Applied Deep Learning, Spring 2022 - Homework 2 郭柏志 R09521205

Q1: Data processing

1. Tokenizer:

Tokenize 的目標是把輸入的句子切成一個個 token,每個 token 有完整的語意,一方面可以將沒看過的字詞切成多個 vocabulary 中存在的 token,一方面方便學習 embedding,hugging-face 提供的 Tokenizer 會將 token encode 成 ids,再透過 embedding layer 將 token 轉為帶有詞語意義的詞向量給模型學習。我使用的 pretrained model 為 bert-base-chinese,bert 的 tokenize algorithm 用的是 WordPiece,與 Byte-Pair Encoding (BPE)類似,會先標準化處理字串,如 unicode 轉換、切割標點符號等,再切割成數個 subword 建立基礎詞彙表,中文句子的 subword 單位為單一中文字,因此基礎詞彙表包含了所有中文字以及非中文字元,接著根據基礎詞彙表,找出能夠最大化訓練集似然度的兩個基礎 A,B 詞彙合併,使得合併的 A,B 有最大的 $\frac{P(AB)}{P(A)P(B)}$ 。

2. Answer Span:

- (a) hugging-face 提供的 Tokenizer 若設定 return_offsets_mapping 為 True Tokenized data 會包含每個 token 對應 question 或 context 的(char start position, char end position), 逐個 token 檢查其 char start position, char end position 便能知道哪個 answer span 的起點與終點對應到哪兩個 token。
- (b) 一個 QA pair 可能因為字數過長被切成多段,預測時將所有分段的 data 的 start_logit, end_logit 存起來,同一個 QA pair 只留下 start_logit + end_logit 最高的前 n_best_size 組,剔除 start_position 大於 end_position 的組合,挑選分數最高者做為答案。

Q2: Modeling with BERTs and their variants

1.

a. Model: bert-base-chinese

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```

b. Performance

Context selection accuracy: 0.973 Question Answering EM (kaggle): 0.744

c. Loss function Cross entropy loss

d. Optimization algorithm, Learning rate, Batch size AdamW with leaning rate = 0.00003, per_device_train_batch_size = 4, gradient accumulation steps = 4

a. Model: hfl/chinese-roberta-wwm-ext

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"transformers_version": "4.17.0",
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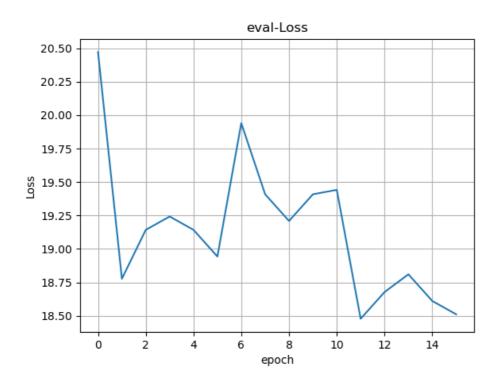
b. Performance

Context selection accuracy: 0.973 Question Answering EM (kaggle): 0.761

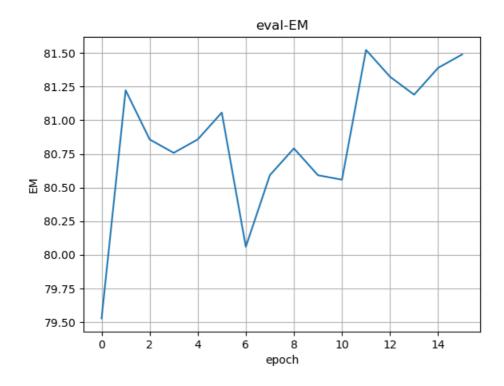
c. Loss function Cross entropy loss

d. Optimization algorithm, Learning rate, Batch size AdamW with leaning rate = 0.00003,
 per_device_train_batch_size = 4,
 gradient accumulation steps = 4

Q3: Curves a. Learning curve of loss



b. Learning curve of EM



O4: Pretrained vs Not Pretrained

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"attention_probs_dropout_prob": 0.1,
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"initializer_range": 0.02,
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```

減少 hidden_dim: $784 \rightarrow 128$, attention_head: $12 \rightarrow 8$, batch_size: $4 \rightarrow 1$, 只從新訓練 QA 任務,local 端得到 EM = 0.066, 由於 Trainsformer 架構龐大,需要大量訓練資料與時間才能訓練得起來,此任務的訓練資料與時間都與 pretrained 相差甚遠,因此若要從 0 開始訓練 bert model 會非常困難。