# Assignment -4 SMS SPAM Classification

Assignment Date	20 November 2022
Team ID	PNT2022TMID25140
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()
```

	<b>v1</b>	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

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

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#### Question-6. Compile the Model

#### **Solution:**

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

Question-7. Fit the Model

#### Solution:

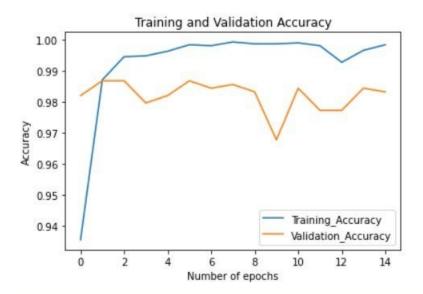
history = model.fit(sequences\_matrix,Y\_train,batch\_size=20,epochs=15, validation split=0.2)

```
Epoch 1/15
Epoch 2/15
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
Fnoch 5/15
Epoch 6/15
168/168 [============ ] - 32s 190ms/step - loss: 0.0065 - accuracy: 0.985 - 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
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: {:0.4f}'.format(accuracy1[0]))
print(' Accuracy: {:0.4f}'.format(accuracy1[1]))
```

loss: 0.1061 Accuracy: 0.9828