# Assignment -4 SMS SPAM Classification

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Team ID	PNP2022TMID03265
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 Jok <mark>ing wif</mark> 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 Dataaset**

### 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_sequencees(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,outpputs=layer)

Question-5. Add Layers (LSTM, Dennse-(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

# **Question-6.Compile the Model**

#### Solution:

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

#### Question-7. Fit the Model

#### Solution:

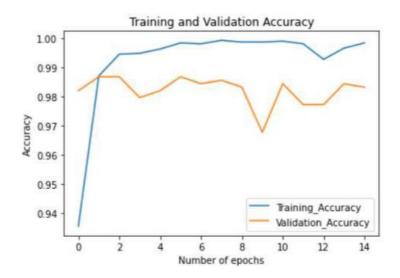
history = model.fit(sequences\_mattrix,Y\_train,batch\_size=20,epochs=15, validation\_split=0.2)

```
Epoch 1/15
acy: 0.9821
Epoch 2/15
168/168 [============] - 31s 185ms/step - loss: 0.0416 - accuracy: 0.9871 - val loss:
                                                                  acy: 0.9868
Epoch 3/15
acy: 0.9868
168/168 [============= - 33s 198ms/step - loss: 0.0155 - accuracy: 0.9949 - val loss:
                                                                  acy: 0.9797
Epoch 5/15
168/168 [============= ] - 32s 188ms/step - loss: 0.0132 - accuracy: 0.9964 - val loss:
                                                                  acy: 0.9821
Epoch 6/15
168/168 [============= - 32s 190ms/step - loss: 0.0065 - accuracy: 0.9985 - val loss:
                                                                  acy: 0.9868
Epoch 7/15
acy: 0.9844
Epoch 8/15
168/168 [==========] - 32s 191ms/step - loss: 0.0045 - accuracy: 0.9994 - val loss:
                                                                  acy: 0.9856
Epoch 9/15
acy: 0.9833
Epoch 10/15
168/168 [============] - 32s 188ms/step - loss: 0.0066 - accuracy: 0.9988 - val_loss:
                                                                  acy: 0.9677
Epoch 11/15
168/168 [============ ] - 33s 194ms/step - loss: 0.0036 - accuracy: 0.9991 - val loss:
                                                                  acy: 0.9844
Epoch 12/15
acy: 0.9773
Epoch 13/15
acy: 0.9773
Epoch 14/15
acy: 0.9844
Epoch 15/15
168/168 [=============] - 32s 188ms/step - loss: 0.0048 - accuracy: 0.9985 - val_loss:
                                                                  acy: 0.9833
```

```
metrics = pd.DataFrame(history.history)
metrics.rename(columns = {'loss': 'Training_Loss', 'accuracy': 'Training_Accuracy', 'val_loss': 'Validation_Loss', 'val_accuracy': 'Validatiion_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\_seqquences(X\_test)

```
test_sequences_matrix = pad_sequuences(test_sequences,maxlen=max_len)
```

# accuracy1 = model.evaluate(test\_sequences\_matrix,Y\_test)

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

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

loss: 0.1061 Accuracy: 0.9828