

SMS SPAM Classification

Download the dataset "spam.csv"

Import required library

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from keras.models import Model
from keras.layers import LSTM, Activation, Dense, Dropout, Input, Embedding
from keras.optimizers import Adam
from keras.preprocessing.text import Tokenizer
from keras.preprocessing import sequence
from keras.utils import pad_sequences
from keras.utils import to_categorical
from keras.callbacks import EarlyStopping
```

Read the Dataset

```
In [2]:
df = pd.read_csv('spam.csv', delimiter=',', encoding='latin-1')
df.head()
```

```
Out[2]:
```

	v1	v2	Unnamed: 2	Unnamed: 3	Unnamed: 4
0	ham	Go until jurong point, crazy.. Available only ...	NaN	NaN	NaN
1	ham	Ok lar... Joking wif u oni...	NaN	NaN	NaN
2	spam	Free entry in 2 a wkly comp to win FA Cup fina...	NaN	NaN	NaN
3	ham	U dun say so early hor... U c already then say...	NaN	NaN	NaN
4	ham	Nah I don't think he goes to usf, he lives aro...	NaN	NaN	NaN

Preprocessing the Dataset

```
df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'],axis=1,inplace=True)
```

```
from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator
```

```
X = df.v2
Y = df.v1
le = LabelEncoder()
Y = le.fit_transform(Y)
Y = Y.reshape(-1,1)
```

```
X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.25)
```

```
max_words = 1000
max_len = 150
tok = Tokenizer(num_words=max_words)
tok.fit_on_texts(X_train)
sequences = tok.texts_to_sequences(X_train)
sequences_matrix = pad_sequences(sequences,maxlen=max_len)
```

Create Model & Add Layers

```
inputs = Input(shape=[max_len])
layer = Embedding(max_words,50,input_length=max_len)(inputs)
layer = LSTM(128)(layer)
layer = Dense(128)(layer)
layer = Activation('relu')(layer)
layer = Dropout(0.5)(layer)
layer = Dense(1)(layer)
layer = Activation('sigmoid')(layer)
model = Model(inputs=inputs,outputs=layer)
```

```
model.summary()
Model: "model"
```

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 150)]	0
embedding (Embedding)	(None, 150, 50)	50000
lstm (LSTM)	(None, 128)	91648
dense (Dense)	(None, 128)	16512
activation (Activation)	(None, 128)	0

dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 1)	129
activation_1 (Activation)	(None, 1)	0

```
=====
Total params: 158,289
Trainable params: 158,289
Non-trainable params: 0
=====
```

Compile the Model

```
In [10]:
model.compile(loss='binary_crossentropy', optimizer=Adam(), metrics=['accuracy'])
```

Fit the Model

```
In [11]:
history = model.fit(sequences_matrix, Y_train, batch_size=20, epochs=15,
                    validation_split=0.2)

Epoch 1/15
168/168 [=====] - 10s 16ms/step - loss: 0.2235 - accuracy: 0.9246 - val_loss: 0.0524 - val_accuracy: 0.9856
Epoch 2/15
168/168 [=====] - 2s 11ms/step - loss: 0.0457 - accuracy: 0.9877 - val_loss: 0.0351 - val_accuracy: 0.9904
Epoch 3/15
168/168 [=====] - 2s 11ms/step - loss: 0.0265 - accuracy: 0.9934 - val_loss: 0.0458 - val_accuracy: 0.9868
Epoch 4/15
168/168 [=====] - 2s 10ms/step - loss: 0.0139 - accuracy: 0.9964 - val_loss: 0.0434 - val_accuracy: 0.9868
Epoch 5/15
168/168 [=====] - 2s 11ms/step - loss: 0.0104 - accuracy: 0.9976 - val_loss: 0.0539 - val_accuracy: 0.9868
Epoch 6/15
168/168 [=====] - 2s 10ms/step - loss: 0.0078 - accuracy: 0.9979 - val_loss: 0.0878 - val_accuracy: 0.9821
Epoch 7/15
168/168 [=====] - 2s 10ms/step - loss: 0.0072 - accuracy: 0.9973 - val_loss: 0.0639 - val_accuracy: 0.9880
Epoch 8/15
168/168 [=====] - 2s 10ms/step - loss: 0.0052 - accuracy: 0.9982 - val_loss: 0.0637 - val_accuracy: 0.9868
Epoch 9/15
168/168 [=====] - 2s 10ms/step - loss: 0.0039 - accuracy: 0.9994 - val_loss: 0.0561 - val_accuracy: 0.9868
Epoch 10/15
168/168 [=====] - 2s 10ms/step - loss: 0.0044 - accuracy: 0.9985 - val_loss: 0.0661 - val_accuracy: 0.9868
```

```

Epoch 11/15
168/168 [=====] - 2s 11ms/step - loss: 0.0010 - ac
curacy: 1.0000 - val_loss: 0.0853 - val_accuracy: 0.9880
Epoch 12/15
168/168 [=====] - 2s 10ms/step - loss: 0.0056 - ac
curacy: 0.9979 - val_loss: 0.0812 - val_accuracy: 0.9856
Epoch 13/15
168/168 [=====] - 2s 10ms/step - loss: 0.0031 - ac
curacy: 0.9991 - val_loss: 0.0779 - val_accuracy: 0.9809
Epoch 14/15
168/168 [=====] - 2s 10ms/step - loss: 8.6841e-04
- accuracy: 1.0000 - val_loss: 0.0909 - val_accuracy: 0.9809
Epoch 15/15
168/168 [=====] - 2s 10ms/step - loss: 0.0028 - ac
curacy: 0.9991 - val_loss: 0.0776 - val_accuracy: 0.9856

```

In [12]:

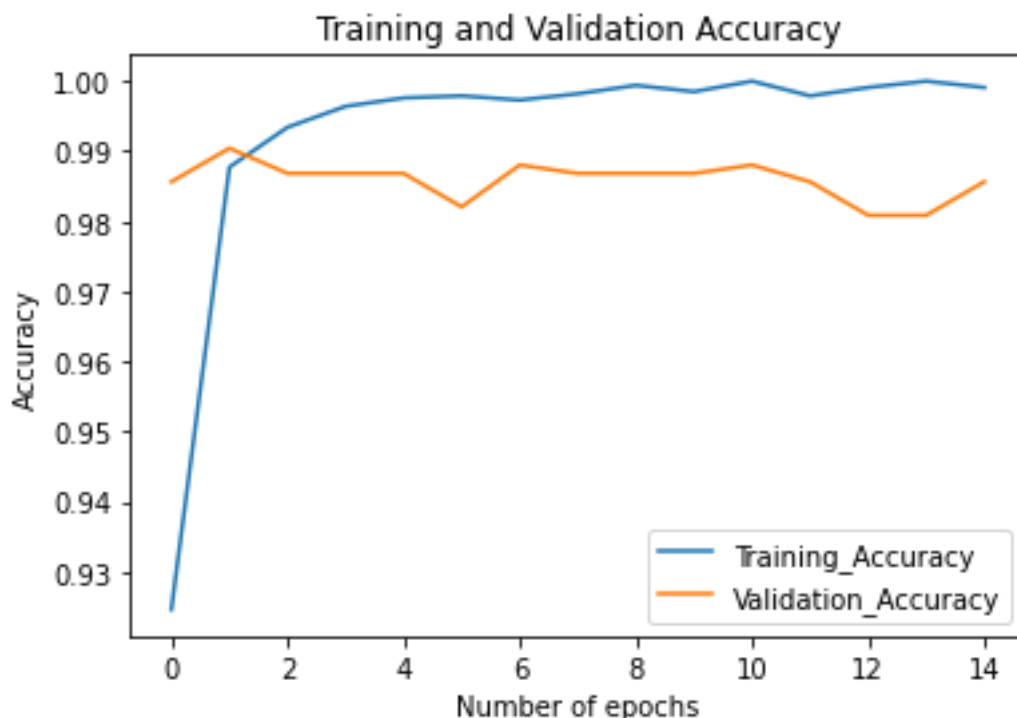
```

metrics = pd.DataFrame(history.history)
metrics.rename(columns = {'loss': 'Training_Loss', 'accuracy':
'Training_Accuracy', 'val_loss': 'Validation_Loss', 'val_accuracy':
'Validation_Accuracy'}, inplace = True)
def plot_graphs1(var1, var2, string):
    metrics[[var1, var2]].plot()
    plt.title('Training and Validation ' + string)
    plt.xlabel ('Number of epochs')
    plt.ylabel(string)
    plt.legend([var1, var2])

```

In [13]:

```
plot_graphs1('Training_Accuracy', 'Validation_Accuracy', 'Accuracy')
```



Save the Model

In [14]:

```
model.save('A4Spam_sms_classifier.h5')
```

Test the Model

In [15]:

```
test_sequences = tok.texts_to_sequences(X_test)
test_sequences_matrix = pad_sequences(test_sequences, maxlen=max_len)
```

In [16]:

```
accuracy1 = model.evaluate(test_sequences_matrix, Y_test)
44/44 [=====] - 0s 7ms/step - loss: 0.1204 - accuracy: 0.9864
```

In [17]:

```
print(' loss: {:.4f}'.format(accuracy1[0]))
print(' Accuracy: {:.4f}'.format(accuracy1[1]))
loss: 0.1204
Accuracy: 0.9864
```