dl-pract-2b

April 28, 2024

```
[30]: from tensorflow.keras.datasets import imdb
[31]: (x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=10000)
     Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-
     datasets/imdb.npz
     [32]: print("Train Shape:",x_train.shape)
     print("Test Shape :",x_test.shape)
     Train Shape : (25000,)
     Test Shape : (25000,)
[33]: print("y_train shape :",y_train.shape)
     print("y_test shape :",y_test.shape)
     y_train shape : (25000,)
     y_test shape : (25000,)
[34]: print(x_train[1])
     [1, 194, 1153, 194, 8255, 78, 228, 5, 6, 1463, 4369, 5012, 134, 26, 4, 715, 8,
     118, 1634, 14, 394, 20, 13, 119, 954, 189, 102, 5, 207, 110, 3103, 21, 14, 69,
     188, 8, 30, 23, 7, 4, 249, 126, 93, 4, 114, 9, 2300, 1523, 5, 647, 4, 116, 9,
     35, 8163, 4, 229, 9, 340, 1322, 4, 118, 9, 4, 130, 4901, 19, 4, 1002, 5, 89, 29,
     952, 46, 37, 4, 455, 9, 45, 43, 38, 1543, 1905, 398, 4, 1649, 26, 6853, 5, 163,
     11, 3215, 2, 4, 1153, 9, 194, 775, 7, 8255, 2, 349, 2637, 148, 605, 2, 8003, 15,
     123, 125, 68, 2, 6853, 15, 349, 165, 4362, 98, 5, 4, 228, 9, 43, 2, 1157, 15,
     299, 120, 5, 120, 174, 11, 220, 175, 136, 50, 9, 4373, 228, 8255, 5, 2, 656,
     245, 2350, 5, 4, 9837, 131, 152, 491, 18, 2, 32, 7464, 1212, 14, 9, 6, 371, 78,
     22, 625, 64, 1382, 9, 8, 168, 145, 23, 4, 1690, 15, 16, 4, 1355, 5, 28, 6, 52,
     154, 462, 33, 89, 78, 285, 16, 145, 95]
[35]: print(y_train[1])
```

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[36]: vocab=imdb.get_word_index()
      print(vocab['the'])
     Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-
     datasets/imdb_word_index.json
     1641221/1641221 [============ ] - Os Ous/step
[37]: class_names=['Negative', 'Positive']
[38]: reverse_index = dict([(value, key) for (key, value) in vocab.items()])
[42]: def decode(review):
           text=""
            for i in review:
            text=text+reverse index[i]
            text=text+" "
            return text
[43]: decode(x_train[1])
[43]: "the thought solid thought senator do making to is spot nomination assumed while
     he of jack in where picked as getting on was did hands fact characters to always
      life thrillers not as me can't in at are br of sure your way of little it
      strongly random to view of love it so principles of guy it used producer of
      where it of here icon film of outside to don't all unique some like of direction
      it if out her imagination below keep of queen he diverse to makes this stretch
      and of solid it thought begins br senator and budget worthwhile though ok and
      awaiting for ever better were and diverse for budget look kicked any to of
     making it out and follows for effects show to show cast this family us scenes
     more it severe making senator to and finds tv tend to of emerged these thing
      wants but and an beckinsale cult as it is video do you david see scenery it in
      few those are of ship for with of wild to one is very work dark they don't do
      dvd with those them "
[44]: def showlen():
           print("Length of first training sample: ",len(x_train[0]))
           print("Length of second training sample: ",len(x_train[1]))
           print("Length of first test sample: ",len(x_test[0]))
           print("Length of second test sample: ",len(x_test[1]))
      showlen()
     Length of first training sample: 218
     Length of second training sample: 189
     Length of first test sample: 68
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Length of second test sample: 260

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[45]: from tensorflow.keras.preprocessing.sequence import pad_sequences
[46]: x train=pad sequences(x train, value=vocab['the'], padding='post', maxlen=256)
     x_test=pad_sequences(x_test, value=vocab['the'], padding='post', maxlen=256)
[47]: showlen()
    Length of first training sample:
    Length of second training sample:
    Length of first test sample:
    Length of second test sample: 256
[48]: decode(x_train[1])
[48]: "the thought solid thought senator do making to is spot nomination assumed while
     he of jack in where picked as getting on was did hands fact characters to always
     life thrillers not as me can't in at are br of sure your way of little it
     strongly random to view of love it so principles of guy it used producer of
     where it of here icon film of outside to don't all unique some like of direction
     it if out her imagination below keep of queen he diverse to makes this stretch
     and of solid it thought begins br senator and budget worthwhile though ok and
     awaiting for ever better were and diverse for budget look kicked any to of
    making it out and follows for effects show to show cast this family us scenes
    more it severe making senator to and finds tv tend to of emerged these thing
     wants but and an beckinsale cult as it is video do you david see scenery it in
     few those are of ship for with of wild to one is very work dark they don't do
     the the the the the the the the the "
[49]: from tensorflow.keras.models import Sequential
     from tensorflow.keras.layers import Dense, Embedding, GlobalAveragePooling1D
[51]: model=Sequential()
     model.add(Embedding(10000,16))
     model.add(GlobalAveragePooling1D())
     model.add(Dense(16,activation='relu'))
     model.add(Dense(1,activation='sigmoid'))
     model.compile(optimizer='adam', loss='binary_crossentropy',metrics=['accuracy'])
     model.summary()
    Model: "sequential"
     Layer (type)
                             Output Shape
                                                    Param #
    ______
```

160000

(None, None, 16)

embedding (Embedding)

```
global_average_pooling1d ( (None, 16)
     GlobalAveragePooling1D)
     dense (Dense)
                             (None, 16)
                                                  272
     dense 1 (Dense)
                            (None, 1)
                                                  17
    ______
    Total params: 160289 (626.13 KB)
    Trainable params: 160289 (626.13 KB)
    Non-trainable params: 0 (0.00 Byte)
[52]: model.fit(x_train, y_train, epochs=4, batch_size=128,__
      ⇔verbose=1,validation_data=(x_test, y_test))
    Epoch 1/4
    196/196 [============== ] - 3s 13ms/step - loss: 0.6686 -
    accuracy: 0.6584 - val_loss: 0.6091 - val_accuracy: 0.7955
    196/196 [============ ] - 2s 10ms/step - loss: 0.4926 -
    accuracy: 0.8340 - val_loss: 0.4098 - val_accuracy: 0.8468
    accuracy: 0.8806 - val loss: 0.3299 - val accuracy: 0.8686
    Epoch 4/4
    accuracy: 0.8985 - val_loss: 0.2999 - val_accuracy: 0.8774
[52]: <keras.src.callbacks.History at 0x7a72a8cd18a0>
[53]: x_{test}[10]
[53]: array([
           1, 1581,
                      34, 7908, 5082,
                                     23,
                                          6, 1374, 1120,
                                                          7, 107,
                  2, 1496,
                                                   7,
           349,
                           11, 5116,
                                     18,
                                         397, 3767,
                                                          4, 107,
            84, 6763, 56,
                           68, 456, 1402,
                                          2,
                                               39,
                                                     4, 1374,
                                                               9,
                           55, 4412, 212,
            35, 204,
                      5,
                                         193,
                                               23,
                                                     4, 326,
                                                              45,
             6, 1109,
                      8, 1738,
                                 2,
                                     15,
                                          29, 199, 1040,
                                                         5, 2684,
                 14, 1403, 212, 1528,
                                     10,
                                          10, 2160,
                                                     2,
            11,
                                                          9,
           452,
                 37,
                            4, 598, 425,
                                          5,
                                               45, 4394, 138,
                       2,
                                                               59,
           214, 467,
                       4, 2391,
                                7, 1738,
                                          2,
                                               19,
                                                    41, 2455, 3028,
             5, 6866, 1489,
                                     18, 101, 1403,
                                                     2, 1514, 5257,
                           90,
                               180,
                                     47, 2586,
             9.
                  4, 564, 871, 322,
                                              27, 274, 326,
                                          87, 162, 2133, 60, 3256,
             9, 150, 112,
                            2,
                               17,
                                     6,
                  4, 7999, 123,
            23,
                                8,
                                     11,
                                          2,
                                               29, 144,
                                                         30, 2961,
           1346,
                  2, 214,
                            4, 326,
                                     7,
                                          2, 1496,
                                                     8, 3767, 533,
                       2, 6229,
                                         7, 265, 285, 5, 233,
             7, 134,
                                10,
                                    10,
            70, 593, 54, 564, 4124,
                                    2, 1625, 27, 1546,
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                                                                 1,
                        1], dtype=int32)
             1,
                   1,
[54]: y_test[10]
[54]: 1
[55]: import numpy as np
     predicted_value=model.predict(np.expand_dims(x_test[10], 0))
     print(predicted_value)
     if predicted_value>0.5:
       final_value=1
     else:
       final_value=0
     print(final_value)
     print(class_names[final_value])
    1/1 [======] - 0s 78ms/step
    [[0.8281511]]
    1
    Positive
[56]: loss, accuracy = model.evaluate(x_test, y_test)
     print("Loss :",loss)
     print("Accuracy (Test Data) :",accuracy*100)
    accuracy: 0.8774
    Loss: 0.299915611743927
    Accuracy (Test Data): 87.74399757385254
[]:
```

4, 114, 3209,

52,

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5, 45, 1139,

5, 7611, 367,

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143, 3760,

7, 328,

89,

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