TaskA_LSTM

November 27, 2024

```
[37]: import nltk
      import keras
      import pandas as pd
      import numpy as np
      from keras.layers import Dense
      from keras.layers import LSTM
      from keras.layers import Embedding
      from keras.layers import TextVectorization
      from keras.models import Sequential
      from keras.preprocessing.sequence import pad_sequences
      from sklearn.model_selection import KFold
      from sklearn.model_selection import train_test_split
 [2]: """
      Download dataset SubtaskA.jsonl from
      https://github.com/mbzuai-nlp/M4GT-Bench.\\
      DATA_PATH = "C:/Users/Admin/Downloads/SubtaskA.jsonl"
      # initialize dataframe
      df = pd.read_json(DATA_PATH, lines=True)
[14]: print(df.source.value_counts())
      print()
      print(df.model.value_counts())
     source
     wikihow
                  36556
     reddit
                  33999
     arxiv
                  33998
     wikipedia
                  31365
     peerread
                  16891
     Name: count, dtype: int64
     model
     human
                65177
     chatGPT
                16892
                14344
     gpt4
```

```
davinci
                 14340
     bloomz
                 14332
     dolly
                 14046
     cohere
                 13678
     Name: count, dtype: int64
[22]: print(df[df.label == 0].model.value_counts())
      print()
      print(df[df.label == 1].model.value_counts())
     model
     human
               65177
     Name: count, dtype: int64
     model
     chatGPT
                 16892
     gpt4
                 14344
                 14340
     davinci
     bloomz
                 14332
     dolly
                 14046
                 13678
     cohere
     Name: count, dtype: int64
[27]: df[['text', 'label']]
[27]:
                                                                    label
                                                              text
      0
              We consider a system of many polymers in solut...
                                                                      1
              We present a catalog of 66 YSOs in the Serpens...
      1
                                                                      1
      2
              Spectroscopic Observations of the Intermediate...
                                                                      1
      3
              We present a new class of stochastic Lie group...
                                                                      1
      4
              ALMA as the ideal probe of the solar chromosph...
                                                                      1
      152804
               The main results presented in this dissertati...
                                                                      0
      152805
               Fine-grained sketch-based image retrieval (FG...
                                                                      0
      152806
               We present the derivation of the NNLO two-par...
                                                                      0
      152807
               The principle of optimism in the face of unce...
                                                                      0
               We consider the setting of prediction with ex...
      152808
                                                                      0
      [152809 rows x 2 columns]
[15]: """
      Pre-process dataframe.
      MAX_VOCAB = 10_000
      MAX LENGTH = 200
      # init text vectorizer
      vectorize_layer = TextVectorization(
```

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max_tokens=MAX_VOCAB,
          standardize='lower_and_strip_punctuation',
          split='whitespace',
          ngrams=None,
          output_mode='int',
          output_sequence_length=MAX_LENGTH,
          pad_to_max_tokens=False,
          vocabulary=None,
          idf_weights=None,
          sparse=False,
          ragged=False,
          encoding='utf-8',
          name=None,
      )
      # create vocabulary
      vectorize_layer.adapt(df['text'])
      vocab = vectorize_layer.get_vocabulary()
[34]: # vectorize text data (in subsets for memory constraints)
      y = df['label']
      subset_size = df.shape[0] // 100
      for i in range(0, df.shape[0], subset_size):
          subset = df['text'][i : i + subset size]
          X.append(vectorize_layer(subset))
      X = np.vstack(X)
      print(X.shape, y.shape)
     (152809, 200) (152809,)
[39]: """
      LSTM model generator.
      EMBEDDING_DIM = 128
      N \text{ HIDDEN} = 100
      OPTIMIZER = 'adam'
      N CLASSES = 2
      def get_model(model_path=None):
          if model_path:
              # load existing model
              model = keras.models.load_model(model_path)
          else:
              # create new model
              model = Sequential()
```

```
embeddings = Embedding(
    input_dim=MAX_VOCAB,
    output_dim=EMBEDDING_DIM,
)

model.add(embeddings)
model.add(LSTM(N_HIDDEN, return_sequences=False))
model.add(Dense(N_CLASSES)) # , activation='softmax'
model.compile(
    loss=keras.losses.SparseCategoricalCrossentropy(from_logits=True),
    optimizer=OPTIMIZER,
    metrics=['accuracy']
)
return model
```

```
[41]: """
      Train and evaluate model.
      # create new model
      model = get_model()
      # create data splits
      X_train, X_test, y_train, y_test = train_test_split(
          Х,
          у,
          test_size=0.15,
          random_state=777,
      # train the model
      model.fit(
         X_train,
          y_train,
          epochs=5,
          batch_size=64
      )
      # final evaluation of the model
      scores = model.evaluate(
          X_test,
          y_test,
          verbose=0
      accuracy = scores[1]
      # report results
      print("Accuracy: %.2f%%" % (accuracy * 100))
```

save model model.save("models/taskA_lstm.keras")

Epoch 1/5

2030/2030 370s 181ms/step - accuracy: 0.7507 - loss: 0.5110

Epoch 2/5

2030/2030 367s 181ms/step - accuracy: 0.8752 - loss: 0.3056

Epoch 3/5

2030/2030 368s 181ms/step - accuracy: 0.9035 - loss: 0.2410

Epoch 4/5

2030/2030 366s 180ms/step - accuracy: 0.9293 - loss: 0.1850

Epoch 5/5

2030/2030 363s 179ms/step - accuracy: 0.9453 - loss: 0.1446

Accuracy: 88.61%

[10]: