

Assignment -4
SMS SPAM Classification

Assignment Date	26 November 2022
Team ID	PNT2022TMID46259
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.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
```

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. Pre processing 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 = 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)
```

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)
```

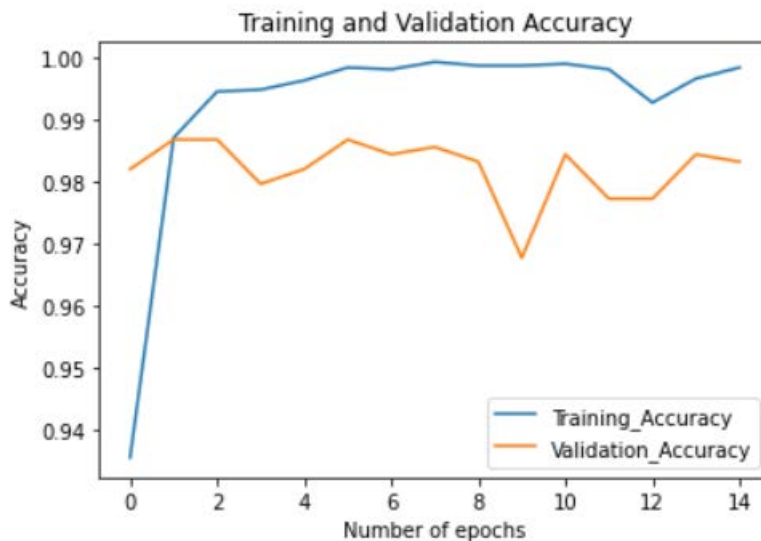
```
metrics.rename(columns = {'loss': 'Training_Loss', 'accuracy': 'Training_Accuracy', 'val_loss': 'Validation_Loss', 'val_accuracy': 'Validation_Accuracy'}, inplace = True)
```

```
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])
```

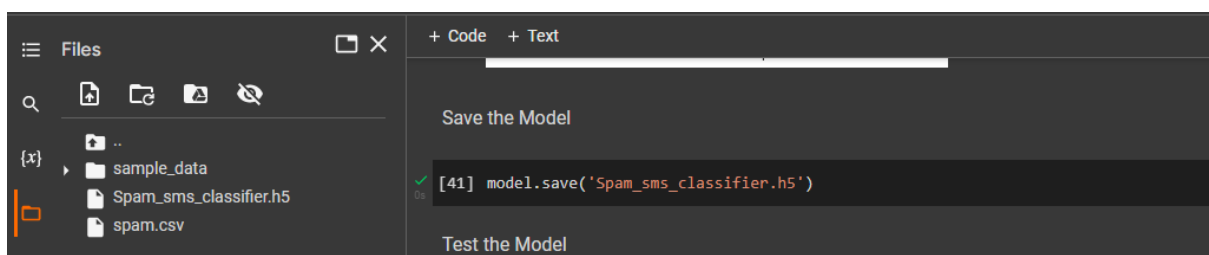
```
plot_graphs1('Training_Accuracy', 'Validation_Accuracy', 'Accuracy')
```



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: {:.4f}'.format(accuracy1[0]))
```

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

```
loss: 0.1061
```

```
Accuracy: 0.9828
```