



```

0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 1 194 1153 194 2 78 228 5 6
1463 4369 2 134 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 2 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 2 5 163 11 3215 2
4 1153 9 194 775 7 2 2 349 2637 148 605 2 2
15 123 125 68 2 2 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 2 5 2 656 245 2350 5 4 2 131
152 491 18 2 32 2 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]

```

```

import tensorflow as tf
# Create the model

embedding_vector_length = 32

model = Sequential()

model.add(Embedding(top_words + 1, embedding_vector_length, input_length=max_review_length))
model.add(LSTM(100))
model.add(Dense(1, activation='sigmoid'))
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])

print(model.summary())

```

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 600, 32)	160032
lstm (LSTM)	(None, 100)	53200
dense (Dense)	(None, 1)	101

```

=====
Total params: 213,333
Trainable params: 213,333
Non-trainable params: 0

```

None

```

hist=model.fit(X_train, y_train, epochs=10, batch_size=64,verbose=1, validation_data=(X_test,y_test)

# Final evaluation of the model
scores = model.evaluate(X_test, y_test, verbose=0)

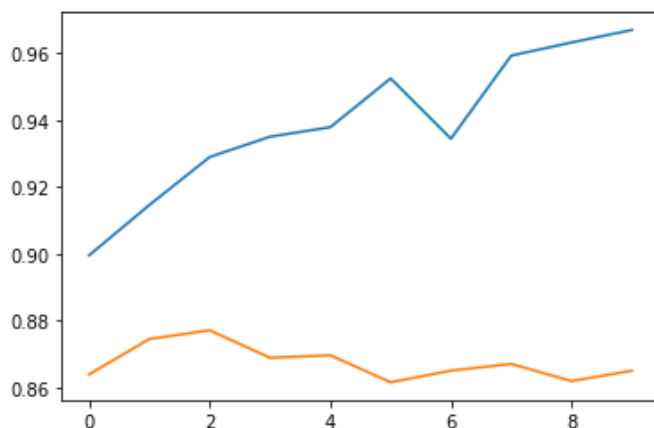
```

```
print("Accuracy: %.2f%%" % (scores[1] 100))
```

```
Epoch 1/10
391/391 [=====] - 17s 43ms/step - loss: 0.2609 - accuracy: 0.8995 -
Epoch 2/10
391/391 [=====] - 18s 46ms/step - loss: 0.2212 - accuracy: 0.9146 -
Epoch 3/10
391/391 [=====] - 17s 44ms/step - loss: 0.1917 - accuracy: 0.9289 -
Epoch 4/10
391/391 [=====] - 17s 43ms/step - loss: 0.1727 - accuracy: 0.9350 -
Epoch 5/10
391/391 [=====] - 18s 46ms/step - loss: 0.1683 - accuracy: 0.9378 -
Epoch 6/10
391/391 [=====] - 17s 43ms/step - loss: 0.1316 - accuracy: 0.9524 -
Epoch 7/10
391/391 [=====] - 18s 47ms/step - loss: 0.1723 - accuracy: 0.9344 -
Epoch 8/10
391/391 [=====] - 18s 46ms/step - loss: 0.1164 - accuracy: 0.9593 -
Epoch 9/10
391/391 [=====] - 28s 71ms/step - loss: 0.1074 - accuracy: 0.9632 -
Epoch 10/10
391/391 [=====] - 18s 47ms/step - loss: 0.0950 - accuracy: 0.9670 -
Accuracy: 86.49%
```

```
import matplotlib.pyplot as plt
y1=hist.history['accuracy']
y2=hist.history['val_accuracy']
plt.plot(y1)
plt.plot(y2)
```

```
[<matplotlib.lines.Line2D at 0x7fd58dbcf9d0>]
```



```
print(X_test[1],y_test[1])
pred=model.predict(X_test)
print(pred)
```

[illegible]

```

0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 1 14 22 3443 6 176 7 2 88 12
2679 23 1310 5 109 943 4 114 9 55 606 5 111 7
4 139 193 273 23 4 172 270 11 2 2 4 2 2801
109 1603 21 4 22 3861 8 6 1193 1330 10 10 4 105
987 35 841 2 19 861 1074 5 1987 2 45 55 221 15
670 2 526 14 1069 4 405 5 2438 7 27 85 108 131
4 2 2 3884 405 9 3523 133 5 50 13 104 51 66
166 14 22 157 9 4 530 239 34 2 2801 45 407 31
7 41 3778 105 21 59 299 12 38 950 5 4521 15 45
629 488 2733 127 6 52 292 17 4 2 185 132 1988 2
1799 488 2693 47 6 392 173 4 2 4378 270 2352 4 1500
7 4 65 55 73 11 346 14 20 9 6 976 2078 7
2 861 2 5 4182 30 3127 2 56 4 841 5 990 692
8 4 1669 398 229 10 10 13 2822 670 2 14 9 31
7 27 111 108 15 2033 19 2 1429 875 551 14 22 9
1193 21 45 4829 5 45 252 8 2 6 565 921 3639 39
4 529 48 25 181 8 67 35 1732 22 49 238 60 135
1162 14 9 290 4 58 10 10 472 45 55 878 8 169
11 374 2 25 203 28 8 818 12 125 4 3077] 1
782/782 [=====] - 7s 9ms/step
[[0.00619938]
[0.999757 ]
[0.13889685]
...
[0.24086878]
[0.07379474]
[0.9452105 ]]

```