

2-imdb

April 23, 2024

```
[14]: from tensorflow.keras.datasets import imdb
```

```
[15]: (train_data, train_label), (test_data, test_label) = imdb.load_data(num_words = 10000)
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz>
17464789/17464789 [=====] - 0s 0us/step

```
[24]: import numpy as np

def vectorize_sequences(sequences, dimensions = 10000):
    results = np.zeros((len(sequences), dimensions))
    for i, sequences in enumerate(sequences):
        results[i, sequences] = 1
    return results

x_train = vectorize_sequences(train_data)
y_train = vectorize_sequences(test_data)
```

```
[25]: y_train = np.asarray(train_label).astype('float32')
y_test = np.asarray(test_label).astype('float32')
```

```
[31]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
```

```
[19]: model = Sequential()
model.add(Dense(16, input_shape=(10000, ), activation = "relu"))
model.add(Dense(16, activation = "relu"))
model.add(Dense(1, activation = "sigmoid"))
```

```
[32]: model.compile(optimizer='adam', loss = 'mse', metrics = ['accuracy'])
```

```
[22]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
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=====
dense (Dense)                (None, 16)                160016

dense_1 (Dense)              (None, 16)                272

dense_2 (Dense)              (None, 1)                 17

=====
Total params: 160,305
Trainable params: 160,305
Non-trainable params: 0
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```

```
[33]: history = model.fit(x_train, y_train, validation_split = 0.3, epochs = 20,
    ↪ verbose = 1, batch_size = 512)
```

```
Epoch 1/20
35/35 [=====] - 2s 39ms/step - loss: 0.0041 - accuracy:
0.9960 - val_loss: 0.1162 - val_accuracy: 0.8652
Epoch 2/20
35/35 [=====] - 1s 35ms/step - loss: 0.0042 - accuracy:
0.9959 - val_loss: 0.1171 - val_accuracy: 0.8648
Epoch 3/20
35/35 [=====] - 1s 32ms/step - loss: 0.0043 - accuracy:
0.9958 - val_loss: 0.1192 - val_accuracy: 0.8636
Epoch 4/20
35/35 [=====] - 1s 31ms/step - loss: 0.0042 - accuracy:
0.9959 - val_loss: 0.1210 - val_accuracy: 0.8619
Epoch 5/20
35/35 [=====] - 1s 31ms/step - loss: 0.0041 - accuracy:
0.9960 - val_loss: 0.1208 - val_accuracy: 0.8628
Epoch 6/20
35/35 [=====] - 1s 42ms/step - loss: 0.0041 - accuracy:
0.9959 - val_loss: 0.1210 - val_accuracy: 0.8619
Epoch 7/20
35/35 [=====] - 2s 47ms/step - loss: 0.0040 - accuracy:
0.9961 - val_loss: 0.1213 - val_accuracy: 0.8621
Epoch 8/20
35/35 [=====] - 1s 31ms/step - loss: 0.0041 - accuracy:
0.9960 - val_loss: 0.1214 - val_accuracy: 0.8620
Epoch 9/20
35/35 [=====] - 1s 32ms/step - loss: 0.0041 - accuracy:
0.9961 - val_loss: 0.1204 - val_accuracy: 0.8655
Epoch 10/20
35/35 [=====] - 1s 32ms/step - loss: 0.0041 - accuracy:
0.9961 - val_loss: 0.1215 - val_accuracy: 0.8633
Epoch 11/20
35/35 [=====] - 1s 37ms/step - loss: 0.0040 - accuracy:
```

0.9961 - val_loss: 0.1224 - val_accuracy: 0.8623
Epoch 12/20
35/35 [=====] - 1s 32ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1223 - val_accuracy: 0.8617
Epoch 13/20
35/35 [=====] - 1s 35ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1223 - val_accuracy: 0.8613
Epoch 14/20
35/35 [=====] - 1s 36ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1223 - val_accuracy: 0.8616
Epoch 15/20
35/35 [=====] - 1s 31ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1223 - val_accuracy: 0.8613
Epoch 16/20
35/35 [=====] - 1s 38ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1224 - val_accuracy: 0.8611
Epoch 17/20
35/35 [=====] - 2s 48ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1224 - val_accuracy: 0.8607
Epoch 18/20
35/35 [=====] - 1s 37ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1224 - val_accuracy: 0.8608
Epoch 19/20
35/35 [=====] - 1s 35ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1224 - val_accuracy: 0.8605
Epoch 20/20
35/35 [=====] - 2s 60ms/step - loss: 0.0039 - accuracy:
0.9961 - val_loss: 0.1224 - val_accuracy: 0.8609