
MLM Fine Tuning

A Multi-Stage Transfer Learning Approach for Personality Classification
Using Masked Language Modeling and Task-Specific Fine-Tuning

FIVE GUYS

김민규 성재영 이영주 이준성 이현준

Method

MLM Fine Tuning

Task

MBTI Classification

Transformer

Attention is all you need

Google (2017)

BERT

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

Google (2018)

ELECTRA

ELECTRA: Pre-Training Text Encoders as Discriminators rather than Generators

Google Research (2020)

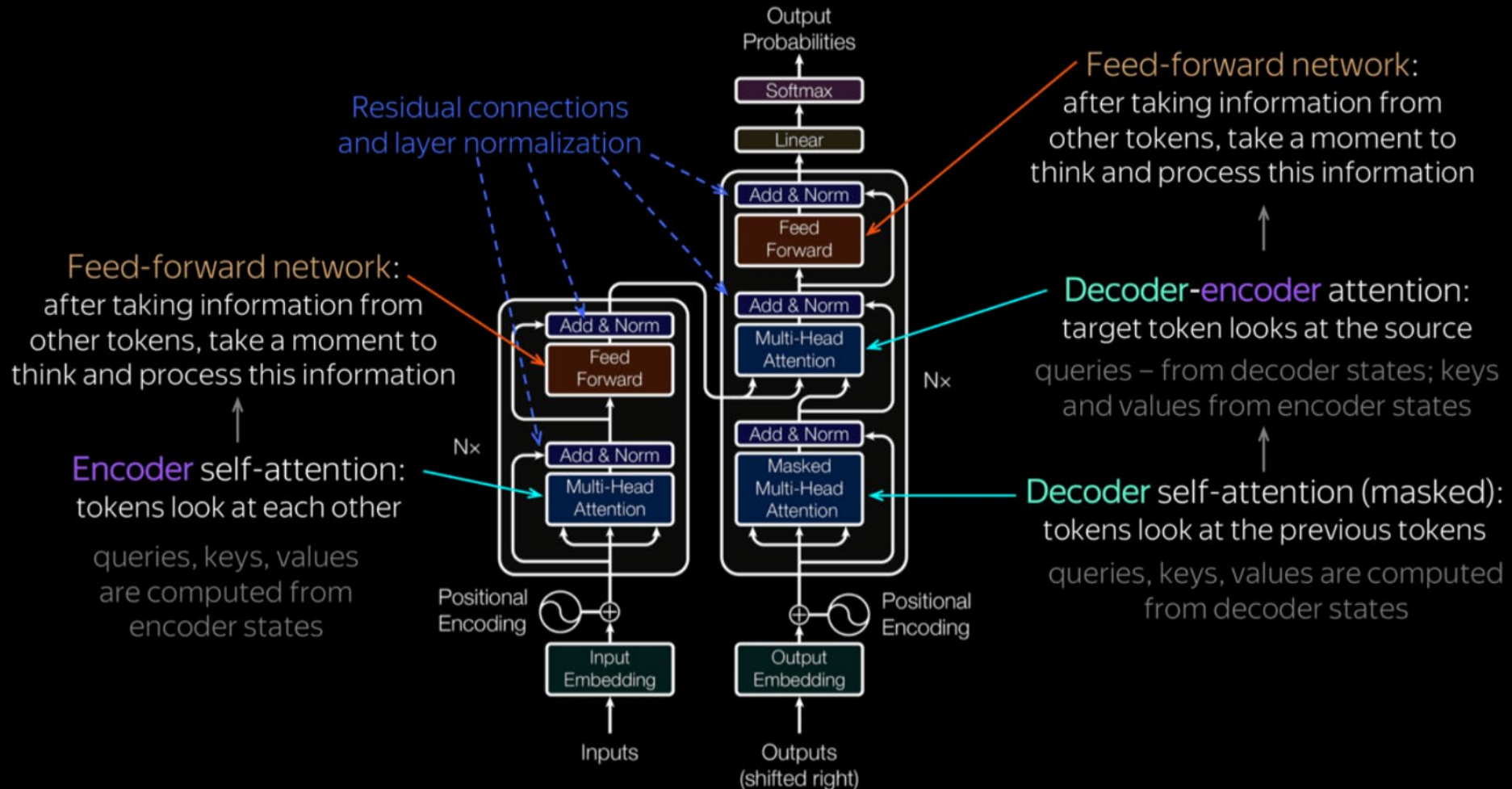
KcELECTRA

KcELECTRA: Korean Comments ELECTRA

Lee Jun Bum (2022)

Transformer

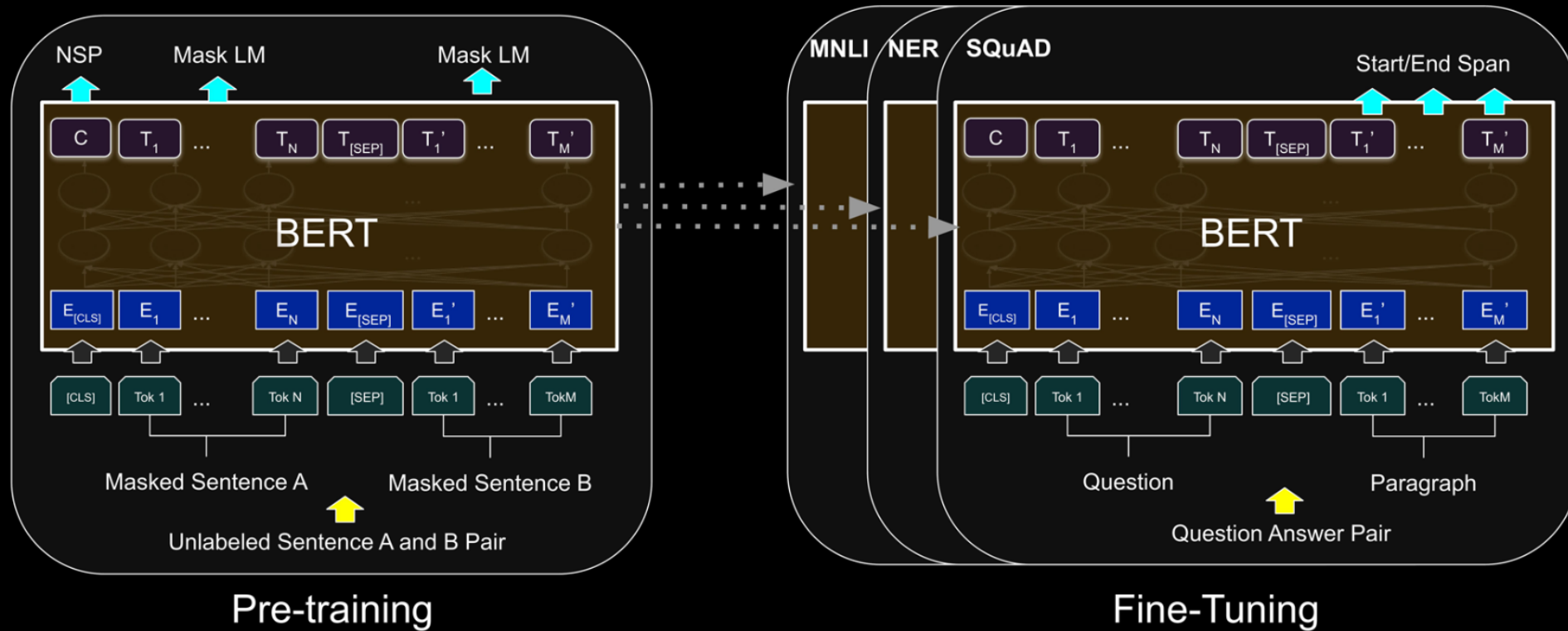
Google (2017) Attention is all you need



BERT

MLM Fine Tuning

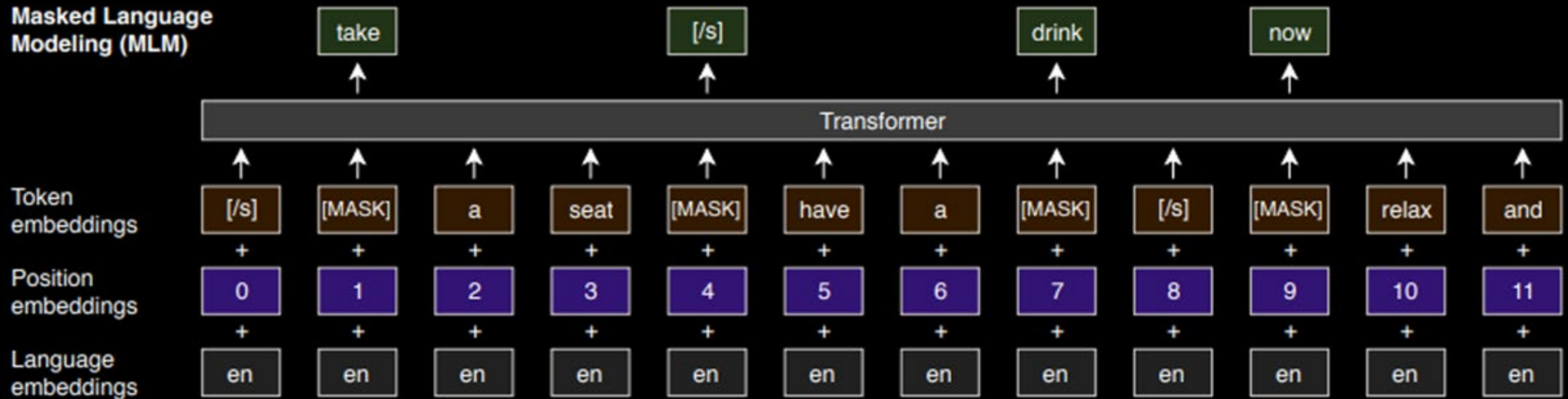
Google (2018) BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding



MLM (Masked Language Model)

NSP (Next Sentence Prediction)

MLM (Masked Language Modeling)



Masking ratio

15 %

MLM in BERT

Masking Rates			Dev Set Results		
MASK	SAME	RND	MNLI Fine-tune	NER Fine-tune	NER Feature-based
80%	10%	10%	84.2	95.4	94.9
100%	0%	0%	84.3	94.9	94.0
80%	0%	20%	84.1	95.2	94.6
80%	20%	0%	84.4	95.2	94.7
0%	20%	80%	83.7	94.8	94.6
0%	0%	100%	83.6	94.9	94.6

Masking ratio

15 %

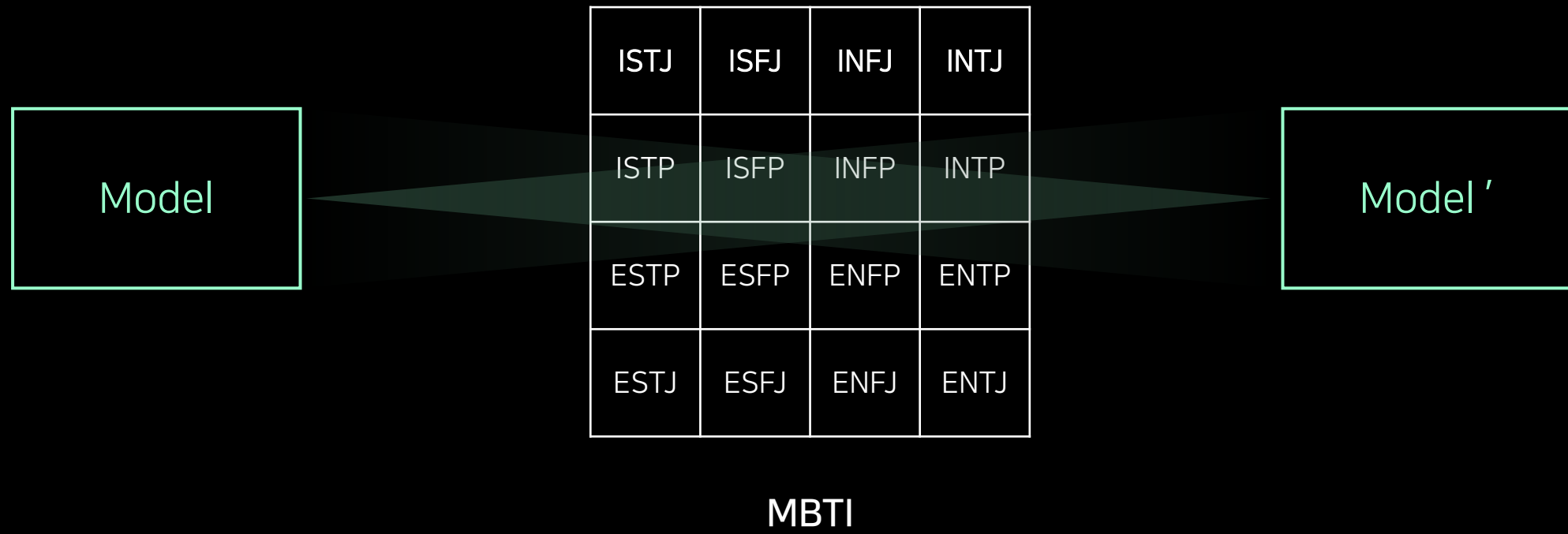
MASK (80%) my dog is hairy → my dog is [MASK]

RND (10%) my dog is hairy → my dog is apple

SAME (10%) my dog is hairy → my dog is hairy

Domain Specific MLM

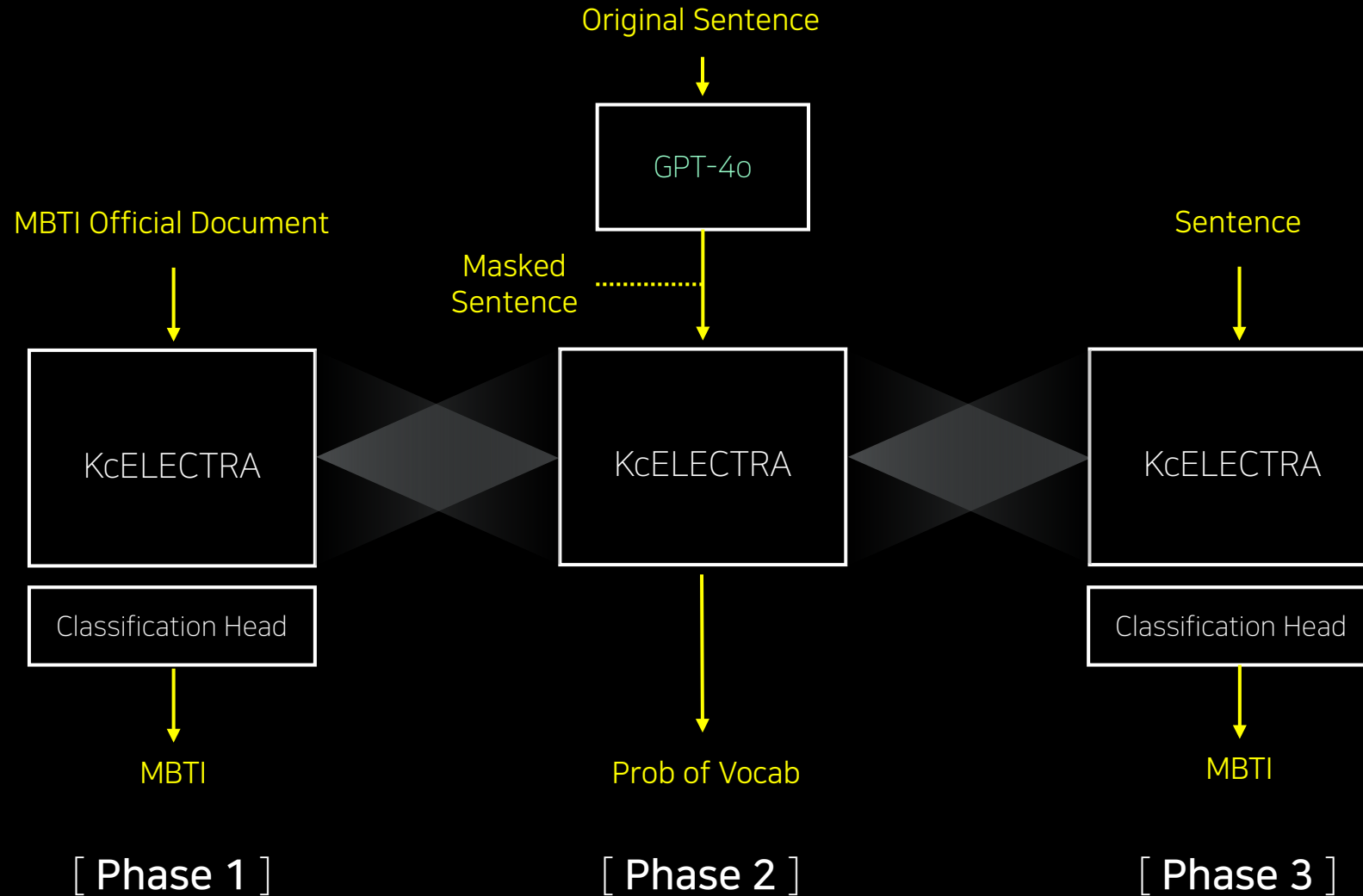
I went to [MASK] after lunch

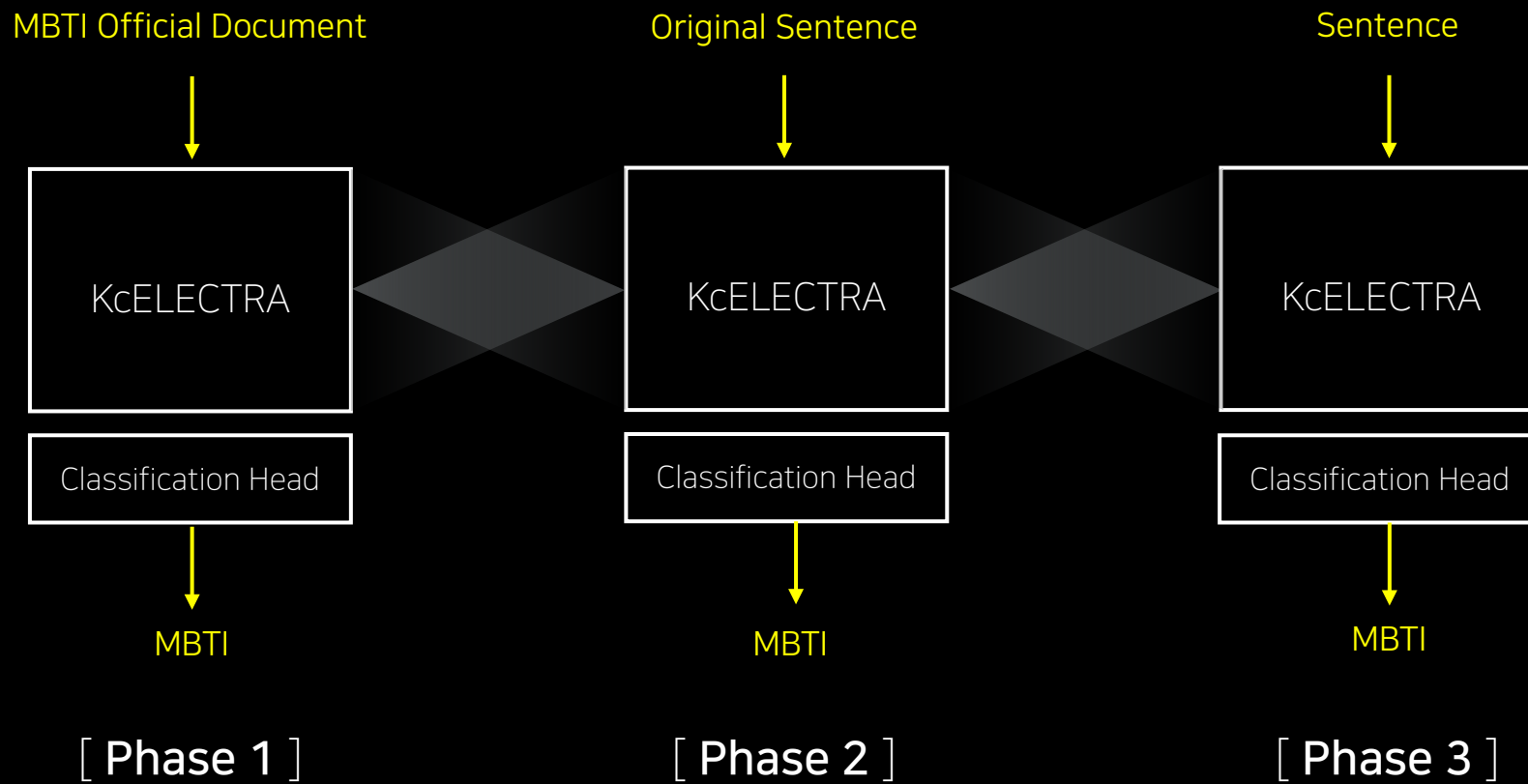


MLM Fine Tuning

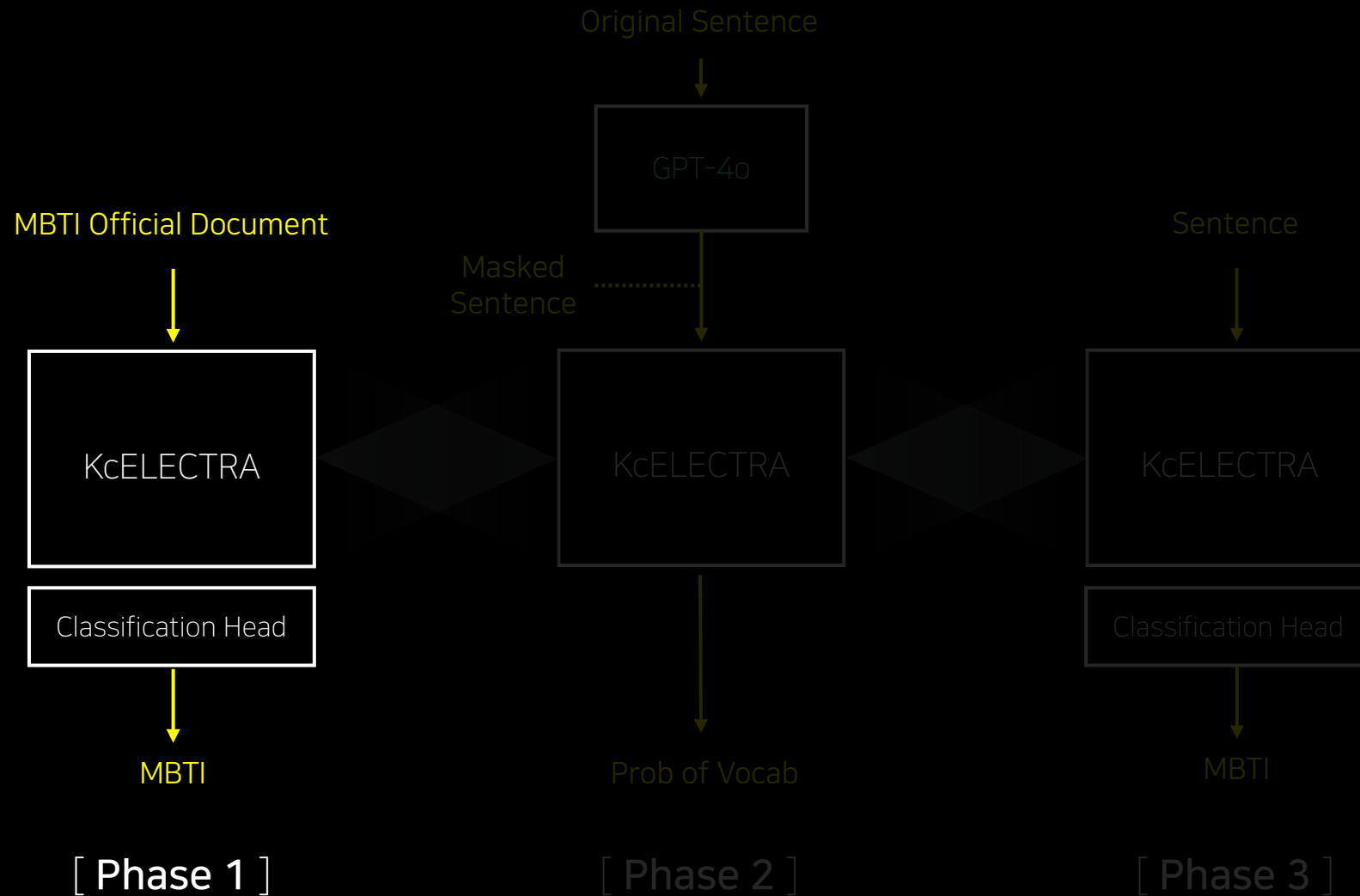
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MLM Fine Tuning





Phase 1



MBTI Official Document Dataset

[Personality Test](#)[Personality Types](#) ▾[Premium Suite](#)[Teams and Practitioners](#) ▾[Resources](#) ▾[Log In](#)[Take the Test](#)

그들은 급격한 합리적 자극 반응 과정에서 사실적이고 즉각적인 현실을 바탕으로 비판적 결정을 내리는 경향이 있습니다.

이것은 학교와 다른 고도로 조직화 된 환경을 ESTP에 대한 도전으로 만듭니다. 확실히 똑똑하지 않기 때문이 아니며, 그곳에서 잘 할 수는 있지만, 공식 교육의 연대적인 강의 접근은 이러한 성격이 일반적으로 즐기는 실습 학습과는 거리가 멉니다.

ESTP

Dataset

1440

=

Article

90

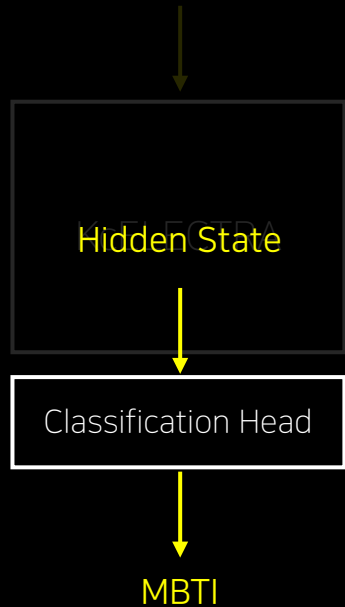
X

MBTI

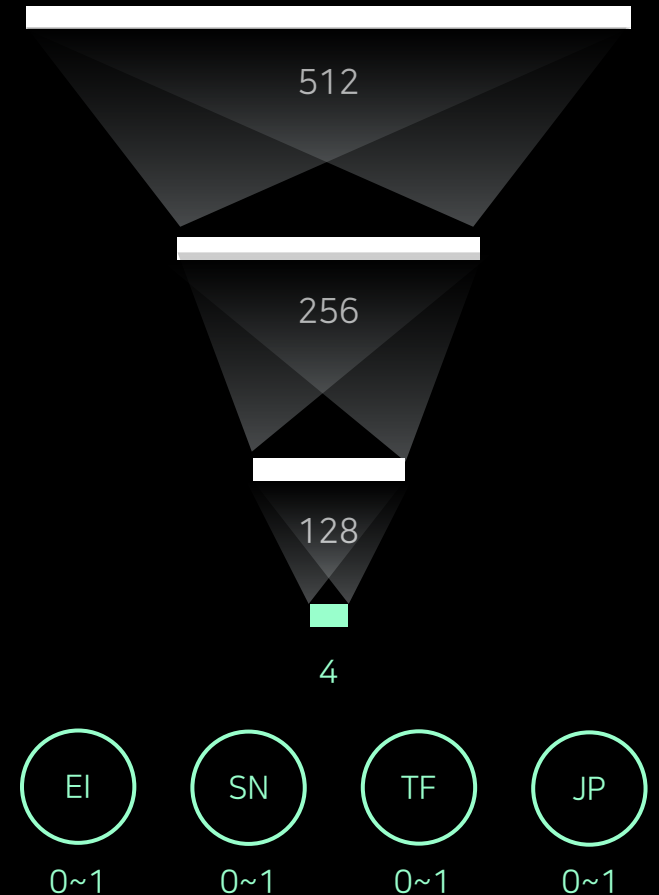
16

Classification Head

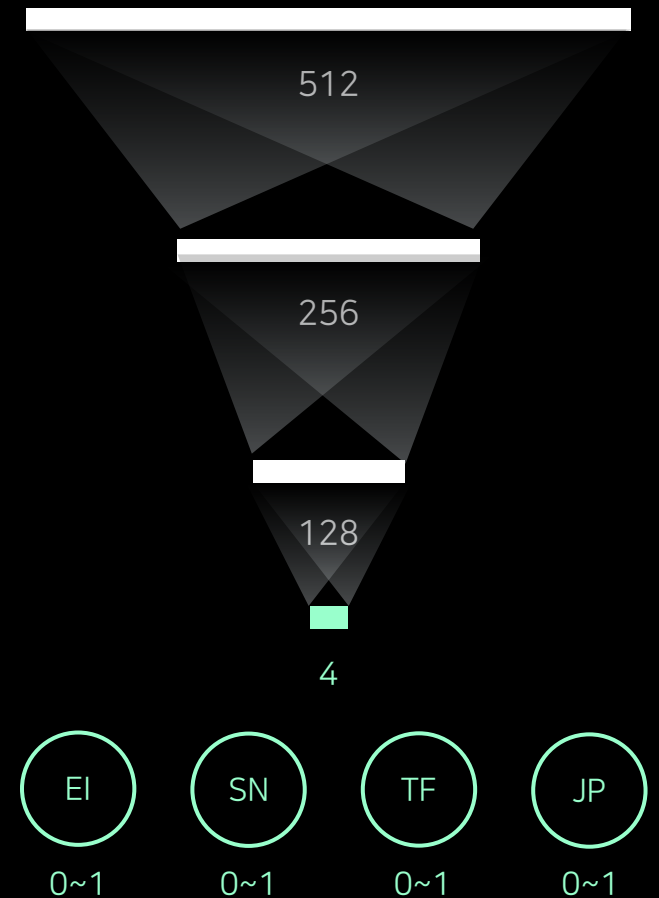
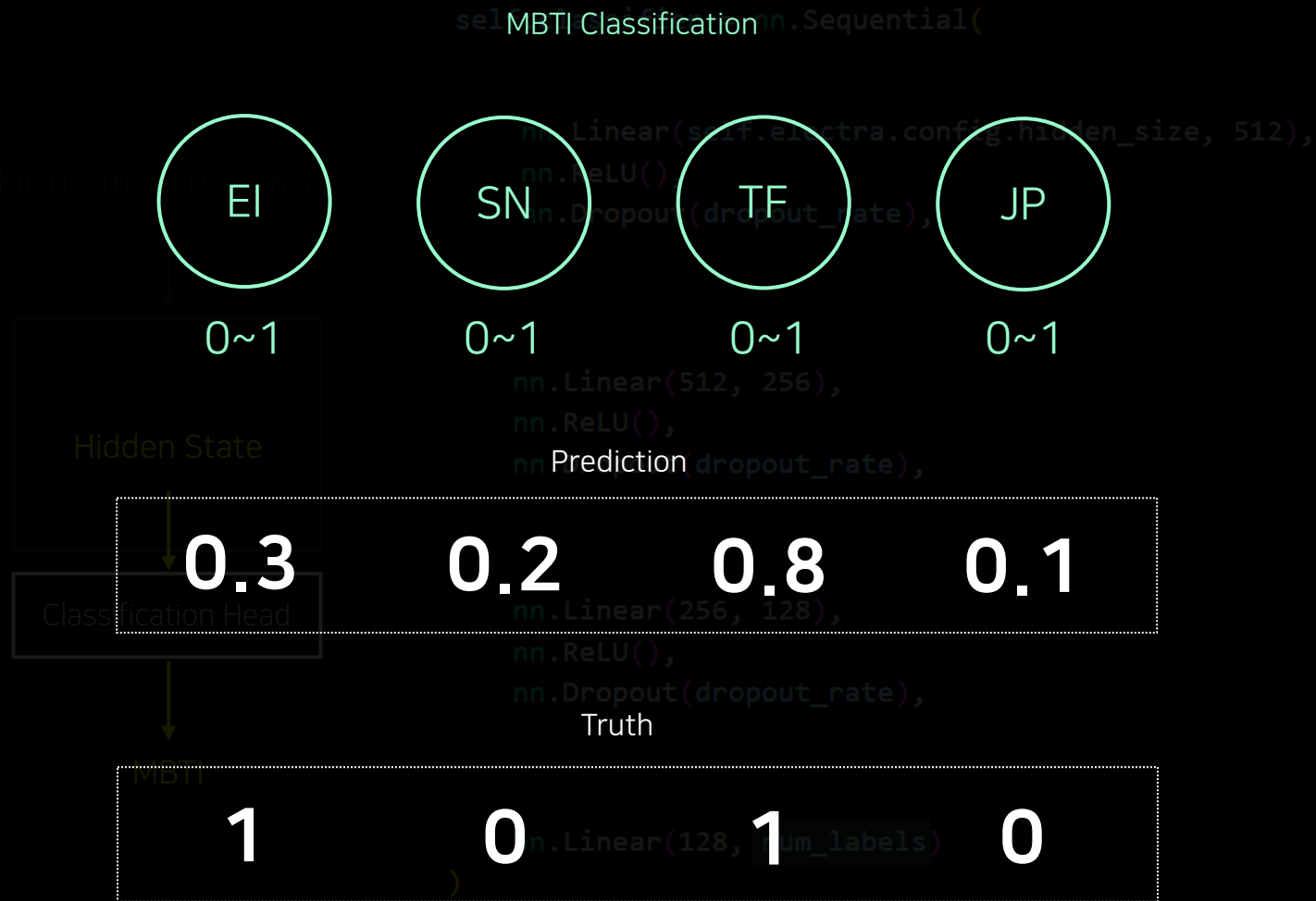
MBTI Official Document



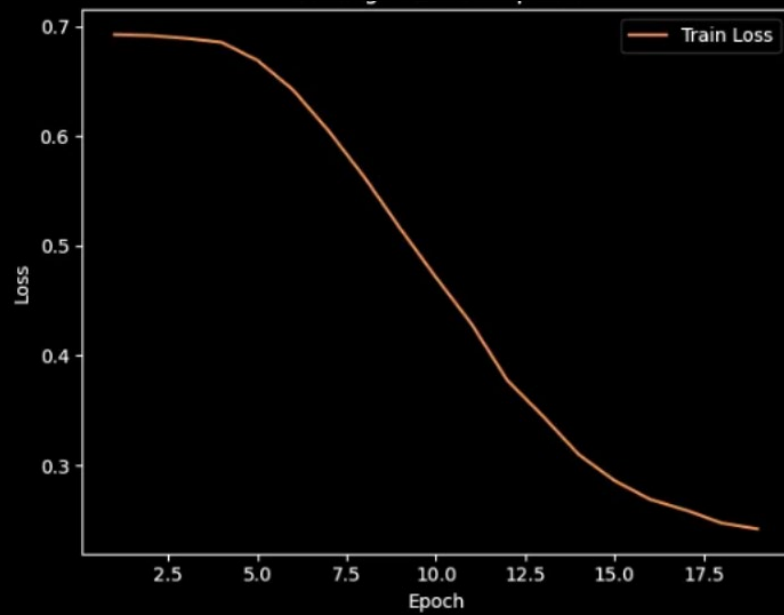
```
self.classifier = nn.Sequential(  
  
    nn.Linear(self.electra.config.hidden_size, 512),  
    nn.ReLU(),  
    nn.Dropout(dropout_rate),  
  
    nn.Linear(512, 256),  
    nn.ReLU(),  
    nn.Dropout(dropout_rate),  
  
    nn.Linear(256, 128),  
    nn.ReLU(),  
    nn.Dropout(dropout_rate),  
  
    nn.Linear(128, num_labels)  
)
```



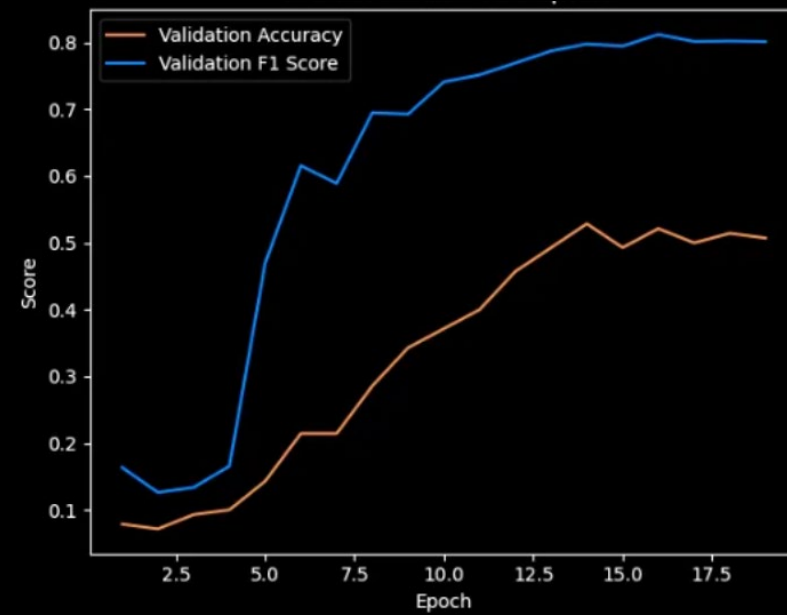
Classification Head



Training

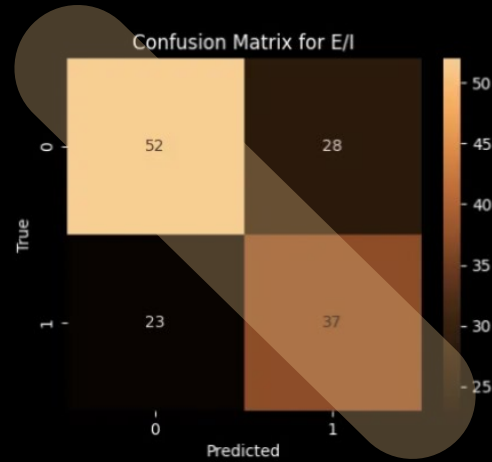


Training Loss Over Epochs

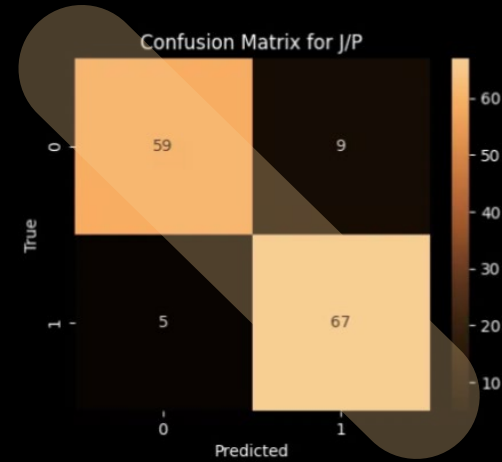


Validation Metrics Over Epochs

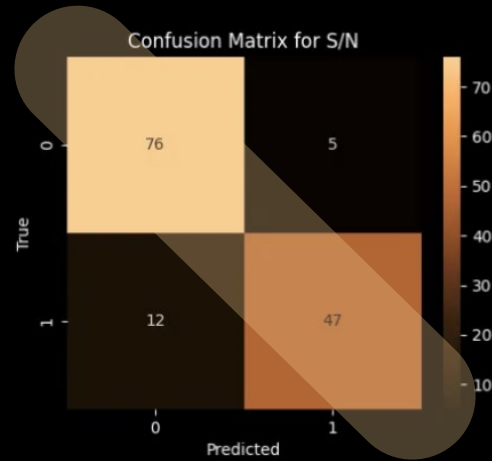
EI



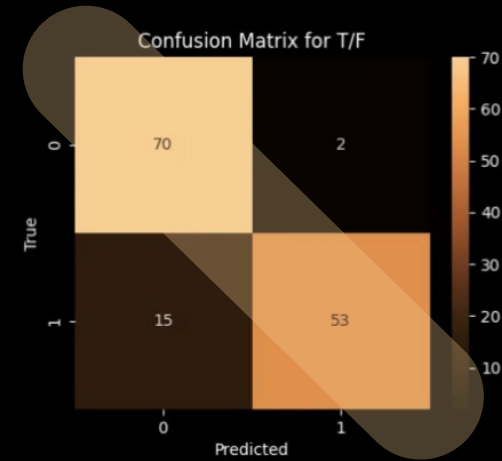
JP

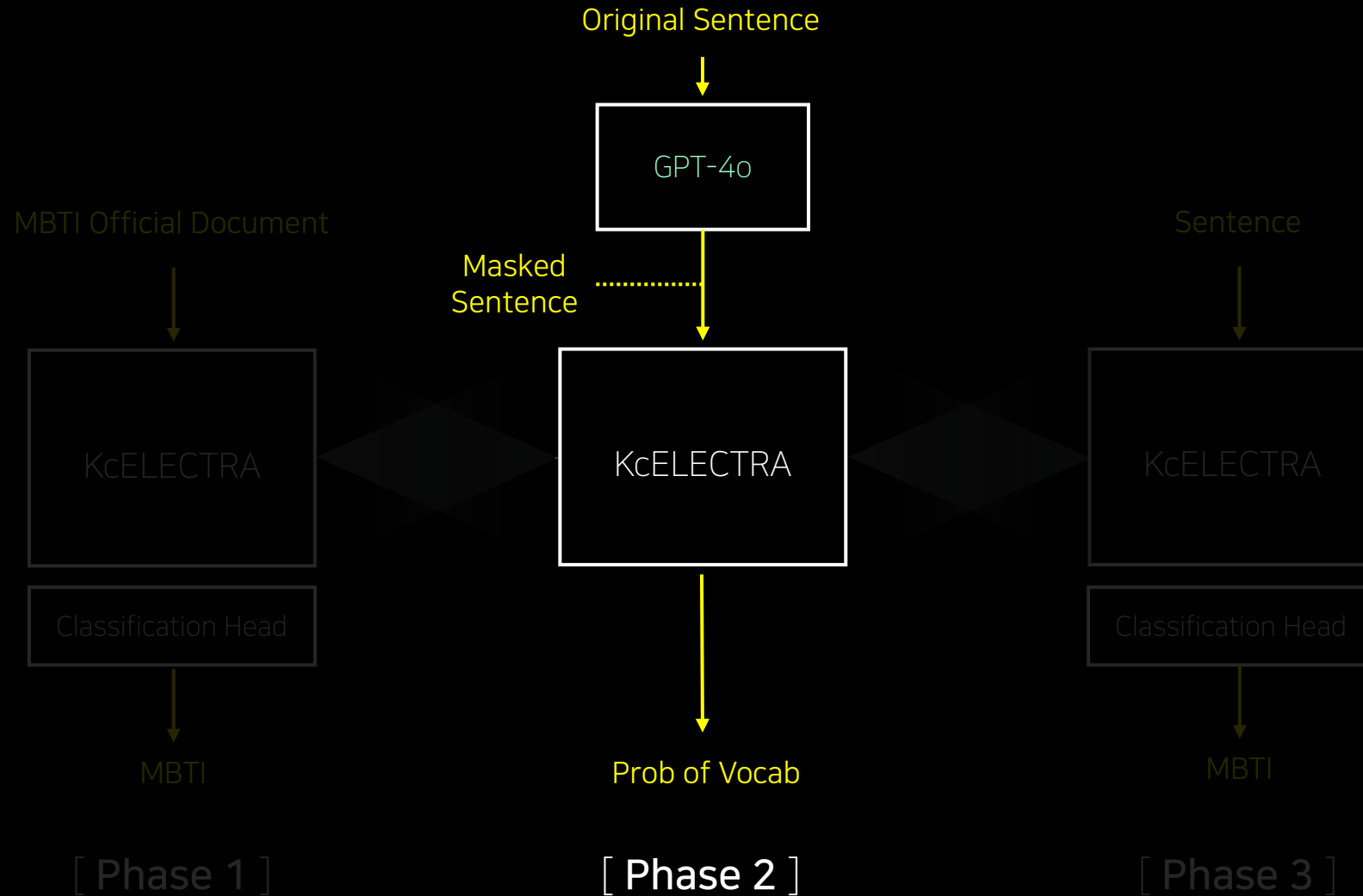


SN

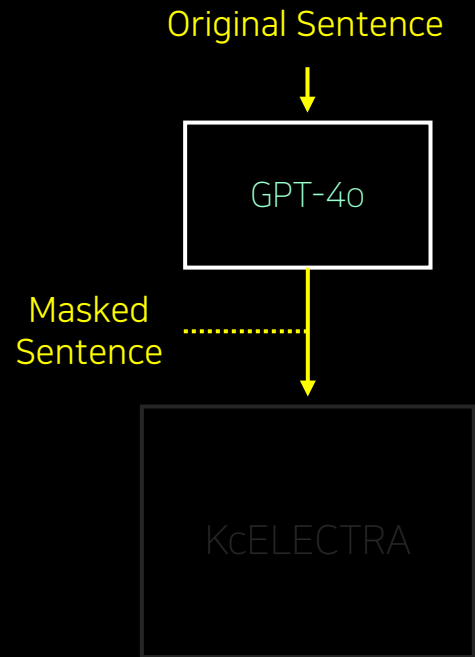


TF





Masked Dataset



Dataset

14,400

Prob of Vocab

=

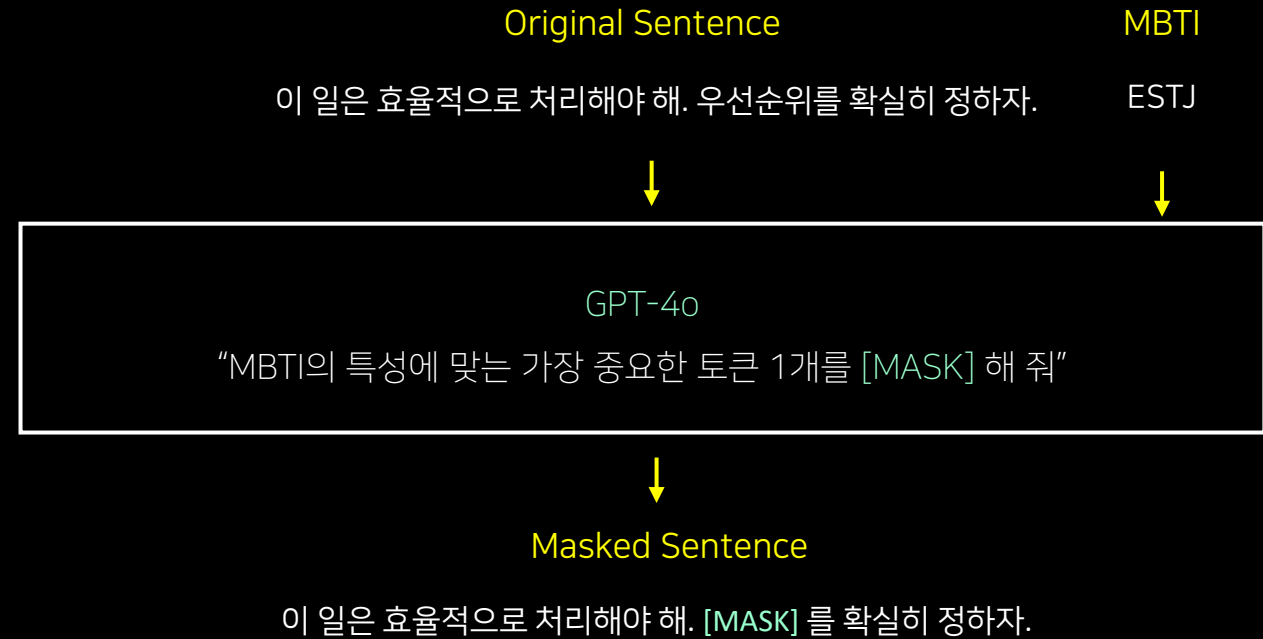
Original Sentence

900

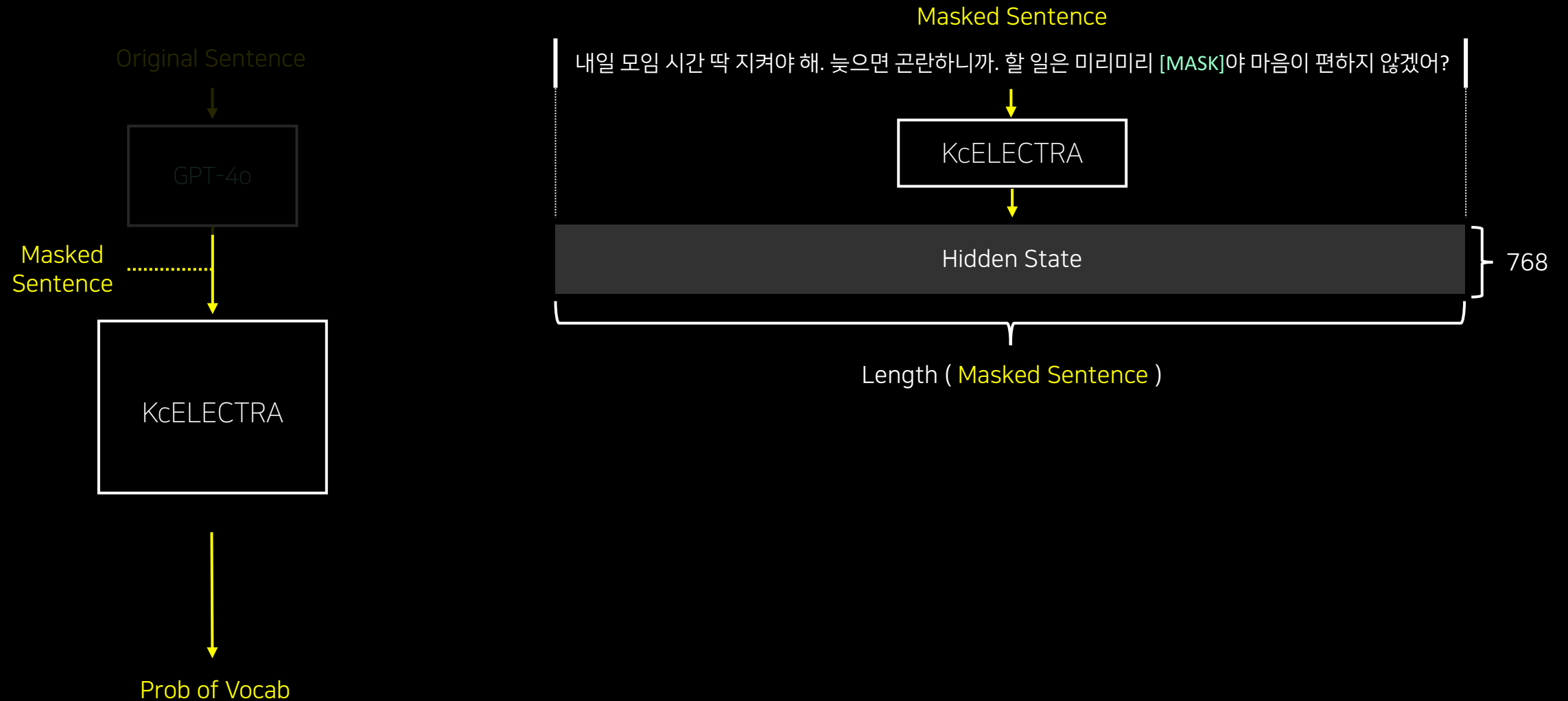
×

MBTI

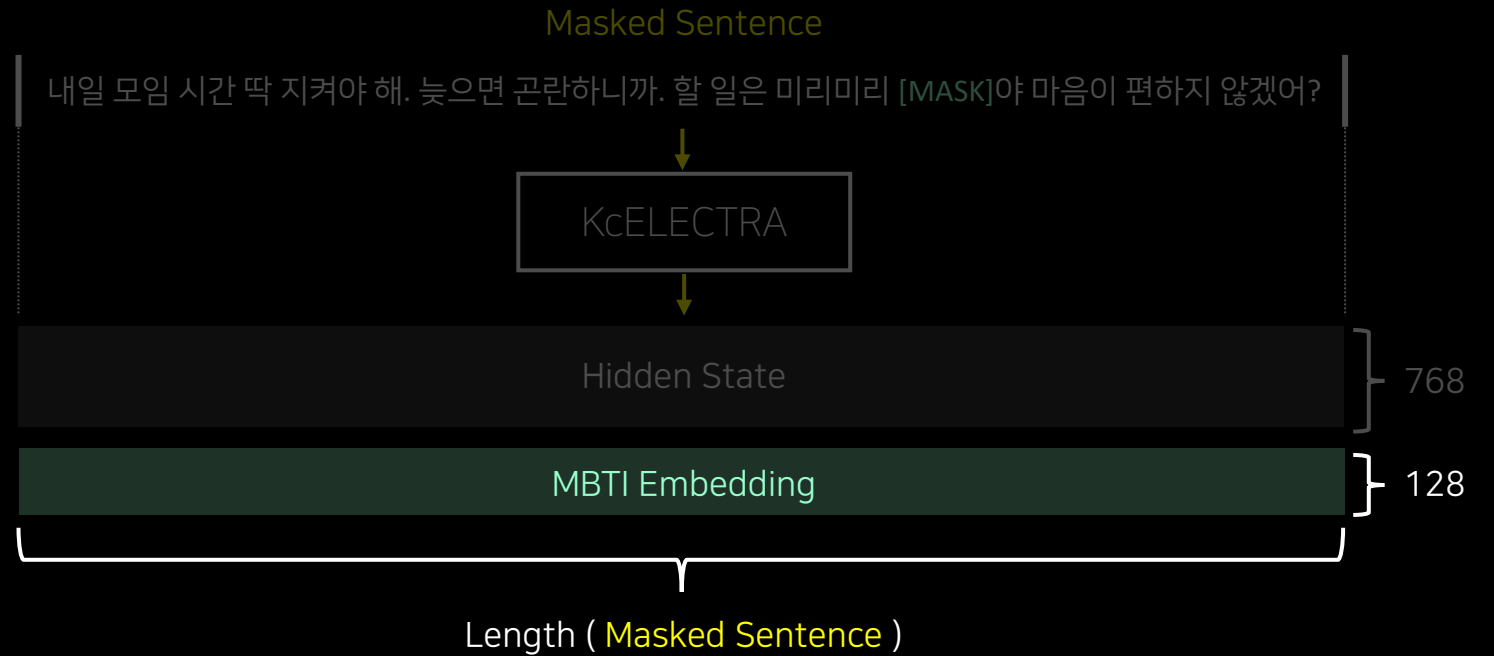
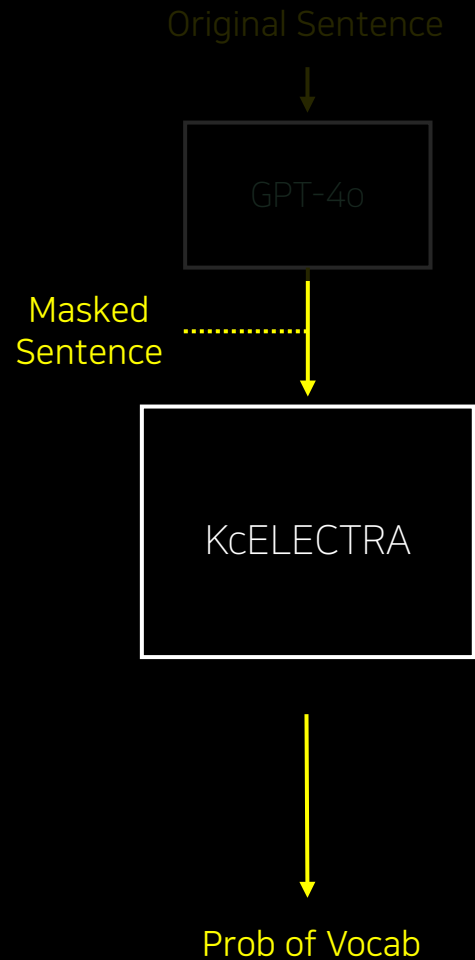
16



Hidden State



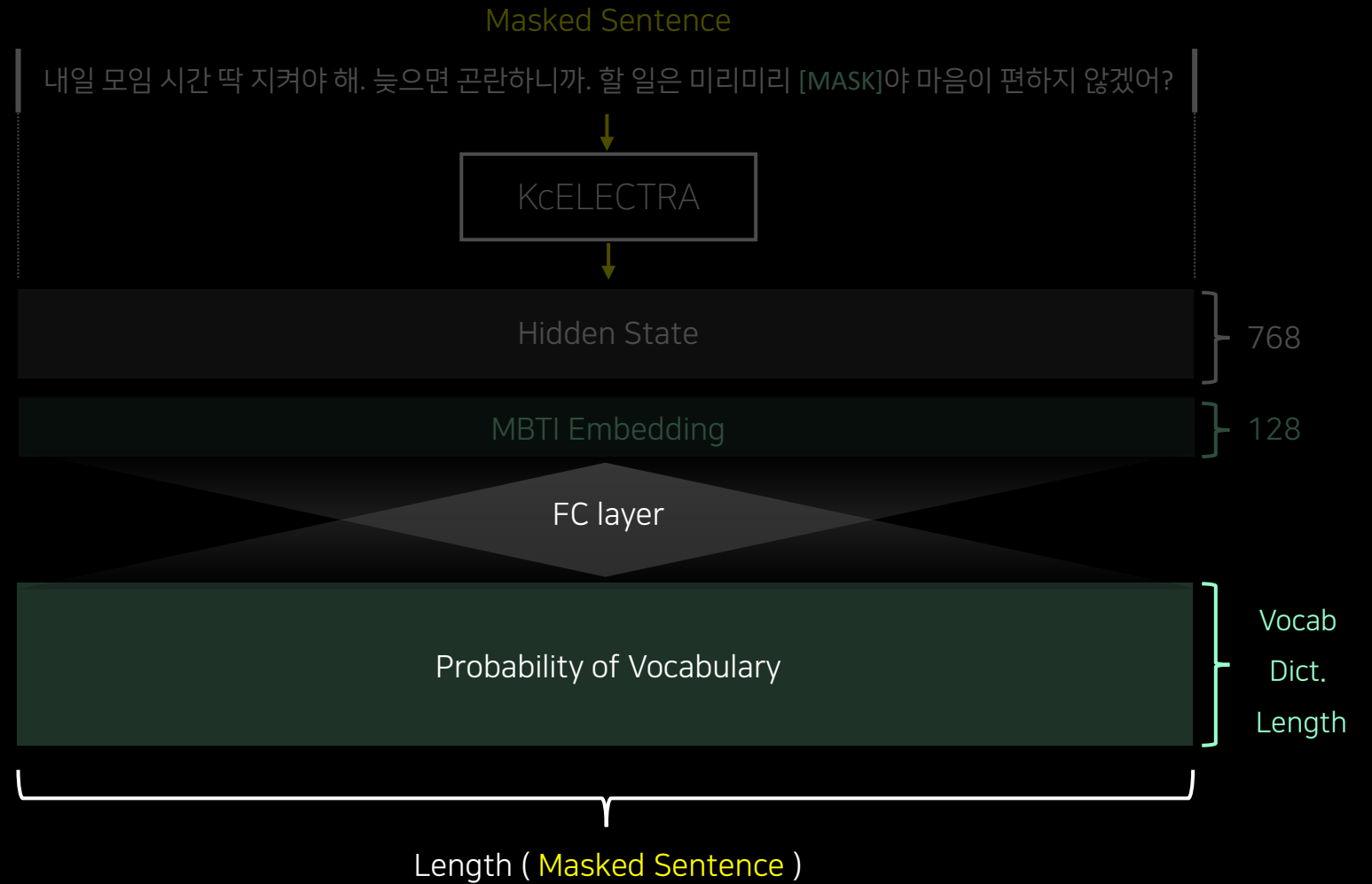
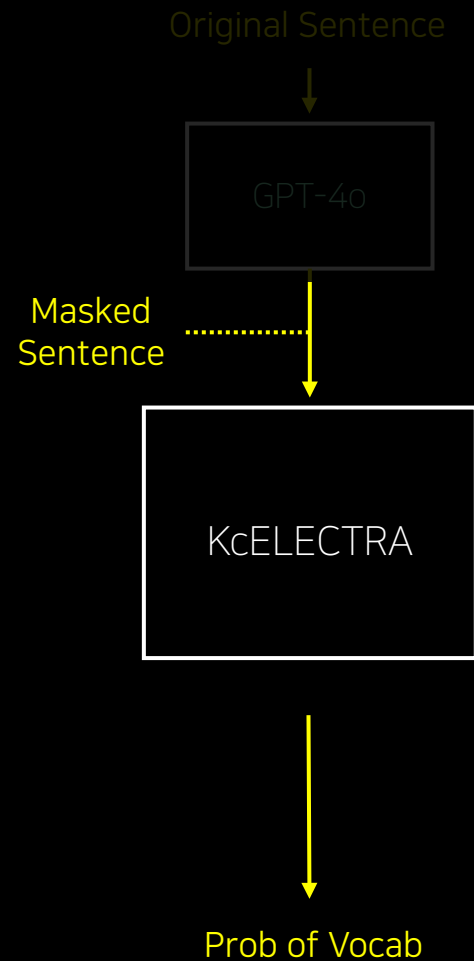
MBTI Embedding



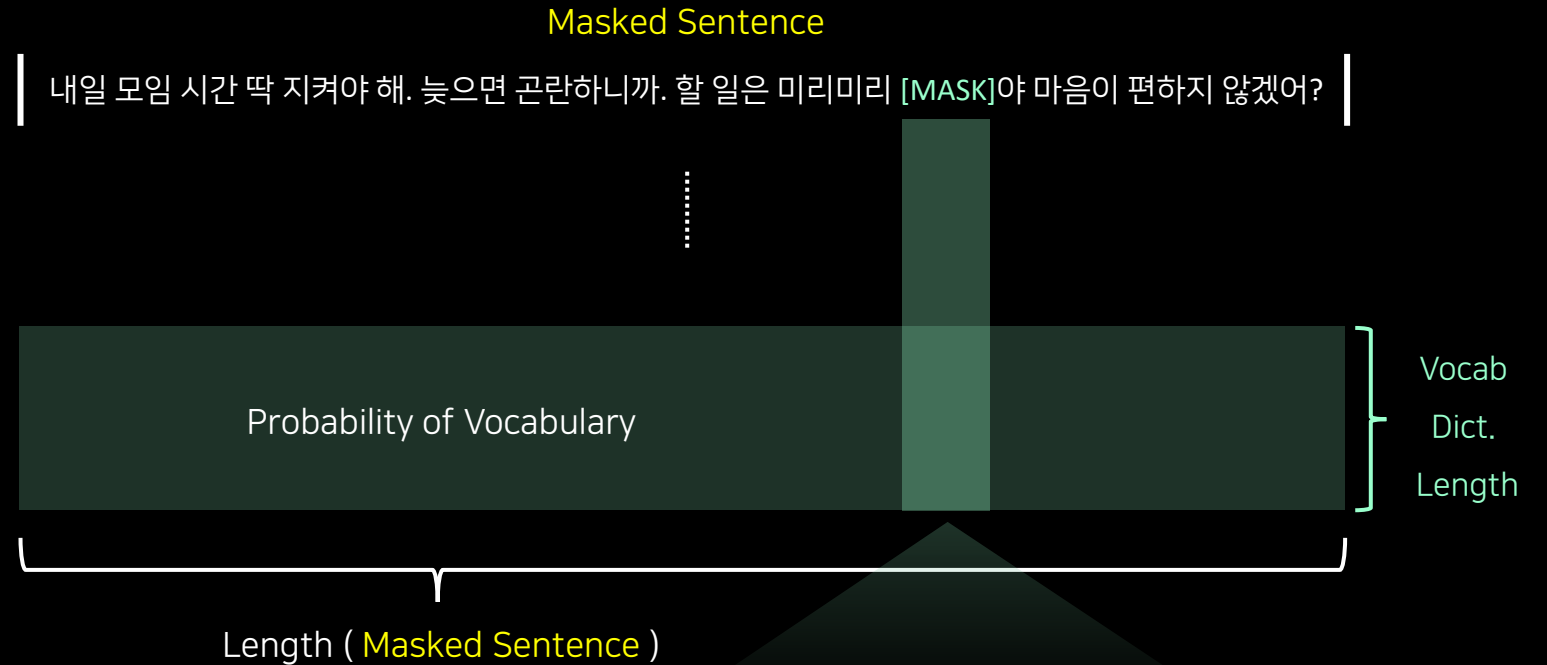
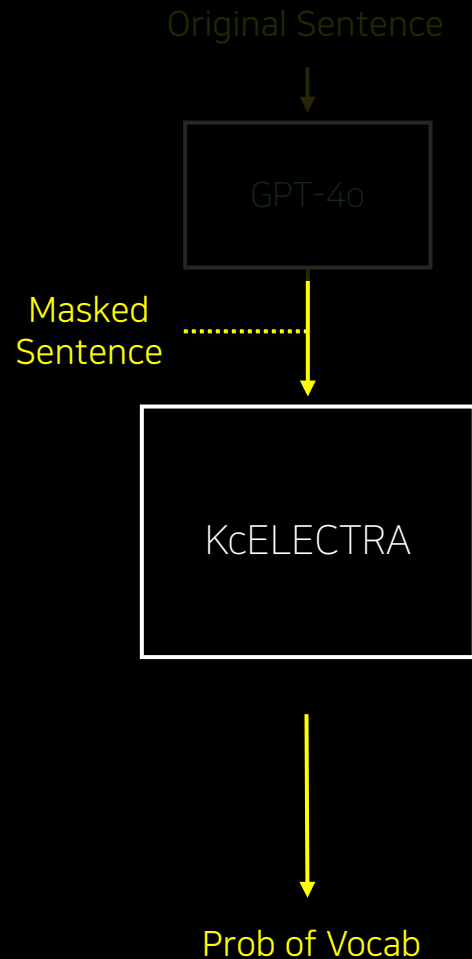
```
MBTI_TYPES = [  
    'INTJ', 'INTP', 'ENTJ', 'ENTP', 'INFJ', 'INFP', 'ENFJ', 'ENFP',  
    'ISTJ', 'ISFJ', 'ESTJ', 'ESFJ', 'ISTP', 'ISFP', 'ESTP', 'ESFP'  
]
```

0~15

Probability of Vocabulary



Probability of Vocabulary



ISTJ

Vocab 1	잘해	(Prob: 0.0143)
Vocab 2	정리해	(Prob: 0.0119)
Vocab 3	기록해	(Prob: 0.0080)
Vocab 4	체계	(Prob: 0.0075)
Vocab 5	세워	(Prob: 0.0071)

Probability of Vocabulary

Masked Sentence

내일 모임 시간 딱 지켜야 해. 늦으면 곤란하니까. 할 일은 미리미리 [MASK]야 마음이 편하지 않겠어?

ISTJ

Vocab 1	잘해	(Prob: 0.0143)
Vocab 2	정리해	(Prob: 0.0119)
Vocab 3	기록해	(Prob: 0.0080)
Vocab 4	체계	(Prob: 0.0075)
Vocab 5	세워	(Prob: 0.0071)

INTJ

Vocab 1	적으로	(Prob: 0.0201)
Vocab 2	돼야	(Prob: 0.0201)
Vocab 3	잘	(Prob: 0.0168)
Vocab 4	체계	(Prob: 0.0164)
Vocab 5	정리	(Prob: 0.0154)

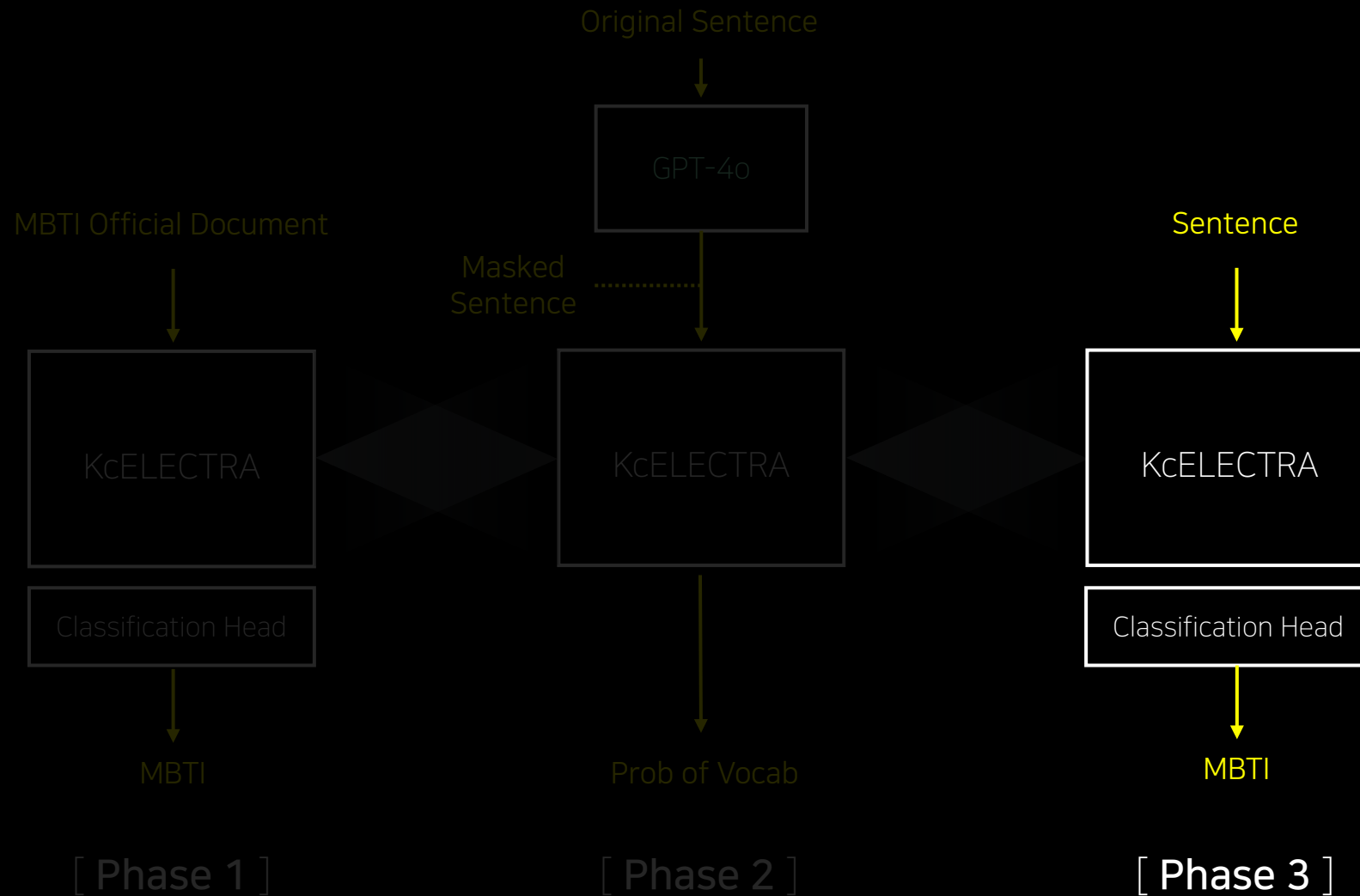
ENFP

Vocab 1	자유롭게	(Prob: 0.0118)
Vocab 2	풀어	(Prob: 0.0111)
Vocab 3	감동	(Prob: 0.0111)
Vocab 4	잘	(Prob: 0.0111)
Vocab 5	신기	(Prob: 0.0085)

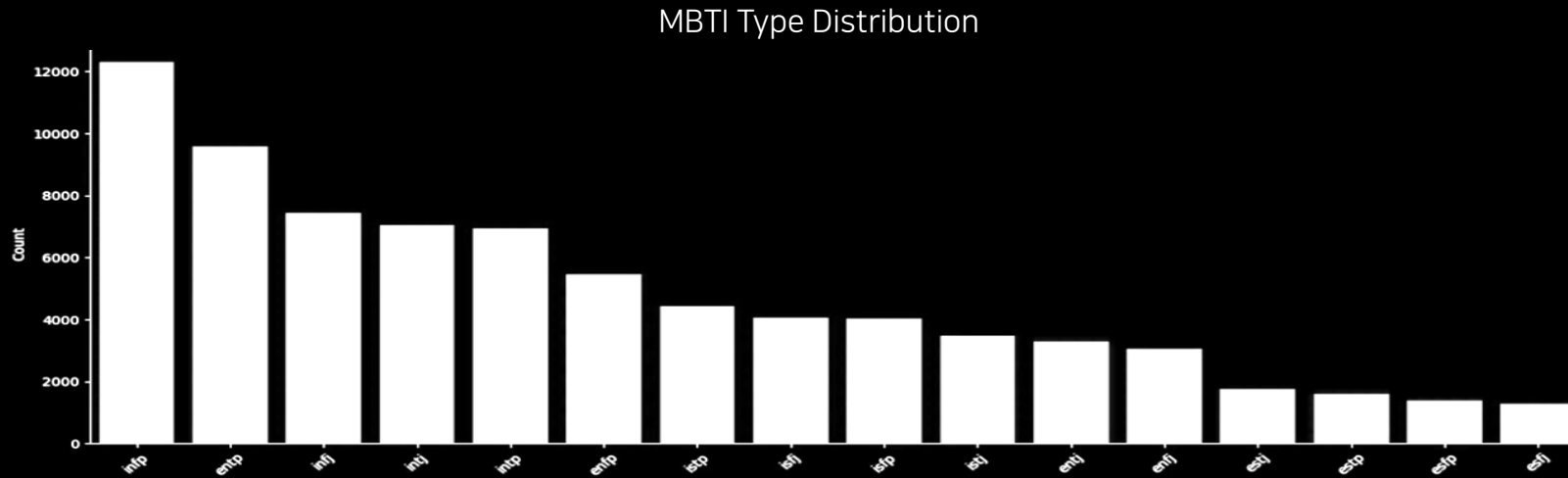
ESFP

Vocab 1	자유롭게	(Prob: 0.0188)
Vocab 2	잘	(Prob: 0.0133)
Vocab 3	풀어	(Prob: 0.0094)
Vocab 4	정리	(Prob: 0.0079)
Vocab 5	효율	(Prob: 0.0068)

Phase 3



Article Dataset



님들이 과자나 치킨 같은 거 사 와서 먹으라고 했는데 동생이 안 먹겠다고 해서 방치해 놔던
먹을거리가 다음날 보니 말도 없이 없어졌을 때 화나 세영???

ISTJ

Article

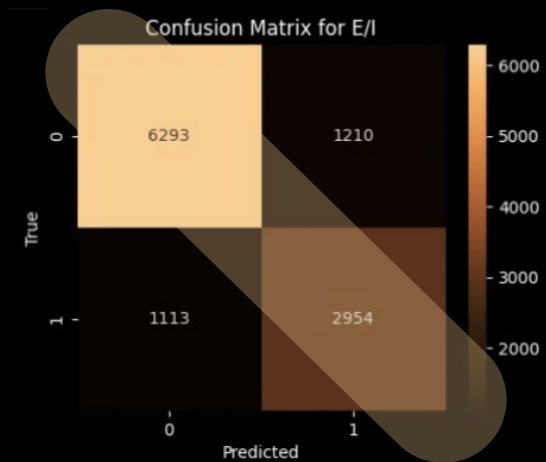
MBTI

80,000

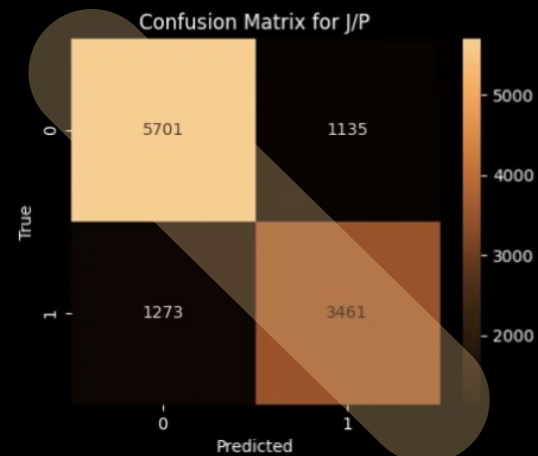
Dataset

MLM Fine Tuning	
Test Loss	1.1752
Test Accuracy	58.26

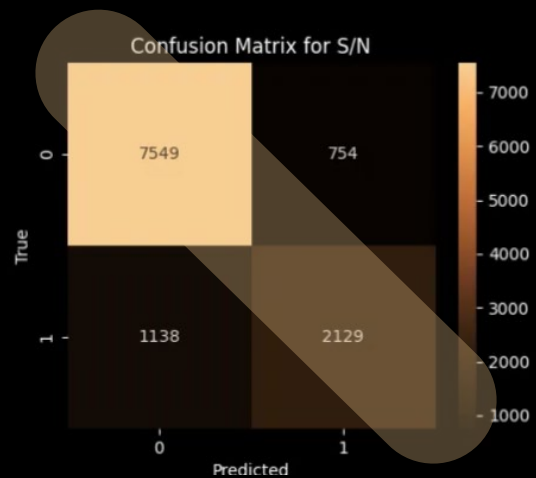
EI



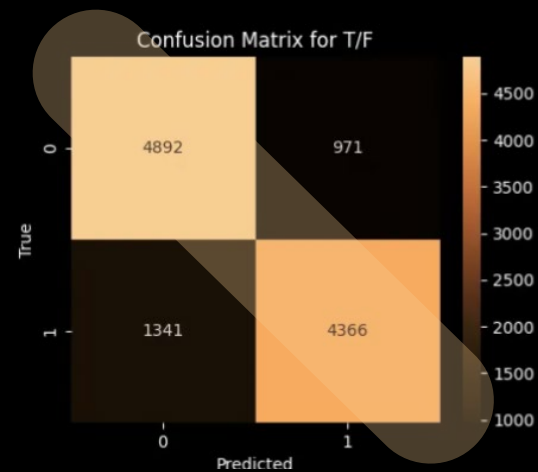
JP



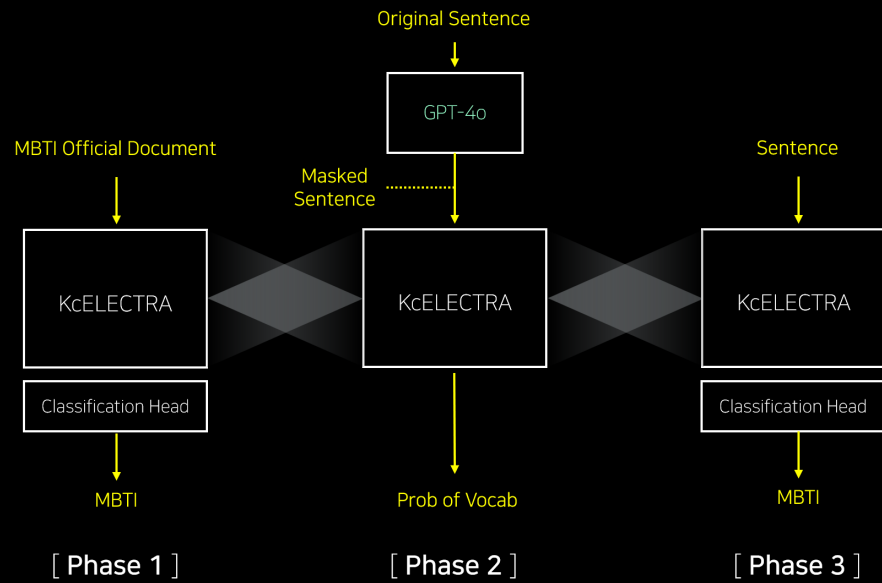
SN



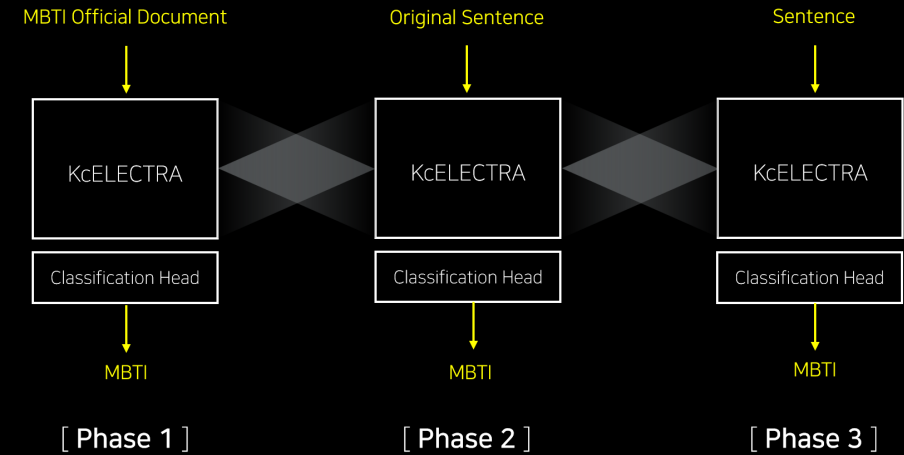
TF



Comparison



MLM Fine Tuning



Fine Tuning

Comparison

	MLM Fine Tuning	Fine Tuning
El Accuracy	80.86	81.31
SN Accuracy	82.31	82.82
TF Accuracy	81.88	81.78
JP Accuracy	81.50	79.29
Test Loss	1.0580	1.2715
Test Accuracy	81.64	81.30

Truth

MFT

FT



김민규

너가 기존에 masking한 방식은 gpt한테 부탁한거라 그랬지??!
오케이오케이 그런 식으로 갈 것 같아 고마워!

ISFJ

ISFP

ESFP



성재영

나 1시넘어서 들어갈것 같은데 너 피곤하면 내일 아침에 할까?

INTP

INTP

INTP



이영주

다 했는데 MBTI 16개별로 나눠서 업로드 할까
아니면 다 합쳐서 업로드 할까? 컴비 강의실만가서 올릴게

ESTP

ESTJ

ENFP



이준성

더깔끔하게 할거면 솔직히 우리 그 마스킹한 데이터셋 랜덤 마스킹 해서 한번
더 돌려보는것도 나쁘지않은디 어떻게생각해 데이터팀한테 부탁해놓을까

INFP

INTP

ENFP



이현준

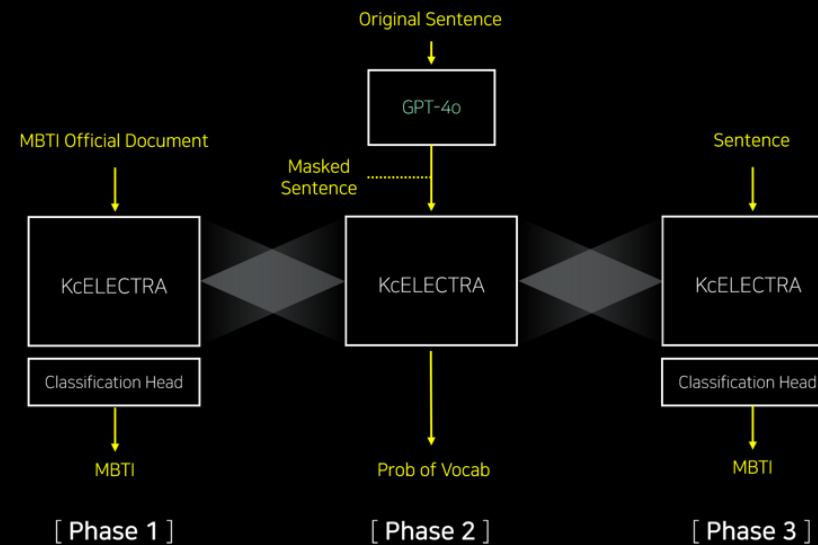
굳굳 두 문장이 딱 적당한 길이인 것 같아
이거 두 문장 번역과정에서 한국어로는 세 문장 돼도 괜찮아!

ISFJ

INFJ

INFP

MLM Fine Tuning



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