

Assignment - 4, authored by Guruprasath G

1. Download the dataset from here.

About the dataset

- Label - Ham or Spam
- Message - Message

```
import warnings
warnings.filterwarnings("ignore")
```

2. Importing Required Library

```
import re
import nltk
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from nltk.stem import WordNetLemmatizer
from nltk.corpus import stopwords
from wordcloud import WordCloud, STOPWORDS
```

3. Read dataset and do Preprocessing

```
df = pd.read_csv("SMSSpamCollection.csv", names=['label', 'message'])
```

```
df.head()
```

	label	message
0	ham	Go until jurong point, crazy.. Available only ...
1	ham	Ok lar... Joking wif u oni...
2	spam	Free entry in 2 a wkly comp to win FA Cup fina...
3	ham	U dun say so early hor... U c already then say...
4	ham	Nah I don't think he goes to usf, he lives aro...

```
df.tail()
```

	label	message
5567	spam	This is the 2nd time we have tried 2 contact u...
5568	ham	Will ü b going to esplanade fr home?
5569	ham	Pity, * was in mood for that. So...any other s...
5570	ham	The guy did some bitching but I acted like i'd...
5571	ham	Rofl. Its true to its name

```
df.info()
```



```

[ 0, 0, 0, ..., 7119, 1101, 3568],
[ 0, 0, 0, ..., 852, 1, 10],
[ 0, 0, 0, ..., 2204, 332, 154]])

from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
y = le.fit_transform(df['label'])

from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test =
train_test_split(padded_seq,y,test_size=0.25,random_state=42)

X_train.shape

(4179, 77)

```

5. Add Layers

```

TOT_SIZE = len(token.word_index) + 1
model = Sequential()
#IP Layer
model.add(Embedding(TOT_SIZE,32,input_length=max_length_sequence))
model.add(LSTM(units=50, activation = 'relu',return_sequences=True))
model.add(Dropout(0.2))
#Layer2
model.add(LSTM(units=60, activation = 'relu'))
model.add(Dropout(0.3))
#output layer
model.add(Dense(units=1, activation='sigmoid'))

model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 77, 32)	227872
lstm (LSTM)	(None, 77, 50)	16600
dropout (Dropout)	(None, 77, 50)	0
lstm_1 (LSTM)	(None, 60)	26640
dropout_1 (Dropout)	(None, 60)	0
dense (Dense)	(None, 1)	61

```

Total params: 271,173
Trainable params: 271,173

```

Non-trainable params: 0

6,7 Compile and Fit the model

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

```
model.fit(X_train, y_train,validation_data=(X_test,y_test), epochs=10)
```

Epoch 1/10

```
131/131 [=====] - 26s 148ms/step - loss:  
0.7357 - accuracy: 0.8643 - val_loss: 0.2063 - val_accuracy: 0.8665
```

Epoch 2/10

```
131/131 [=====] - 19s 145ms/step - loss:  
0.1576 - accuracy: 0.9691 - val_loss: 36236.3438 - val_accuracy:  
0.9856
```

Epoch 3/10

```
131/131 [=====] - 16s 122ms/step - loss:  
16435255.0000 - accuracy: 0.9864 - val_loss: 0.2386 - val_accuracy:  
0.9907
```

Epoch 4/10

```
131/131 [=====] - 19s 144ms/step - loss:  
0.1788 - accuracy: 0.9935 - val_loss: 21.0637 - val_accuracy: 0.9892
```

Epoch 5/10

```
131/131 [=====] - 16s 119ms/step - loss:  
1.2619 - accuracy: 0.9945 - val_loss: 4.3433 - val_accuracy: 0.9892
```

Epoch 6/10

```
131/131 [=====] - 15s 114ms/step - loss:  
0.0952 - accuracy: 0.9947 - val_loss: 4.9345 - val_accuracy: 0.9899
```

Epoch 7/10

```
131/131 [=====] - 15s 112ms/step - loss:  
0.0582 - accuracy: 0.9959 - val_loss: 10.5201 - val_accuracy: 0.9871
```

Epoch 8/10

```
131/131 [=====] - 15s 113ms/step - loss:  
0.2283 - accuracy: 0.9962 - val_loss: 5.4732 - val_accuracy: 0.9856
```

Epoch 9/10

```
131/131 [=====] - 16s 124ms/step - loss:  
0.0805 - accuracy: 0.9971 - val_loss: 8.9843 - val_accuracy: 0.9892
```

Epoch 10/10

```
131/131 [=====] - 16s 119ms/step - loss:  
0.2049 - accuracy: 0.9974 - val_loss: 25.0567 - val_accuracy: 0.9828
```

<keras.callbacks.History at 0x1cfd890cac0>

```
model.evaluate(X_test,y_test)
```

```
44/44 [=====] - 2s 37ms/step - loss: 25.0567  
- accuracy: 0.9828
```

```
[25.056659698486328, 0.9827709794044495]
```

8. Save the Model

```
from pickle import dump,load
tfid = 'tfid.sav'
lstm = 'lstm.sav'

dump(token,open(tfid,'wb'))
model.save('nlp.h5')
```

9. Test the Model

```
def preprocess(raw_mess):
    review = re.sub('[^a-zA-Z]', ' ', raw_mess)
    review = review.lower()
    review = review.split()
    review = [lemmatizer.lemmatize(i) for i in review if not i in
set(stopwords.words('english'))]
    review = ' '.join(review)
    return review

def predict(mess):
    vect = load(open(tfid,'rb'))
    classifier = load_model('nlp.h5')
    clean = preprocess(mess)
    text_to_seq = token.texts_to_sequences([mess])
    padded_seq = pad_sequences(text_to_seq, maxlen=77, padding="pre")
    pred = classifier.predict(padded_seq)
    return pred
```

```
msg = input("Enter a message: ")
predi = predict(msg)
if predi >= 0.6:
    print("It is a spam")
else:
    print("Not a spam")
```

```
Enter a message: Hey hi how are you?
1/1 [=====] - 1s 658ms/step
Not a spam
```

```
msg = input("Enter a message: ")
predi = predict(msg)
if predi >= 0.6:
    print("It is a spam")
else:
    print("Not a spam")
```

```
Enter a message: You won a cash reward of 3000rs!!!
1/1 [=====] - 1s 552ms/step
It is a spam
```