Assignment -4
SMS SPAM Classification

Assignment Date	10 NOVEMBER 2022
Team ID	PNT2022TMID46604
Project Name	AI BASED DISCOURSE FOR BANKING INDUSTRY
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Maximum Marks	2 Marks

Question-1. Import required library

Solution:

```
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.layersimport LSTM,Activation, Dense, Dropout, Input, Embedding
from keras.optimizers import Adam
fromkeras.preprocessing.textimportTokenizer
from keras.preprocessing import sequence
from keras.utils import pad_sequences
from keras.utils import to_categorical
from keras.callbacks import EarlyStopping
```

Question-2. Read the Dataset

Solution:

```
df = pd.read_csv('spam.csv',delimiter=',',encoding='latin-1')
df.head()
```

	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

Question-3. Preprocessing the Dataset

Solution:

```
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 =                    tok.fit_on_texts(X_tr
Tokenizer(num_wordain)
s=max                    words)
_

sequences = tok.texts_to_sequences(X_train)
sequences_matrix = pad_sequenc
                                    es(sequences,maxlen=max_len)
```

Question-4. Create Model

Solution:

```
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)

Question-5. Add Layers (LSTM, Dense-(Hidden Layers), Output)

Solution:

model.summary()

```
Model: "model_1"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 input_2 (InputLayer)        [(None, 150)]             0

 embedding_1 (Embedding)     (None, 150, 50)           50000

 lstm_1 (LSTM)               (None, 128)               91648

 dense_2 (Dense)             (None, 128)               16512

 activation_2 (Activation)   (None, 128)               0

 dropout_1 (Dropout)         (None, 128)               0

 dense_3 (Dense)             (None, 1)                 129

 activation_3 (Activation)   (None, 1)                 0

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

Question-6. Compile the Model

Solution:

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

Question-7. Fit the Model

Solution:

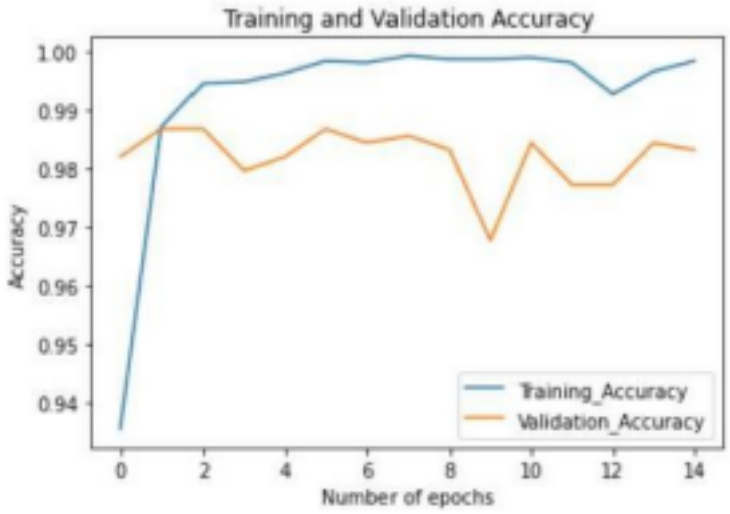
```
history = model.fit(sequences_matrix,Y_train,batch_size=20,epochs=15,
        validation_split=0.2)
```

```
Epoch 1/15
168/168 [==============================] - 34s 190ms/step - loss: 0.1980 - accuracy: 0.9354 - val_loss: 0.0649 - val_accuracy: 0.9821
Epoch 2/15
168/168 [==============================] - 31s 185ms/step - loss: 0.0416 - accuracy: 0.9871 - val_loss: 0.0513 - val_accuracy: 0.9868
Epoch 3/15
168/168 [==============================] - 31s 186ms/step - loss: 0.0217 - accuracy: 0.9946 - val_loss: 0.0613 - val_accuracy: 0.9868
Epoch 4/15
168/168 [==============================] - 33s 198ms/step - loss: 0.0155 - accuracy: 0.9949 - val_loss: 0.0779 - val_accuracy: 0.9797
Epoch 5/15
168/168 [==============================] - 32s 188ms/step - loss: 0.0132 - accuracy: 0.9964 - val_loss: 0.0661 - val_accuracy: 0.9821
Epoch 6/15
168/168 [==============================] - 32s 190ms/step - loss: 0.0065 - accuracy: 0.9985 - val_loss: 0.0772 - val_accuracy: 0.9868
Epoch 7/15
168/168 [==============================] - 32s 192ms/step - loss: 0.0057 - accuracy: 0.9982 - val_loss: 0.0811 - val_accuracy: 0.9844
Epoch 8/15
168/168 [==============================] - 32s 191ms/step - loss: 0.0045 - accuracy: 0.9994 - val_loss: 0.0877 - val_accuracy: 0.9856
Epoch 9/15
168/168 [==============================] - 32s 189ms/step - loss: 0.0046 - accuracy: 0.9988 - val_loss: 0.1282 - val_accuracy: 0.9833
Epoch 10/15
168/168 [==============================] - 32s 188ms/step - loss: 0.0066 - accuracy: 0.9988 - val_loss: 0.1191 - val_accuracy: 0.9677
Epoch 11/15
168/168 [==============================] - 33s 194ms/step - loss: 0.0036 - accuracy: 0.9991 - val_loss: 0.1149 - val_accuracy: 0.9844
Epoch 12/15
168/168 [==============================] - 31s 186ms/step - loss: 0.0131 - accuracy: 0.9982 - val_loss: 0.1019 - val_accuracy: 0.9773
Epoch 13/15
168/168 [==============================] - 31s 187ms/step - loss: 0.0251 - accuracy: 0.9928 - val_loss: 0.1015 - val_accuracy: 0.9773
Epoch 14/15
168/168 [==============================] - 31s 187ms/step - loss: 0.0081 - accuracy: 0.9967 - val_loss: 0.1005 - val_accuracy: 0.9844
Epoch 15/15
168/168 [==============================] - 32s 188ms/step - loss: 0.0048 - accuracy: 0.9985 - val_loss: 0.0985 - val_accuracy: 0.9833
```

```
metrics = pd.DataFrame(history.history)
metrics.rename(columns ={'loss':'Training_Loss', 'accuracy':'Training_Accuracy', 'val_loss': 'Valida
tion_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])

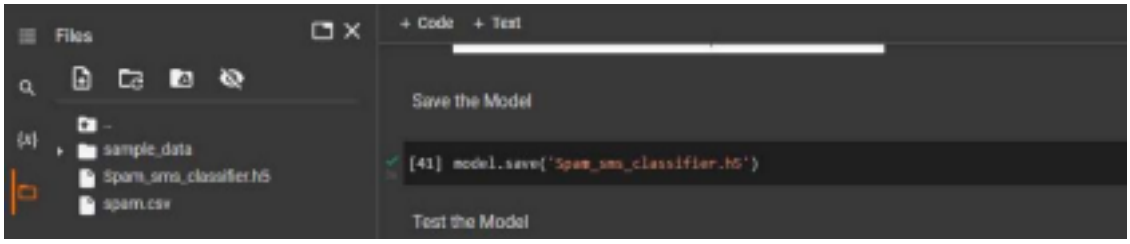
plot_graphs1('Training_Accuracy', 'Validation_Accuracy', 'Accuracy')



Question-8. Save The Model

Solution:

model.save('Spam_sms_classifier.h5')



Question-9. Test The Model

Solution:

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

accuracy1 = model.evaluate(test_sequences_matrix,Y_test)

```
44/44 [==============================] - 4s 82ms/step - loss: 0.1061 - accuracy: 0.9828
```

```
print(' loss: {:0.4f}'.format(accuracy1[0]))

print(' Accuracy: {:0.4f}'.format(accuracy1[1]))
```

```
loss: 0.1061
Accuracy: 0.9828
```