# Assignment -4 SMS SPAM Classification

Assignment Date	26 October 2022
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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. 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 =
    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()

1 7		i	7	. 11
Model	:	mod	eT	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

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

#### Solution:

model.compile(loss='binary\_crossentropy',optimizer=Adam(),metrics=['accuracy'])

Question-7. Fit the Model

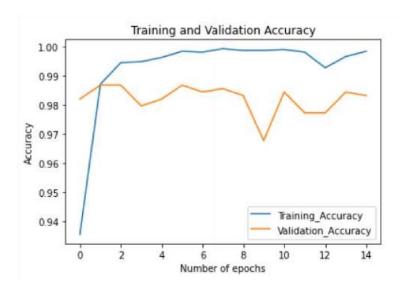
#### Solution:

history = model.fit(sequences\_matrix,Y\_train,batch\_size=20,epochs=15, validation split=0.2)

```
Epoch 1/15
Fpoch 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 [===
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
Fnoch 10/15
Epoch 11/15
168/168 [===========] - 33s 194ms/step - loss: 0.0036 - accuracy: 0.991 - val_loss: 0.1149 - val_accuracy: 0.9844
Epoch 12/15
Fnoch 13/15
168/168 [========] - 31s 187ms/step - loss: 0.0251 - accuracy: 0.928 - 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
```

```
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