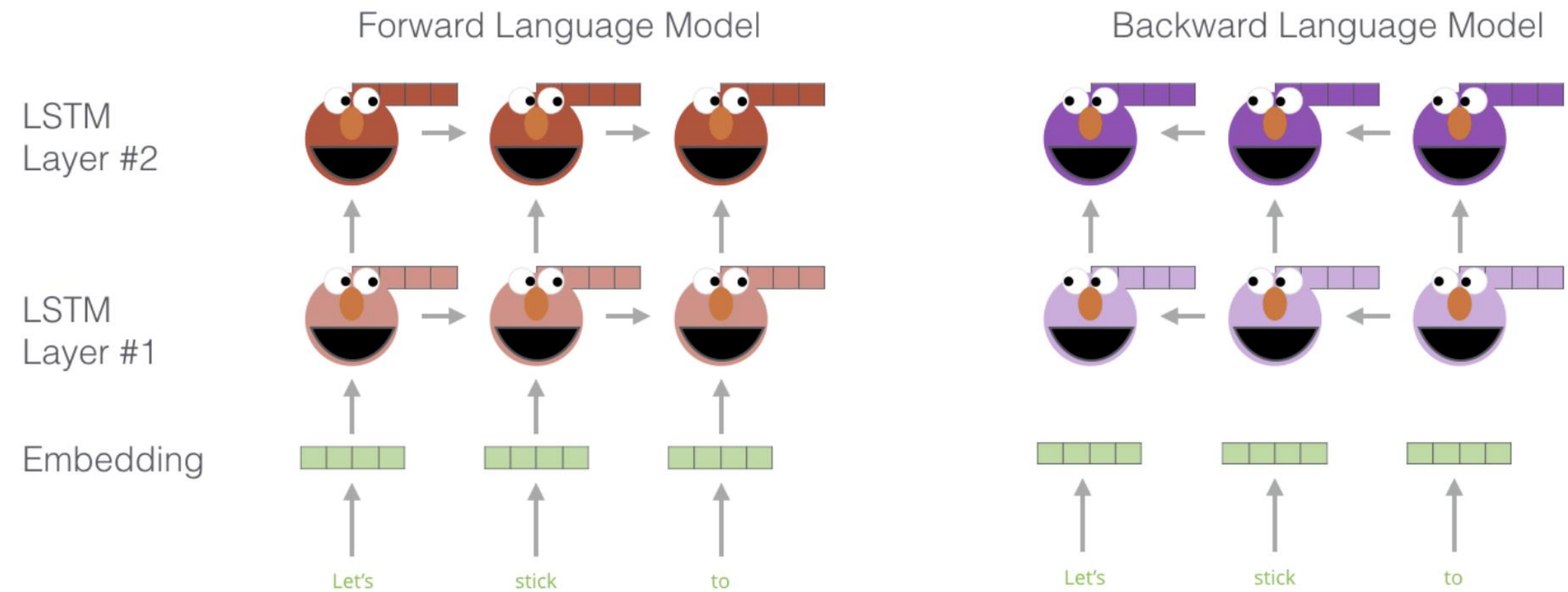
Blue



[0.2 0.8 -1.2]

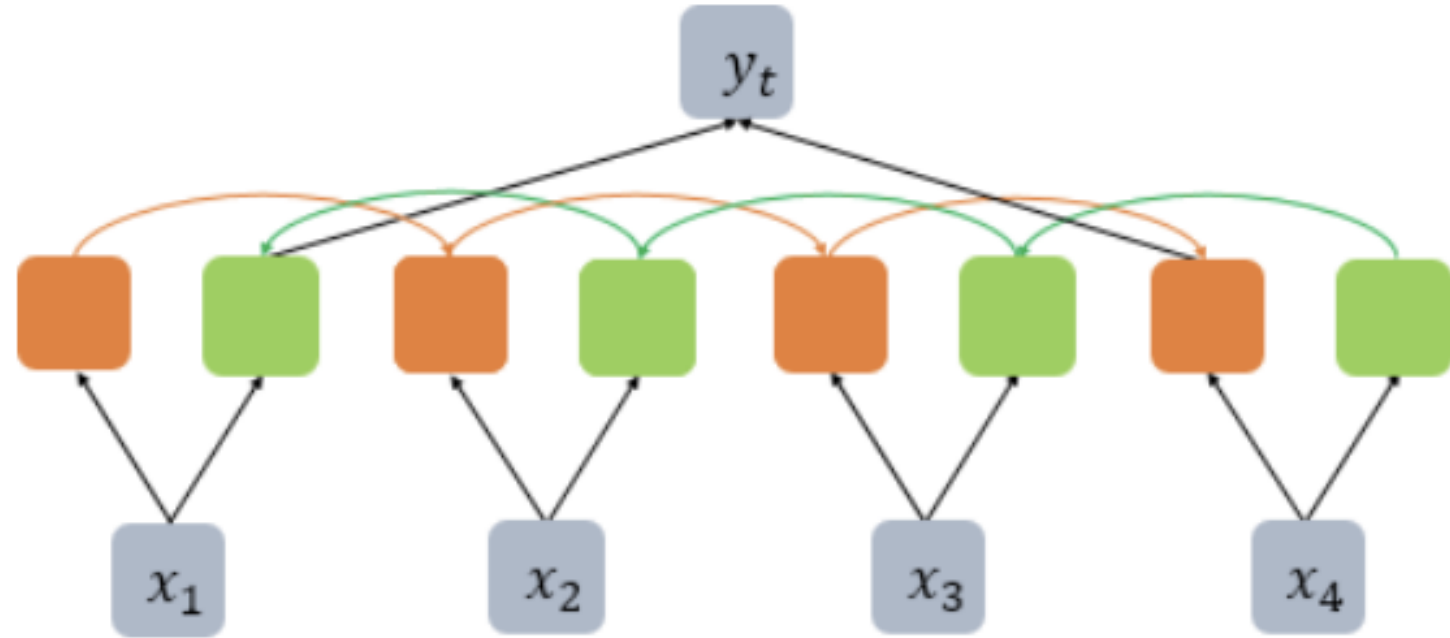


ELMo

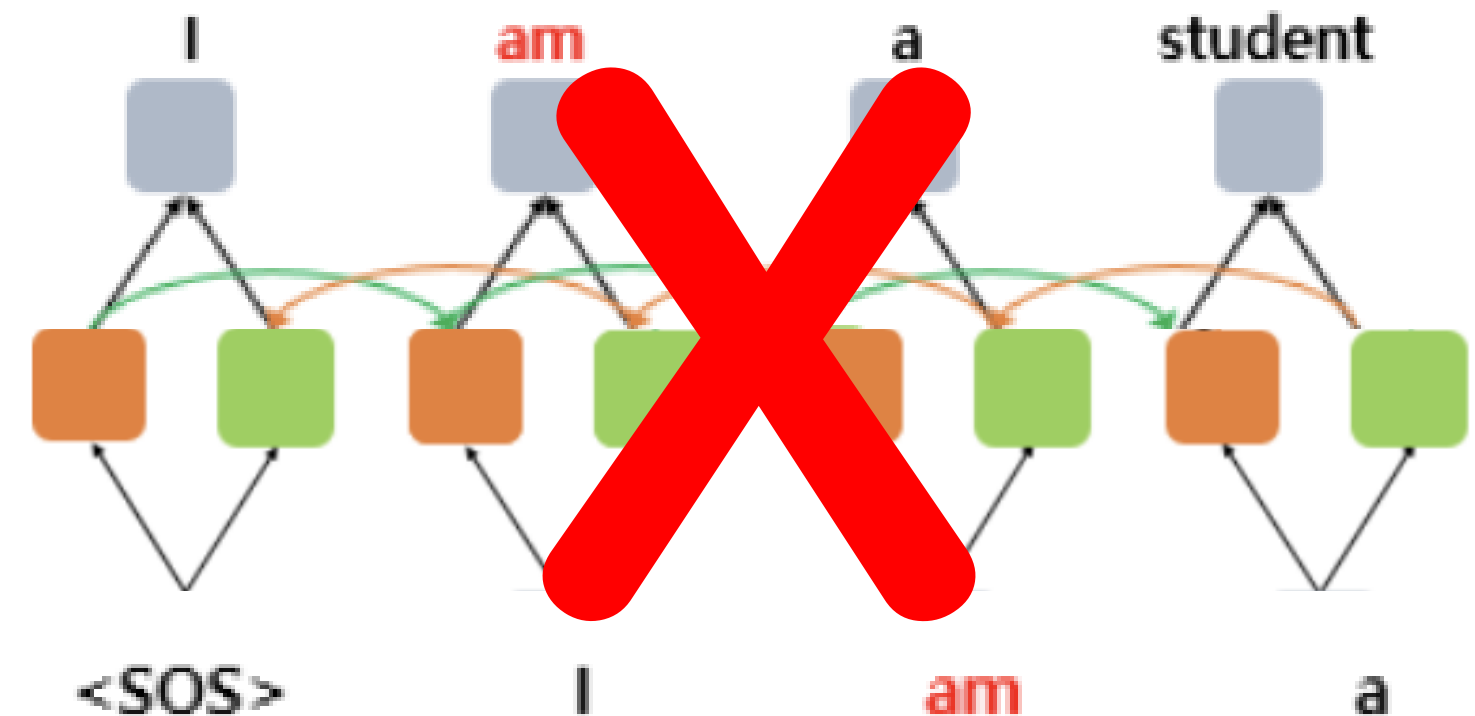


# Pre-Training :

---



양방향RNN



양방향RNNLM

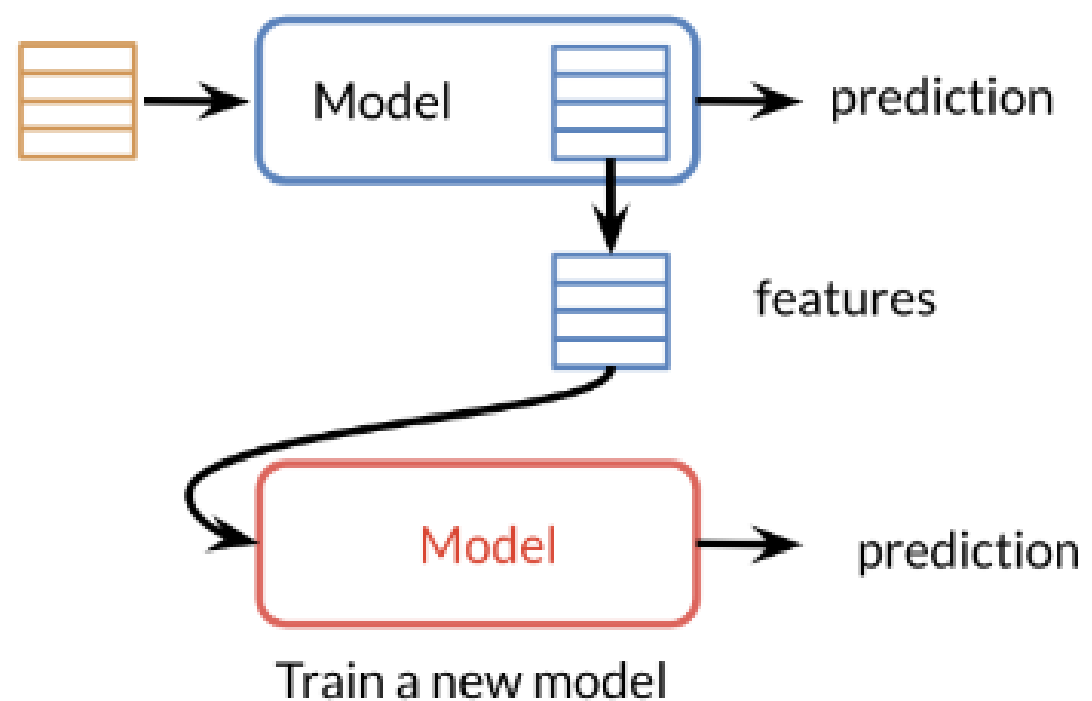
양방향성을 유지하면서 LM로서 역할을 할 수 있는 모델

: For Contextual Consideration

# Pre-Training :

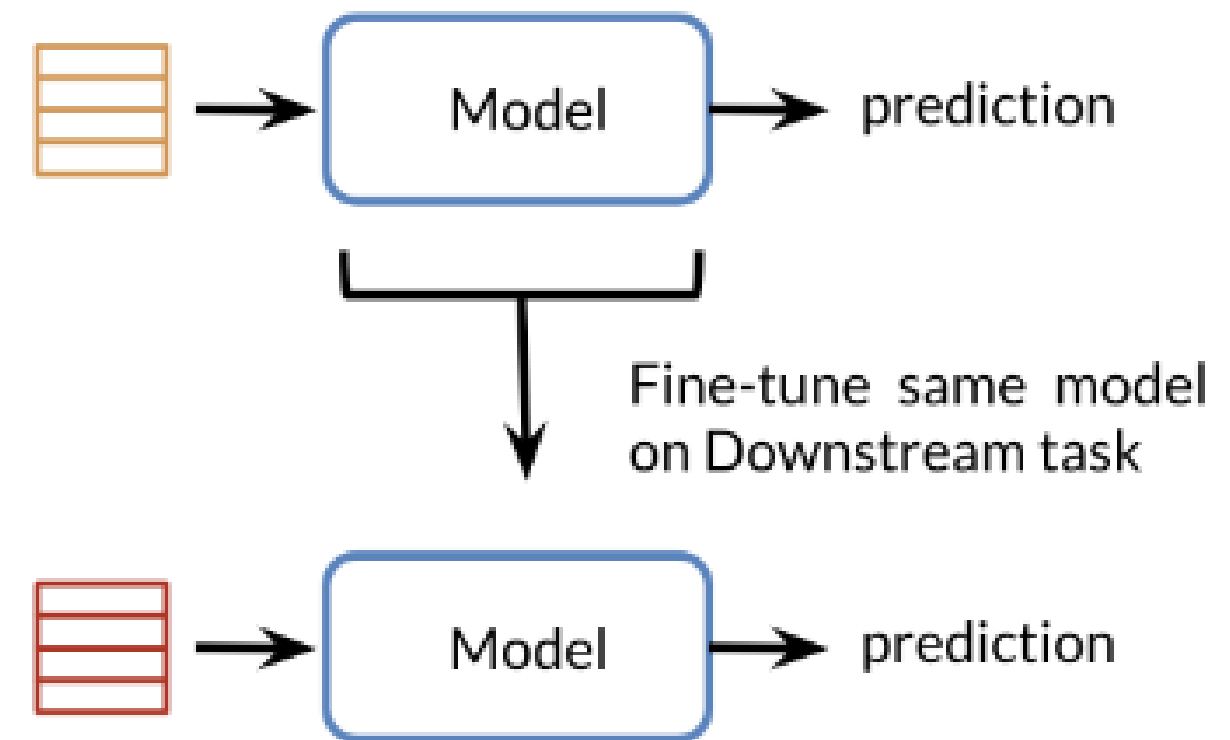
---

## Feature Based



train word embeddings and then using those features (i.e. word vectors) on a different task.

## Fine Tuning



"update" the meaning of words based on your specific domain.

# BERT

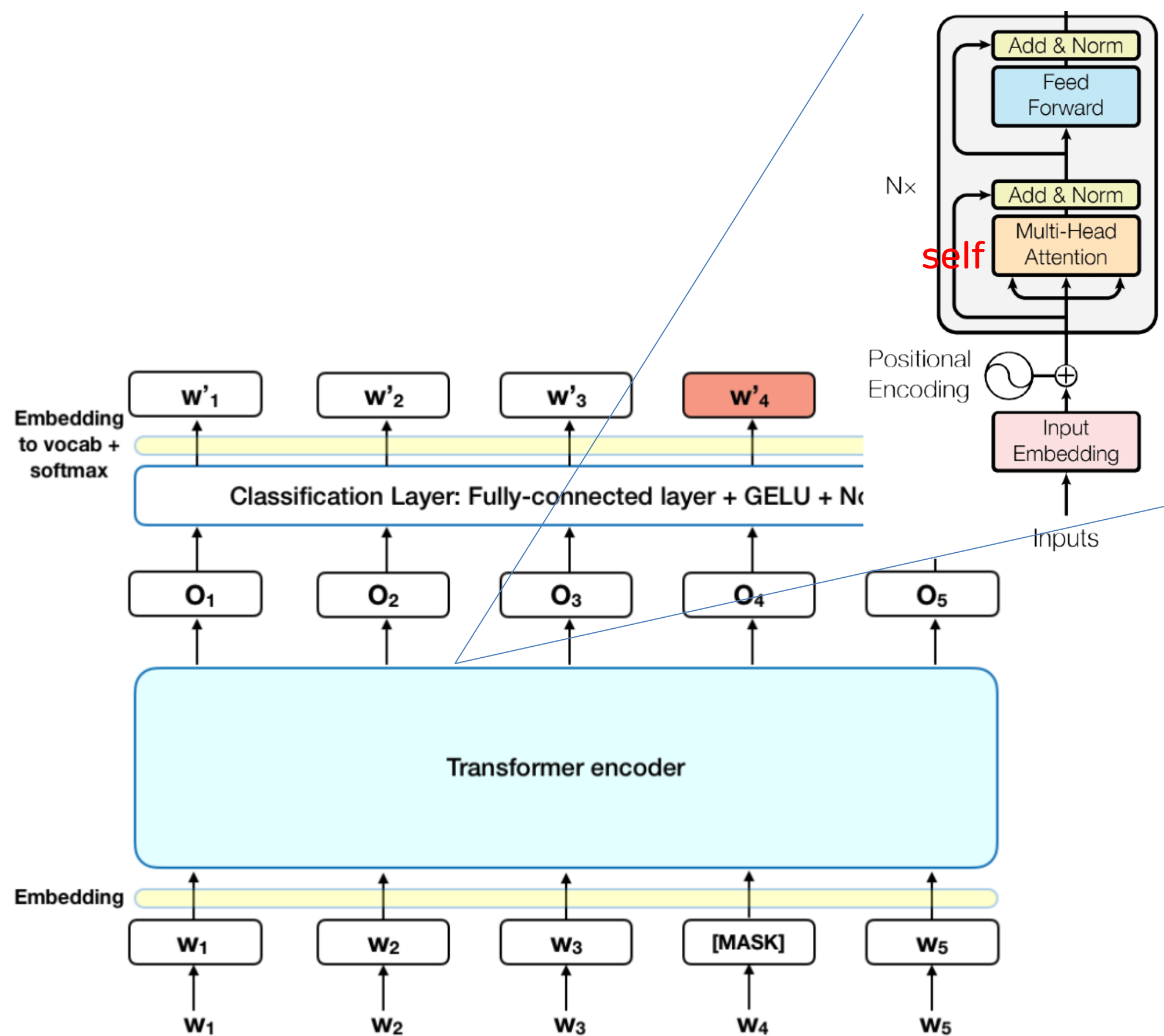
## 0. Architecture

- 초기 트랜스포머 모델 :  $L = 6, H = 512, A = 8$
- BERT-Base :  $L=12, H=768, A=12$
- BERT-Large :  $L=24, H=1024, A=16$

$L$  : num\_layers

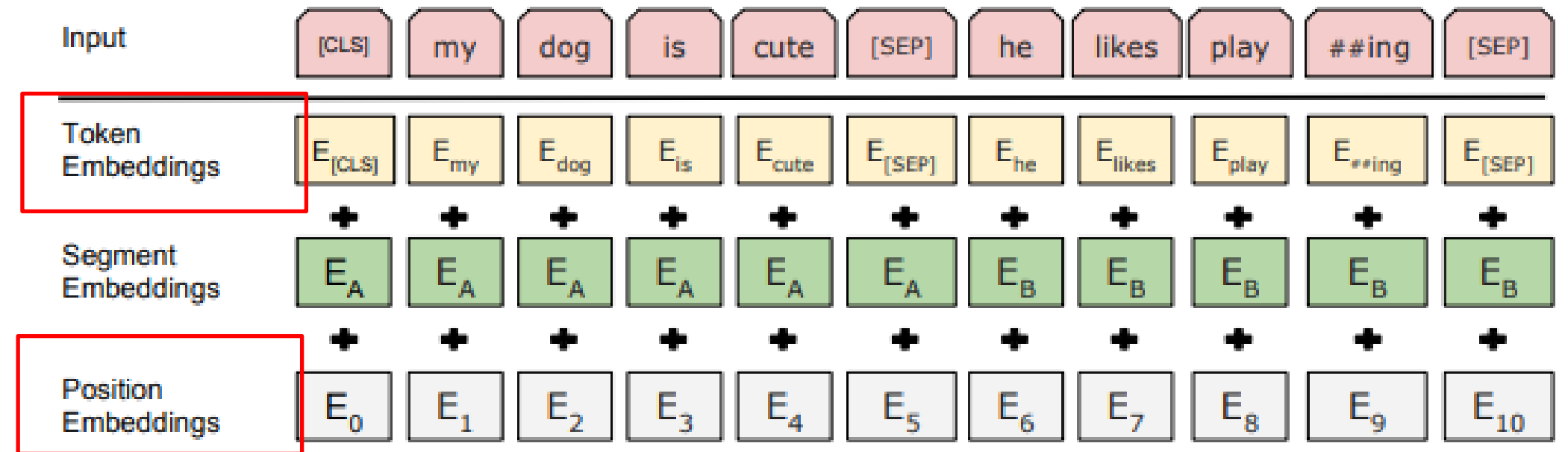
$H$  : hidden size ( $d_{\text{model}}$ )

$A$  : num attention heads



# BERT

## 1. Embedding



### WordPiece Embedding

- 토큰이 단어 집합에 존재 => 분리하지 않음

ex) 그, 나, 너, is, want, here,

- 토큰이 단어 집합에 없음 => 토큰을 서브워드로 분리

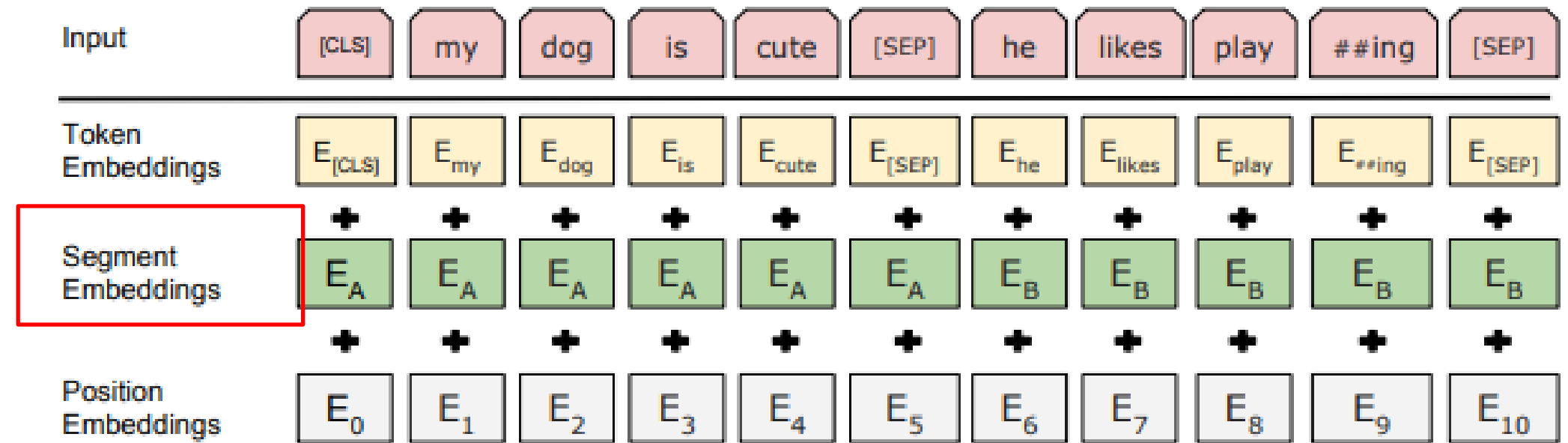
개막공연 : '개막' '##공연' , embeddings : 'em', '##bed', '##ding', '##s'

### Position Embedding

- Transformer와 유사
- pos을 sin이나 cos함수를 이용하지 않고 별도의 임베딩 층을 만들어 학습을 통해 위치 정보 만든다

# BERT

## 1. Embedding



### Segment Embedding

- 2개의 문장을 구분하기 위한 임베딩

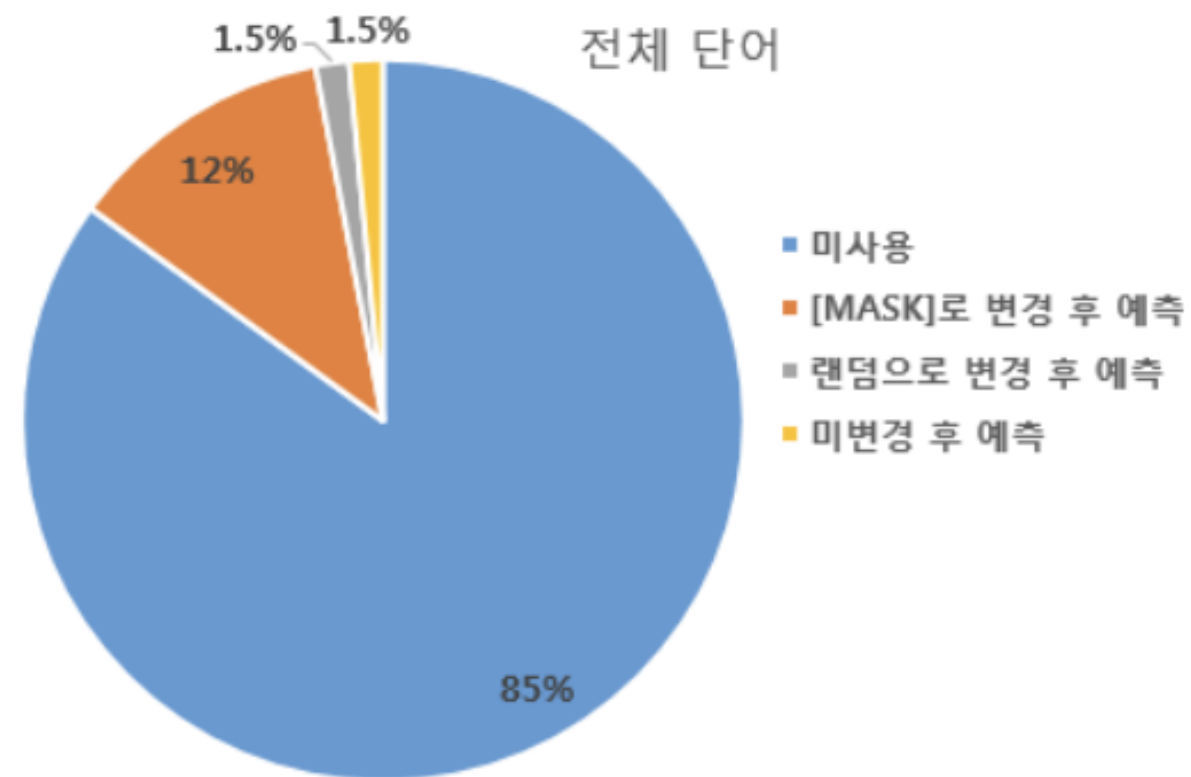
[SEP]에 의해 구분됨

a "sentence" can be an arbitrary span of contiguous text, rather than an actual linguistic sentence

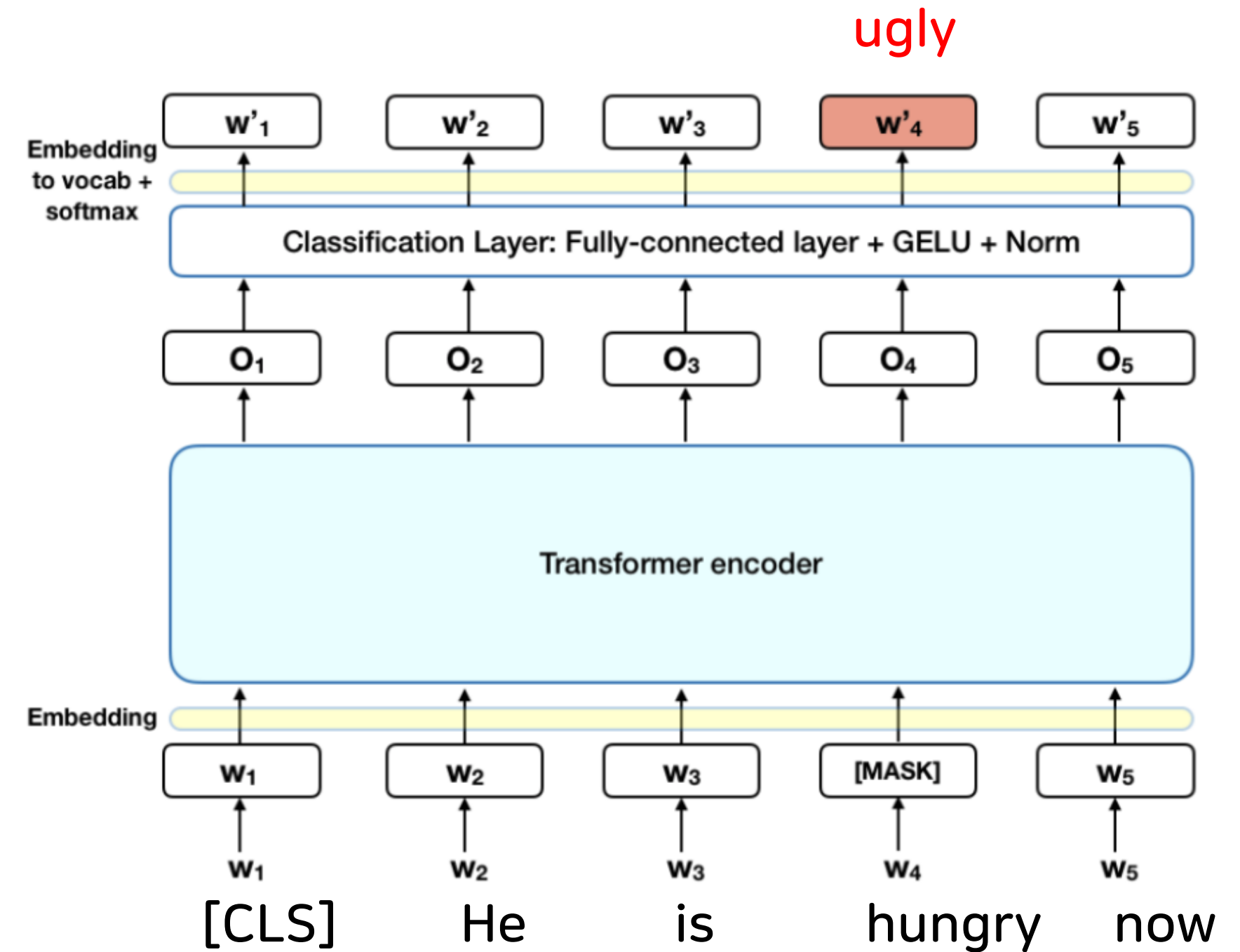
# BERT :

## 2. Pre-training

### Masked Language Model



Pre-training과 fine tuning 간의  
간극을 해소시키기 위해





# BERT

## 2. Pre-training

### Masked Language Model

```
1 from transformers import TFBertForMaskedLM
2 from transformers import AutoTokenizer
```

```
[3] 1 # 예전에 학습되었던, 미리 만들어진 model과 tokenizer
    2 model = TFBertForMaskedLM.from_pretrained('bert-large-uncased')
    3 tokenizer = AutoTokenizer.from_pretrained("bert-large-uncased")
```

```
[8] 1 from transformers import FillMaskPipeline
    2 pip = FillMaskPipeline(model=model, tokenizer=tokenizer)
```

```
1 pip("Love is like the [MASK]. You can't see it but you can feel it")
```

```
[{'score': 0.7495366334915161,
  'token': 3103,
  'token_str': 'sun',
  'sequence': "love is like the sun. you can't see it but you can feel it"},
 {'score': 0.031841378659009933,
  'token': 4231,
  'token_str': 'moon',
  'sequence': "love is like the moon. you can't see it but you can feel it"},
 {'score': 0.03020692989230156,
  'token': 3612,
  'token_str': 'wind',
  'sequence': "love is like the wind. you can't see it but you can feel it"},
 {'score': 0.024551361799240112,
  'token': 4153,
  'token_str': 'ocean',
  'sequence': "love is like the ocean. you can't see it but you can feel it"},
 {'score': 0.017378853633999825,
  'token': 3712,
  'token_str': 'sky',
  'sequence': "love is like the sky. you can't see it but you can feel it"}]
```

```
[11] 1 pip("Where are you come from? I'm from [MASK].")
```

```
[{'score': 0.05109766870737076,
  'token': 2182,
  'token_str': 'here',
  'sequence': "where are you come from? i'm from here."},
 {'score': 0.028318142518401146,
  'token': 2662,
  'token_str': 'california',
  'sequence': "where are you come from? i'm from california."},
 {'score': 0.027601782232522964,
  'token': 3146,
  'token_str': 'texas',
  'sequence': "where are you come from? i'm from texas."},
 {'score': 0.019426632672548294,
  'token': 2414,
  'token_str': 'london',
  'sequence': "where are you come from? i'm from london."},
 {'score': 0.018204212188720703,
  'token': 2563,
  'token_str': 'england',
  'sequence': "where are you come from? i'm from england."}]
```

# BERT :

## 2. Pre-training

### | Next Sentence Prediction

I am going outside

S1

I will be back after 6

S2

: IsNextSentence

I am going outside

S1

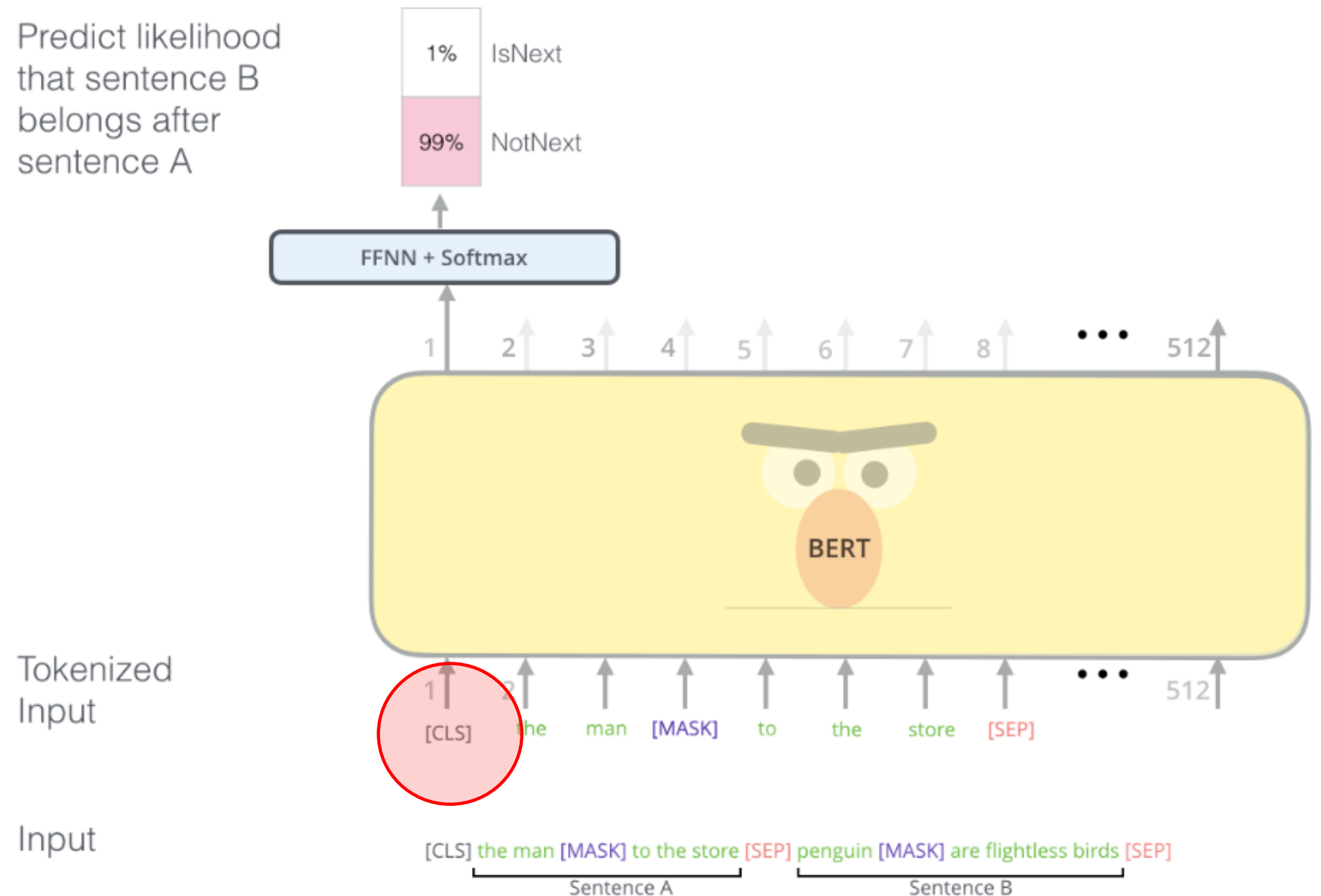
Where are you come from?

S6

: NotNextSentence

-앞뒤가 이어지는 문장: 아닌 문장의 비율을 5:5로 설정

Predict likelihood  
that sentence B  
belongs after  
sentence A



# BERT

## 2. Pre-training

### Next Sentence Prediction

```
[13] 1 import tensorflow as tf
      2 from transformers import TFBertForNextSentencePrediction
      3 from transformers import AutoTokenizer
```

```
[13] 1 model = TFBertForNextSentencePrediction.from_pretrained('bert-base-uncased')
      2 tokenizer = AutoTokenizer.from_pretrained('bert-base-uncased')
```

```
[14] 1 prompt = "Now if one were to determine what attribute the German people share with a beast, #
      2         it would be the cunning and the predatory instinct of a hawk."
      3 next_sentence = "But if one were to determine what attributes the Jews share with a beast, it would be that of the rat. #
      4         If a rat were to walk in here right now as I'm talking, would you treat it with a saucer of your delicious milk?"
```

```
[20] 1 logits = model(encoding['input_ids'], token_type_ids=encoding['token_type_ids'])[0]
      2 softmax = tf.keras.layers.Softmax()
      3 probs = softmax(logits)
      4 print(probs)
```

```
tf.Tensor([[9.9999774e-01 2.2407737e-06]], shape=(1, 2), dtype=float32)
```

```
[21] 1 print('최종 예측 레이블 :', tf.math.argmax(probs, axis=-1).numpy())
```

최종 예측 레이블 : [0]

```
[22] 1 # 상관없는 두 개의 문장
      2 next_sentence = "The sky is blue due to the shorter wavelength of blue light."
      3 encoding = tokenizer(prompt, next_sentence, return_tensors='tf')
      4
      5 logits = model(encoding['input_ids'], token_type_ids=encoding['token_type_ids'])[0]
      6
      7 softmax = tf.keras.layers.Softmax()
      8 probs = softmax(logits)
      9 print('최종 예측 레이블 :', tf.math.argmax(probs, axis=-1).numpy())
```

최종 예측 레이블 : [1]

# BERT :

## 3. Fine-tuning

QA : Question and Answer 질의응답

NER : Named Entity Recognition 개체명 인식

MNLI : Multi-Genre Natural Language Inference

