Assignment10_3

November 7, 2021

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[10]: import tensorflow.compat.v1 as tf
      tf.disable_v2_behavior()
      from keras.preprocessing.text import Tokenizer
      from keras.preprocessing.sequence import pad_sequences
      import numpy as np
      import matplotlib.pyplot as plt
      from pathlib import Path
      from keras.models import Sequential
      from keras.layers import Embedding, Flatten, Dense
      import os
      from contextlib import redirect_stdout
      import time
      start_time = time.time()
      from keras.layers import LSTM
      from keras.datasets import imdb
      from keras.preprocessing import sequence
 [2]: imdb_dir = Path('/home/jovyan/dsc650/data/external/imdb/aclImdb/')
      test_dir = os.path.join(imdb_dir, 'test')
      train_dir = os.path.join(imdb_dir, 'train')
      results_dir = Path('results').joinpath('model_1')
      results_dir.mkdir(parents=True, exist_ok=True)
 [3]: max_features = 10000
     maxlen = 500
      batch_size = 32
      max_words = 1000
      training_samples = 200
      validation_samples = 10000
 [4]: labels = []
      texts = []
      for label_type in ['neg', 'pos']:
          dir_name = os.path.join(test_dir, label_type)
          for fname in sorted(os.listdir(dir_name)):
              if fname[-4:] == '.txt':
                  f = open(os.path.join(dir_name, fname), encoding="utf8")
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texts.append(f.read())
                 f.close()
                 if label_type == 'neg':
                     labels.append(0)
                 else:
                     labels.append(1)
[5]: tokenizer = Tokenizer(num_words=max_words)
     tokenizer.fit_on_texts(texts)
     sequences = tokenizer.texts_to_sequences(texts)
     print('Loading data... ')
     word_index = tokenizer.word_index
     print('Found %s unique tokens.' % len(word_index))
     data = pad_sequences(sequences, maxlen=maxlen)
     labels = np.asarray(labels)
     print('Shape of data tensor:', data.shape)
     print('Shape of label tensor:', labels.shape)
     indices = np.arange(data.shape[0])
     np.random.shuffle(indices)
     data = data[indices]
     labels = labels[indices]
    Loading data...
    Found 87393 unique tokens.
    Shape of data tensor: (25000, 500)
    Shape of label tensor: (25000,)
[6]: | #x_train
     input_train = data[:training_samples]
     y_train = labels[:training_samples]
     input_test = data[training_samples: training_samples + validation_samples]
     #y_val
     y_test = labels[training_samples: training_samples + validation_samples]
     print('input_train shape:', input_train.shape)
     print('input_test shape:', input_test.shape)
    input_train shape: (200, 500)
    input_test shape: (10000, 500)
[7]: # from page 205 - Listing 6.27
     model = Sequential()
     model.add(Embedding(max_features, 32))
     model.add(LSTM(32))
     model.add(Dense(1, activation='sigmoid'))
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model.compile(optimizer='rmsprop', loss='binary_crossentropy', metrics=['acc'])

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history=model.fit(input_train, y_train, epochs=10,__
 ⇒batch_size=32,validation_data=(input_test, y_test))
result_model_file = results_dir.joinpath('pre_trained_glove_model_LSTM.h5')
model.save_weights(result_model_file)
WARNING:tensorflow:From /opt/conda/lib/python3.8/site-
packages/tensorflow/python/keras/initializers/initializers_v1.py:58: calling
RandomUniform.__init__ (from tensorflow.python.ops.init_ops) with dtype is
deprecated and will be removed in a future version.
Instructions for updating:
Call initializer instance with the dtype argument instead of passing it to the
constructor
Train on 200 samples, validate on 10000 samples
Epoch 1/10
/opt/conda/lib/python3.8/site-
packages/tensorflow/python/keras/engine/training.py:2325: UserWarning:
`Model.state updates` will be removed in a future version. This property should
not be used in TensorFlow 2.0, as `updates` are applied automatically.
 warnings.warn('`Model.state updates` will be removed in a future version. '
200/200 [============ ] - 23s 114ms/sample - loss: 0.6941 -
acc: 0.5000 - val_loss: 0.6921 - val_acc: 0.5051
Epoch 2/10
acc: 0.5450 - val_loss: 0.6907 - val_acc: 0.5545
acc: 0.7900 - val_loss: 0.6876 - val_acc: 0.6027
Epoch 4/10
acc: 0.8400 - val_loss: 0.6793 - val_acc: 0.6118
Epoch 5/10
acc: 0.8250 - val_loss: 0.6476 - val_acc: 0.6766
Epoch 6/10
acc: 0.8700 - val loss: 0.6322 - val acc: 0.6365
Epoch 7/10
200/200 [============== ] - 21s 106ms/sample - loss: 0.4548 -
acc: 0.8700 - val_loss: 0.6299 - val_acc: 0.6561
Epoch 8/10
acc: 0.9400 - val_loss: 1.0575 - val_acc: 0.5094
Epoch 9/10
200/200 [============= ] - 21s 106ms/sample - loss: 0.3532 -
acc: 0.9200 - val_loss: 0.5785 - val_acc: 0.6949
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Epoch 10/10
    acc: 0.9050 - val_loss: 0.5757 - val_acc: 0.6961
[8]: # Save the summary to file
    summary_file = results_dir.joinpath('Assignment_10.3_ModelSummary.txt')
    with open(summary_file, 'w') as f:
        with redirect_stdout(f):
           model.summary()
[9]: # Plots
    acc = history.history['acc']
    val_acc = history.history['val_acc']
    loss = history.history['loss']
    val loss = history.history['val loss']
    epochs = range(1, len(acc) + 1)
    plt.plot(epochs, acc, 'bo', label='Training acc')
    plt.plot(epochs, val_acc, 'b', label='Validation acc')
    plt.title('Training and validation accuracy')
    plt.legend()
    plt.figure()
    plt.plot(epochs, loss, 'bo', label='Training loss')
    plt.plot(epochs, val_loss, 'b', label='Validation loss')
    plt.title('Training and validation loss')
    plt.legend()
    img_file = results_dir.joinpath('Assignment_10.3_Model Accuracy Validation.png')
    plt.savefig(img_file)
    plt.show()
```



