CS 4372 ASSIGNMENT 4

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Number of free late days used: 0

Note: You are allowed a <u>total</u> of 4 free late days for the <u>entire semester</u>. You can use at most 2 for each assignment.

After that, there will be a penalty of 10% for each late day.

The Use of Transformers in Chat Bots

The input text (The Return of Sherlock Holmes, in our case) is tokenized and then fed into the transformer model (AlbertForQuestionAnswering). Each token is then encoded into embeddings, so that the model can understand the position and context of each token. These features of a token are what allows AlbertForQuestionAnswering to be a powerful model for question-answering. Transformers can process the input to find contextual links between the question and relevant parts of the text, and can use its fine-tuned understanding to predict what the answer is likely to be. The most probable answer is outputted with a confidence score.

The AlbertForQuestionAnswering Model

AlbertForQuestionAnswering has a transformer encoder architecture. It has multiple layers of transformer encoders that use self-attention to understand relationships between tokens. AlbertForQuestionAnswering also implements parameter sharing, which allows the model to reduce memory usage. This is done by implementing the same parameters for all layers, instead of different parameters for each layer, thus giving us a more compact model. The overall structure of AlbertForQuestionAnswering includes input embeddings and stacked transformer encoded layers, which produces probable outputs. A core component of the AlbertForQuestionAnswering model, and other models as well, is self-attention.

Hyper-parameter Tuning

| Iteration | hyper-params | Training loss | Validation loss | Score (on book q) | Answers to Qs | | |
|-----------|---|--|-------------------|--|---|-----------------------|--|
| | 1 NA | NA | NA | Q1: 0.7764191627502441 Q2: 0.01717839017510414 Q3: 0.01087028719484806 | Q1: 1894 Q2: Sherlock Holmes Q3: THE ADVENTURE OF THE EMPTY HOUSE | Using ALBERT untouche | |
| | learning rate: 2e-5 train batch size: 16 eval batch size: 16 training epochs: 5 weight decay: 0.01 | epoch 3: 2.681100 epoch 4: 1.135000 | epoch 4: 1.550689 | | Q1: 1894 Q2: Sherlock Holmes Q3: Sherlock Holmes | | |
| | learning rate: 2e-5 train batch size: 16 eval batch size: 16 training epochs: 4 weight decay: 0.01 | epoch 3: 2.585500 | epoch 3: 1.570670 | Q1: 0.9458447694778442 Q2: 0.720957338809967 Q3: 0.7049626708030701 | Q1: 1894 Q2: young Adair Q3: the Park Lane Mystery | | |
| | learning rate: 2e-5 train batch size: 16 eval batch size: 16 training epochs: 4 weight decay: 0.10 | epoch 3: 2.599900 | epoch 3: 1.580346 | Q1: 0.8227441310882568 Q2: 0.6674026250839233 Q3: 0.44598376750946045 | Q1: 1894 Q2: Lady Maynooth Q3: the Park Lane Mystery | | |
| | learning rate: 2e-5 train batch size: 16 eval batch size: 16 training epochs: 4 weight decay: 0.001 | epoch 2: 2.596300 epoch 3: 2.596300 | epoch 3: 1.584609 | Q1: 0.8897749185562134 Q2: 0.4277809262275696 Q3: 0.8366639614105225 | Q1: 1894 Q2: Lady Maynooth Q3: the Park Lane Mystery | | |
| | learning rate: 2e-3 train batch size: 16 eval batch size: 16 training epochs: 4 weight decay: 0.01 | epoch 2: 5.967700 epoch 3: 5.967700 | epoch 3: 5.950644 | Q1: 3.857876072288491e-05 | Q3: wound which must have caused instantaneou | | |
| | learning rate: 2e-7 train batch size: 16 eval batch size: 16 training epochs: 4 weight decay: 0.01 | epoch 3: 5.862200 | epoch 3: 5.766145 | Q1: 0.0001003671932267025 Q2: 9.018548007588834e-05 Q3: 9.10957096493803e-05 | Q2: Street end of | | |

Results

The hyper-parameters used in the second and third iteration seem to produce the best results from our trials. We believe the third iteration to be the most accurate in the answers it produced. The third iteration also showed a consistently high confidence score.

Result Analysis

Using the AlbertForQuestionAnswering model untouched, the only high scoring answer was to the first question, while the next two had confidence scores of only 0.01. This showed us that we needed to tune the AlbertForQuestionAnswering model to our data in order to get correct answers and higher confidence scores. The second and third iterations showed much improvement, however the second iteration had some inconsistencies. Although it scored very high for the first question (0.98), it was not able to assess the next question as well. We see the third iteration gave us consistently higher confidence. It is possible, however, that the model consistently scored worse on the second question because not enough data was given. If we chose a wider sample of the text to use to train the model, there is a chance that all questions could have scored higher.

We conclude that the third iteration's hyper parameters were the most useful in obtaining accurate results from our model.

Assumptions

Instead of using the entire book, we used only a portion to save time and space. Therefore, the model considers only a portion of the text to answer questions.