## Assignment -4

## **Python Programming**

Assignment Date	20 October 2022
Student Name	Ajay Sai Ram N
Student Roll Number	212219060010
Maximum Marks	2 Marks

- 1. Download the dataset "spam.csv"
- 2. Import required library

```
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 pad_sequences
from keras.utils import to_categorical
from keras.callbacks import EarlyStopping
```

3. Read the data set

```
In [13]:
    df = pd.read_csv('spam.csv',delimiter=',',encoding='latin-1')
    df.head()
```

t[13]:		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

### Preprocessing the dataset

# 4. CREATE MODEL & ADD LAYERS

```
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 = Dense(1)(layer)
layer = Activation('sigmoid')(layer)
model = Model(inputs=inputs,outputs=layer)
```

In [20]:

```
model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 150)]	0
embedding (Embedding)	(None, 150, 50)	50000
1stm (LSTM)	(None, 128)	91648
dense (Dense)	(None, 128)	16512
activation (Activation)	(None, 128)	0
dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 1)	129
activation_1 (Activation)	(None, 1)	0
Total params: 158,289 Trainable params: 158,289 Non-trainable params: 0		

#### 5.Compile the Model

In [24]:

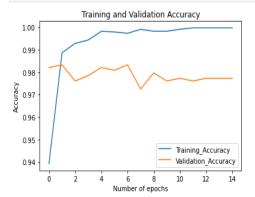
```
model.compile(loss='binary_crossentropy',optimizer=Adam(),metrics=['accuracy'])
```

#### **6.FIT THE MODEL**

```
168/168 [===========] - 35s 189ms/step - loss: 0.1840 - accuracy: 0.9393 - val_loss: 0.0674 - val_accuracy: 0.9821
Fnoch 2/15
168/168 [===:
     Epoch 3/15
168/168 [===
     Epoch 4/15
168/168 [===
      Epoch 5/15
     168/168 [====
Epoch 6/15
Epoch 7/15
168/168 [====
     Epoch 8/15
Epoch 9/15
168/168 [====
     Epoch 10/15
168/168 [============] - 31s 187ms/step - loss: 0.0051 - accuracy: 0.9982 - val loss: 0.1406 - val accuracy: 0.9761
Epoch 11/15
168/168 [===:
      =========] - 31s 186ms/step - loss: 0.0041 - accuracy: 0.9991 - val_loss: 0.1406 - val_accuracy: 0.9773
Epoch 12/15
168/168 [===========] - 31s 187ms/step - loss: 0.0025 - accuracy: 0.9997 - val_loss: 0.1444 - val_accuracy: 0.9761
Epoch 13/15
Epoch 14/15
168/168 [===:
     Epoch 15/15
```

```
In [26]:
    metrics = pd.DataFrame(history.history)
    metrics.rename(columns = {'loss': 'Training_Loss', 'accuracy': 'Training_Accuracy', 'val_loss': 'Validation_Loss', 'val_accuracy': 'Validation_Accuracy'
    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])
```

```
In [30]: plot_graphs1('Training_Accuracy', 'Validation_Accuracy', 'Accuracy')
```



#### 7. SAVE THE MODEL

```
In [33]: model.save('A4Spam_sms_classifier.h5')
```

## 8.TEST THE MODEL

```
In [34]: test_sequences = tok.texts_to_sequences(X_test)
test_sequences_matrix = pad_sequences(test_sequences,maxlen=max_len)

In [35]: accuracy1 = model.evaluate(test_sequences_matrix,Y_test)
```

```
44/44 [========] - 4s 80ms/step - loss: 0.1045 - accuracy: 0.9864
```

```
In [40]: print(' loss: {:0.4f}'.format(accuracy1[0]))
    print(' Accuracy: {:0.4f}'.format(accuracy1[1]))
```

loss: 0.1045 Accuracy: 0.9864