ERNIE 2.0: A CONTINUAL PRE-TRAINING FRAMEWORK FOR LANGUAGE UNDERSTANDING

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https://ratsgo.github.io

Motivation

대부분의 pretrained model이 co-occurrence로만 학습 named entity, sentence order, semantic similarity 등도 중요

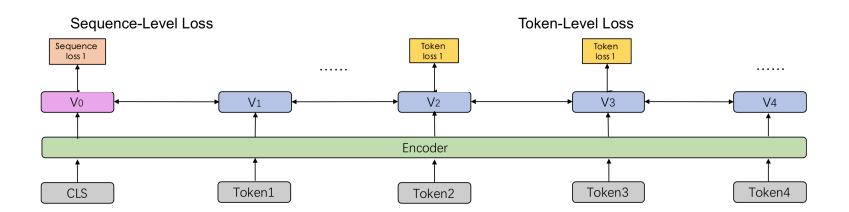
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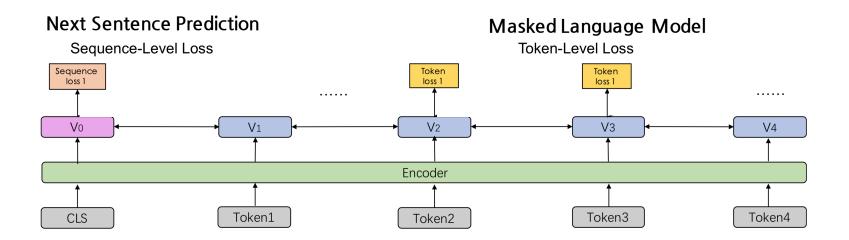
- → Knowledge Integration (ERNIE)
- → Continual Learning (ERNIE2.0)

Continual Learning

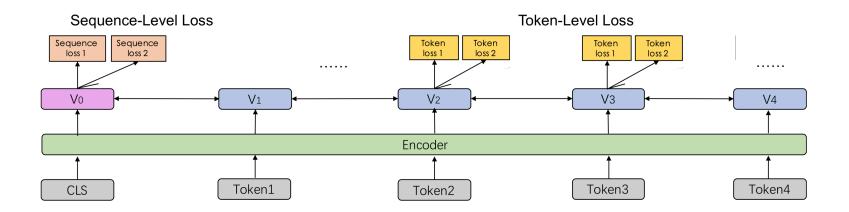
Continual Learning



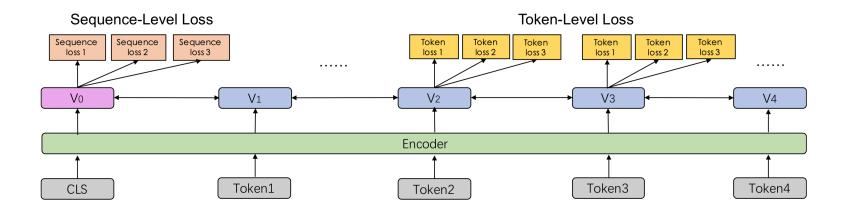
Continual Learning



Continual Learning

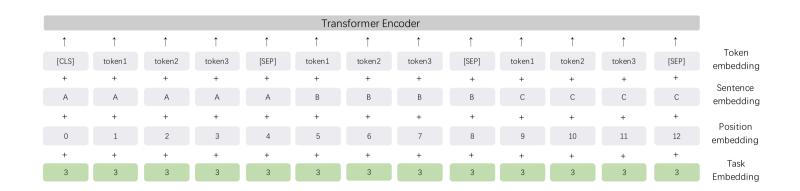


Continual Learning



Task Embedding

Task별 임베딩을 입력 레이어에서 더함

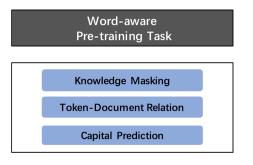


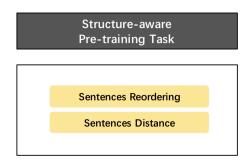
Pre-train Tasks

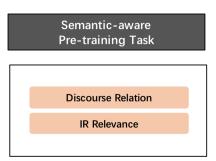
Word-aware Pre-training task: lexical information

Structure-aware Pre-training task: syntactic information

Semantic-aware Pre-training task: semantic information







Word-aware Pre-training task

Knowledge Masking Task

개체(named entity), 구(phrase) 단위 마스킹 (ERNIE1.0) continual learning의 최초 모델에 적용

Capitalization Prediction Task

해당 단어가 대문자인지 소문자인지 예측

Token-Document Relation Prediction task

해당 토큰이 원본 문서의 다른 segment에 나타나는지 예측 키워드(동일 문서에 자주 나타나는 단어) 파악 가능

• Structure-aware Pre-training task

Sentence Reordering Task

원래 세그먼트를 1~m개로 나누고 랜덤 셔플한 뒤 순서 맟추기 문장 간 관계를 파악하게 됨

• Structure-aware Pre-training task

Sentence Reordering Task

원래 세그먼트를 1~m개로 나누고 랜덤 셔플한 뒤 순서 맟추기 문장 간 관계를 파악하게 됨

→ m=2이라면 1! + 2! = 3, 즉 3범주 분류 문제로 정의

[100]: 세그먼트를 하나로 나눈 것

[0 1 0]: 두 세그먼트 가운데 현재 예측 대상 세그먼트가 첫번째

Sentence Distance task

3범주 분류 문제

0=동일 문서 앞뒤 문장

1=앞뒤는 아니지만 동일 문서

2=다른 문서에서 뽑힌 두 문장

• Semantic-aware Pre-training task

Discource Relation Task

두 문장이 의미적으로, 비유적으로 유사한지 여부 예측

Sentence Distance task

앞 문장을 쿼리/뒤 문장을 검색 문서 제목으로 간주, 3범주 분류 문제 0=강한 관련성, 쿼리를 실제 검색했을 때 해당 문서 제목을 클릭 1=약한 관련성, 해당 쿼리를 던지면 검색은 되지만 미클릭 2=관련성 제로

• Pre-train Data

English

Wikipedia, BookCorpus, Reddit, Discovery

Chinese

Encyclopedia, news, dialogue, 검색 등 (바이두 데이터)

Pre-train Settings

```
Layer: 12(base), 24(large)
Head: 12(base), 16(large)
```

Dimension : 768(base), 1024(large) Adam optimizer : $\beta_1 = 0.9$, $\beta_2 = 0.98$

Batch size: 393216 tokens

Learning rate: 5e-5(English), 1.28e-4(Chinese)

Warm-up steps: 4000 (task 달라질 때마다)

• Results (English)

	BASE model		LARGE model					
Task(Metrics)	Test			Dev		Test		
	BERT	ERNIE 2.0	BERT	XLNet	ERNIE 2.0	BERT	ERNIE 2.0	
CoLA (Matthew Corr.)	52.1	55.2	60.6	63.6	65.4	60.5	63.5	
SST-2 (Accuracy)	93.5	95.0	93.2	95.6	96.0	94.9	95.6	
MRPC (Accurary/F1)	84.8/88.9	86.1/89.9	88.0/-	89.2/-	89.7/-	85.4/89.3	87.4/90.2	
STS-B (Pearson Corr./Spearman Corr.)	87.1/85.8	87.6/86.5	90.0/-	91.8/-	92.3/-	87.6/86.5	91.2/90.6	
QQP (Accuracy/F1)	89.2/71.2	89.8/73.2	91.3/-	91.8/-	92.5/-	89.3/72.1	90.1/73.8	
MNLI-m/mm (Accuracy)	84.6/83.4	86.1/85.5	86.6/-	89.8/-	89.1/-	86.7/85.9	88.7/88.8	
QNLI (Accuracy)	90.5	92.9	92.3	93.9	94.3	92.7	94.6	
RTE (Accuracy)	66.4	74.8	70.4	83.8	85.2	70.1	80.2	
WNLI (Accuracy)	65.1	65.1	-	-	-	65.1	67.8	
AX(Matthew Corr.)	34.2	37.4	_	-	-	39.6	48.0	
Score	78.3	80.6	-	-	-	80.5	83.6	

• Results (Chinese)

Task	Metrics	BERT_{BASE}		ERNIE 1.0_{BASE}		ERNIE 2.0_{BASE}		ERNIE 2.0_{LARGE}	
		Dev	Test	Dev	Test	Dev	Test	Dev	Test
CMRC 2018	EM/F1	66.3/85.9	-	65.1/85.1	-	69.1/88.6	-	71.5/89.9	-
DRCD	EM/F1	85.7/91.6	84.9/90.9	84.6/90.9	84.0/90.5	88.5/93.8	88.0/93.4	89.7/94.7	89.0/94.2
DuReader	EM/F1	59.5/73.1	-	57.9/72.1	-	61.3/74.9	-	64.2/77.3	-
MSRA-NER	F1	94.0	92.6	95.0	93.8	95.2	93.8	96.3	95.0
XNLI	Accuracy	78.1	77.2	79.9	78.4	81.2	79.7	82.6	81.0
ChnSentiCorp	Accuracy	94.6	94.3	95.2	95.4	95.7	95.5	96.1	95.8
LCQMC	Accuracy	88.8	87.0	89.7	87.4	90.9	87.9	90.9	87.9
BQ Corpus	Accuracy	85.9	84.8	86.1	84.8	86.4	85.0	86.5	85.2
NLPCC-DBQA	MRR/F1	94.7/80.7	94.6/80.8	95.0/82.3	95.1/82.7	95.7/84.7	95.7/85.3	95.9/85.3	95.8/85.8