# **Importing Model building libraries**

```
import pandas as pd
import numpy as np
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 RMSprop
from keras.preprocessing.text import Tokenizer
from keras_preprocessing import sequence
from keras.utils import to_categorical
from keras.models import load_model
```

### **Importing NLTK libraries**

```
import csv
import tensorflow as tf
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
import nltk
nltk.download('stopwords')
from nltk.corpus import stopwords
STOPWORDS = set(stopwords.words('english'))

[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

### Reading dataset and preprocessing

df.drop(['Unnamed: 2','Unnamed: 3', 'Unnamed: 4'],axis=1,inplace=True)

```
from google.colab import drive
 drive.mount('/content/drive')
Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force remount=True).
cd/content/drive/MyDrive/Colab Notebooks
/content/drive/MyDrive/Colab Notebooks
 df = pd.read_csv('/content/drive/MyDrive/AI_IBM/spam.csv',delimiter=',',encoding='latin-1')
df.head()
                                          v2 Unnamed: 2 Unnamed: 3 Unnamed: 4
                                                   NaN
                                                               NaN
0 ham Go until jurong point, crazy.. Available only ...
                        Ok lar... Joking wif u oni...
                                                   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...
```

```
RangeIndex: 5572 entries, 0 to 5571
Data columns (total 2 columns):
# Column Non-Null Count Dtype
0 v1 5572 non-null object
1 v2 5572 non-null object
1 v2 5572
dtypes: object(2)
memory usage: 87.2+ KB
df.groupby(['v1']).size()
ν1
        4825
spam
dtype: int64
 #Label Encoding Required Column
 X = df.v2
 Y = df.v1
 le = LabelEncoder()
 Y = le.fit_transform(Y)
Y = Y.reshape(-1,1)
 # Test and train data split
X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.15)
 # Tokenisation function
 max_words = 1000
max_len = 150
tok = Tokenizer(num_words=max_words)
 tok = Tokenizer(num_words=max_words)
tok.fit_on_texts(X_train)
sequences = tok.texts_to_sequences(X_train)
 sequences_matrix = sequence.pad_sequences(sequences,maxlen=max_len)
```

#### **Create Model**

## Add layers (LSTM , Dense-(HiddenLayers), Ouput)

```
inputs = Input(name='InputLayer',shape=[max_len])
layer = Embedding(max_words,50,input_length=max_len)(inputs)
layer = LSTM(64)(layer)
layer = Dense(256,name='FullyConnectedLayer1')(layer)
layer = Activation('relu')(layer)
layer = Dropout(0.5)(layer)
layer = Dense(1,name='OutputLayer')(layer)
layer = Activation('sigmoid')(layer)
model = Model(inputs=inputs,outputs=layer)
model.summary()
model.compile(loss='binary_crossentropy',optimizer=RMSprop(),metrics=['accuracy'])
Model: "model"
Layer (type)
                           Output Shape
                                                     Param #
InputLayer (InputLayer) [(None, 150)]
embedding (Embedding)
                          (None, 150, 50)
                                                    50000
                           (None, 64)
                                                    29440
FullyConnectedLayer1 (Dense (None, 256)
                                                     16640
```

```
activation (Activation) (None, 256) 0

dropout (Dropout) (None, 256) 0

OutputLayer (Dense) (None, 1) 257

activation_1 (Activation) (None, 1) 0
```

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Total params: 96,337 Trainable params: 96,337 Non-trainable params: 0

#### model.fit(sequences\_matrix,Y\_train,batch\_size=128,epochs=25,validation\_split=0.2)

```
Epoch 1/25
30/30 [====
                     :=========] - 28s 720ms/step - loss: 0.3323 - accuracy: 0.8772 - val_loss: 0.1085 - val_accuracy: 0.9715
Epoch 2/25
30/30 [===
                                         18s 588ms/step - loss: 0.0818 - accuracy: 0.9807 - val_loss: 0.0794 - val_accuracy: 0.9800
Epoch 3/25
30/30 [===
                                       - 12s 384ms/step - loss: 0.0421 - accuracy: 0.9884 - val loss: 0.0518 - val accuracy: 0.9842
Epoch 4/25
                      =========] - 9s 291ms/step - loss: 0.0293 - accuracy: 0.9921 - val_loss: 0.0461 - val_accuracy: 0.9884
30/30 [=====
Epoch 5/25
30/30 [===:
                                         9s 288ms/step - loss: 0.0261 - accuracy: 0.9921 - val_loss: 0.0517 - val_accuracy: 0.9873
Epoch 6/25
30/30 [===:
                                         9s 291ms/step - loss: 0.0161 - accuracy: 0.9952 - val_loss: 0.0582 - val_accuracy: 0.9863
Enoch 7/25
30/30 [===
                                       - 9s 291ms/step - loss: 0.0110 - accuracy: 0.9971 - val_loss: 0.0660 - val_accuracy: 0.9895
Epoch 8/25
                                       - 11s 369ms/step - loss: 0.0087 - accuracy: 0.9974 - val loss: 0.0765 - val accuracy: 0.9863
30/30 [===:
Epoch 9/25
30/30 [=====
                    ==========] - 9s 294ms/step - loss: 0.0059 - accuracy: 0.9982 - val_loss: 0.0815 - val_accuracy: 0.9884
Epoch 10/25
30/30 [====
                             =======] - 9s 290ms/step - loss: 0.0051 - accuracy: 0.9987 - val_loss: 0.0902 - val_accuracy: 0.9852
EDOCH 11/25
30/30 [==
                                        9s 318ms/step - loss: 0.0038 - accuracy: 0.9987 - val_loss: 0.0964 - val_accuracy: 0.9884
Epoch 12/25
30/30 [===:
                                        9s 290ms/step - loss: 0.0039 - accuracy: 0.9984 - val_loss: 0.1214 - val_accuracy: 0.9863
Epoch 13/25
30/30 [===:
                                   =] - 11s 363ms/step - loss: 0.0011 - accuracy: 0.9997 - val_loss: 0.1153 - val_accuracy: 0.9895
Enoch 14/25
30/30 [====
                                        9s 294ms/step - loss: 6.9965e-04 - accuracy: 0.9997 - val_loss: 0.1322 - val_accuracy: 0.9873
Epoch 15/25
30/30 [=====
                       ========] - 9s 292ms/step - loss: 0.7710 - accuracy: 0.9739 - val loss: 0.1286 - val accuracy: 0.9884
Epoch 16/25
30/30 [====
                                 ====] - 9s 294ms/step - loss: 5.0771e-04 - accuracy: 0.9997 - val_loss: 0.1294 - val_accuracy: 0.9895
Epoch 17/25
30/30 [====
                             :======] - 9s 296ms/step - loss: 2.4364e-04 - accuracy: 1.0000 - val loss: 0.1362 - val accuracy: 0.9895
Epoch 18/25
                                      - 9s 293ms/step - loss: 7.7019e-05 - accuracy: 1.0000 - val loss: 0.1435 - val accuracy: 0.9863
30/30 [====
Epoch 19/25
30/30 [=====
                         ========] - 9s 294ms/step - loss: 4.9329e-05 - accuracy: 1.0000 - val_loss: 0.1585 - val_accuracy: 0.9863
Epoch 20/25
                       =========] - 9s 310ms/step - loss: 3.0667e-05 - accuracy: 1.0000 - val loss: 0.1735 - val accuracy: 0.9863
30/30 [=====
Epoch 21/25
30/30 [=====
                       ========] - 9s 316ms/step - loss: 1.8201e-05 - accuracy: 1.0000 - val_loss: 0.1857 - val_accuracy: 0.9852
Epoch 22/25
30/30 [=====
                        ========] - 9s 295ms/step - loss: 7.7908e-06 - accuracy: 1.0000 - val loss: 0.2049 - val accuracy: 0.9884
Epoch 23/25
30/30 [===:
                           :======] - 9s 295ms/step - loss: 7.4443e-06 - accuracy: 1.0000 - val_loss: 0.2257 - val_accuracy: 0.9873
Epoch 24/25
30/30 [====
                          =======] - 9s 298ms/step - loss: 1.8775e-04 - accuracy: 1.0000 - val_loss: 0.2443 - val_accuracy: 0.9810
Epoch 25/25
30/30 [===
                          ========] - 9s 292ms/step - loss: 1.6095e-06 - accuracy: 1.0000 - val_loss: 0.2496 - val_accuracy: 0.9810
```