

```
In [1]: # Credits: https://machinelearningmastery.com/sequence-classification-lstm-recurrent-neural-networks-python-keras/
# LSTM for sequence classification in the IMDB dataset
import numpy
from keras.datasets import imdb
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import LSTM
from keras.layers.embeddings import Embedding
from keras.preprocessing import sequence
import warnings
warnings.filterwarnings("ignore")
# fix random seed for reproducibility
numpy.random.seed(7)
```

Using TensorFlow backend.

```
In [2]: #Dataset taken from : https://keras.io/datasets/#imdb-movie-reviews-sentiment-classification

# load the dataset but only keep the top n words, zero the rest
top_words = 5000
(X_train, y_train), (X_test, y_test) = imdb.load_data(nb_words=top_words)
```

/usr/local/lib/python3.6/dist-packages/keras/datasets/imdb.py:49: UserWarning: The `nb_words` argument in `load_data` has been renamed `num_words`.
warnings.warn('The `nb_words` argument in `load_data` '

```
In [3]: print(X_train[1])
print(type(X_train[1]))
print(len(X_train[1]))
```

```
[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, 2
```

```

1, 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, 490
1, 19, 4, 1002, 5, 89, 29, 952, 46, 37, 4, 455, 9, 45, 43, 38, 1543, 19
05, 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, 22
0, 175, 136, 50, 9, 4373, 228, 2, 5, 2, 656, 245, 2350, 5, 4, 2, 131, 1
52, 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]
<class 'list'>
189

```

```

In [4]: # truncate and/or pad input sequences
# maximum length review be 600
max_review_length = 600
#padding maximum length review to X_train and X_test
X_train = sequence.pad_sequences(X_train, maxlen=max_review_length)
X_test = sequence.pad_sequences(X_test, maxlen=max_review_length)

print(X_train.shape)
print(X_train[1])

```

```

(25000, 600)
[ 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  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 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]

```

```

In [5]: # create the model
embedding_vecor_length = 32
model = Sequential()
model.add(Embedding(top_words, embedding_vecor_length, input_length=max
_review_length))
model.add(LSTM(100))
#Using Sigmoid Activation
model.add(Dense(1, activation='sigmoid'))
# Optimizer used : Adam
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['a

```

```
ccuracy'])
print(model.summary())
```

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 600, 32)	160000
lstm_1 (LSTM)	(None, 100)	53200
dense_1 (Dense)	(None, 1)	101
Total params: 213,301		
Trainable params: 213,301		
Non-trainable params: 0		
None		

```
In [6]: model.fit(X_train, y_train, nb_epoch=10, batch_size=64)
# Final evaluation of the model
scores = model.evaluate(X_test, y_test, verbose=0)
print("Accuracy: %.2f%%" % (scores[1]*100))
```

```
/usr/local/lib/python3.6/dist-packages/keras/models.py:981: UserWarning: The `nb_epoch` argument in `fit` has been renamed `epochs`.
warnings.warn('The `nb_epoch` argument in `fit` '
```

```
Epoch 1/10
25000/25000 [=====] - 295s 12ms/step - loss: 0.4567 - acc: 0.7825
Epoch 2/10
20160/25000 [=====>.....] - ETA: 57s - loss: 0.3694 - acc: 0.8343
25000/25000 [=====] - 295s 12ms/step - loss: 0.3647 - acc: 0.8385
Epoch 3/10
25000/25000 [=====] - 297s 12ms/step - loss: 0.5037 - acc: 0.7639
Epoch 4/10
4416/25000 [====>.....] - ETA: 4:03 - loss: 0.3698 - acc: 0.8385
25000/25000 [=====] - 297s 12ms/s
```

```

tep - loss: 0.3554 - acc: 0.8503
Epoch 5/10
24832/25000 [=====>.] - ETA: 1s - loss: 0.2673 -
acc: 0.896125000/25000 [=====] - 296s 12ms/ste
p - loss: 0.2671 - acc: 0.8961
Epoch 6/10
25000/25000 [=====] - 297s 12ms/step - loss:
0.2477 - acc: 0.9051
Epoch 7/10
5824/25000 [=====>.....] - ETA: 3:47 - loss: 0.1967
- acc: 0.926525000/25000 [=====] - 296s 12ms/s
tep - loss: 0.2131 - acc: 0.9180
Epoch 8/10
25000/25000 [=====] - 296s 12ms/step - loss:
0.1861 - acc: 0.9304
Epoch 9/10
256/25000 [.....] - ETA: 4:51 - loss: 0.1830
- acc: 0.929725000/25000 [=====] - 297s 12ms/s
tep - loss: 0.1659 - acc: 0.9392
Epoch 10/10
22784/25000 [=====>...] - ETA: 26s - loss: 0.1636
- acc: 0.938125000/25000 [=====] - 296s 12ms/s
tep - loss: 0.1657 - acc: 0.9374
Accuracy: 86.24%

```

Observation:

1. The training accuracy at the end of first epoch is 78%.
2. At the end of second epoch it is 83%.
3. At last it is 93%.