Assignment -3

Problem Statement :- SMS SPAM Classification

Assignment Date	27 October 2022	
Student Name	S.Karthika	
Student Roll Number	912419104014	
Maximum Marks	2 Marks	

Problem Statement:

Over recent years, as the popularity of mobile phone devices has increased, Short Message Service (SMS) has grown into a multi-billion dollar industry. At the same time, reduction in the cost of messaging services has resulted in growth in unsolicited commercial advertisements (spams) being sent to mobile phones. Due to Spam SMS, Mobile service providers suffer from some sort of financial problems as well as it reduces calling time for users. Unfortunately, if the user accesses such Spam SMS they may face the problem of virus or malware. When SMS arrives at mobile it will disturb mobile user privacy and concentration. It may lead to frustration for the user. So Spam SMS is one of the major issues in the wireless communication world and it grows day by day.

Import required library

Solution:

import numpy as np
import pandas as pd # for data preprocessing

Import required library

```
[189] import numpy as np
import pandas as pd # for data preprocessing
```

Question-1:

Read dataset and do pre-processing

Solution:

```
# Read dataset
spam=pd.read_csv('/content/spam.csv',encoding="ISO-8859-1")
spam
```

▼ Read dataset and do pre-processing



preprocessing
spam.head()

pre processing



spam.columns

```
[95] spam.columns

Index(['v1', 'v2', 'Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], dtype='object')
```

spam=spam.drop(columns=["Unnamed: 2","Unnamed: 3","Unnamed: 4"])
spam.info()

```
[96] spam=spam.drop(columns=["Unnamed: 2","Unnamed: 3","Unnamed: 4"])
 / [97] spam.info()
        <class 'pandas.core.frame.DataFrame'>
        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
        dtypes: object(2)
        memory usage: 87.2+ KB
        spam.isnull().sum()
        # scaling
        from sklearn.preprocessing import MinMaxScaler
        scl = MinMaxScaler()
        scl tr=scl.fit transform
        scl tr
       spam.isnull().sum()
   C→ v1
       v2
       dtype: int64

√ [99] # scaling

       from sklearn.preprocessing import MinMaxScaler
       scl = MinMaxScaler()
       scl_tr=scl.fit_transform
       scl_tr
       <bound method TransformerMixin.fit_transform of MinMaxScaler()>
        spam["Text Length"].describe()
/ [48] spam["Text Length"].describe()
                5572.000000
         count
                 80.118808
59.690841
         mean
         std
         min
                      2.000000
                   36.000000
61.000000
         25%
         50%
        75% 121.000000
max 910.000000
```

Name: Text Length, dtype: float64

```
import seaborn as sns
    import matplotlib.pyplot as plt
    fig=plt.figure()
    sns.kdeplot(
     x=spam["Text Length"],
     hue=spam["v1"]
     plt.title("ham & spam messege length comparision")
     plt.show()
       import seaborn as sns
       import matplotlib.pyplot as plt
       fig=plt.figure()
       sns.kdeplot(
          x=spam["Text Length"],
hue=spam["v1"]
      plt.title("ham & spam messege length comparision")
      plt.show()
  ⊏→
                   ham & spam messege length comparision
         0.012
         0.010
                                                    spam
         0.008
         0.006
         0.004
         0.000
from sklearn.preprocessing import LabelEncoder
X = spam.v2
Y = spam.v1
le = LabelEncoder()
Y = le.fit transform(Y)
Y = Y.reshape(-1,1)
from sklearn.model_selection import train_test_split
X train,X test,Y train,Y test = train test split(X,Y,test size=0.15)
from keras.preprocessing.text import Tokenizer
from keras.preprocessing import sequence
from keras.utils import pad_sequences
from keras.callbacks import EarlyStopping
max_words = 1000
max len = 150
tok = Tokenizer(num_words=max_words)
seq = tok.fit_on_texts(X_train)
sequences = tok.texts_to_sequences(X_train)
sequences matrix = pad sequences(sequences,maxlen=max len)
```

```
[135] from sklearn.preprocessing import LabelEncoder
       X = spam.v2
       Y = spam.v1
       le = LabelEncoder()
       Y = le.fit_transform(Y)
       Y = Y.reshape(-1,1)
[140] from sklearn.model_selection import train_test_split
       X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.15)
✓ [149] from keras.preprocessing.text import Tokenizer
        from keras.preprocessing import sequence
       from keras.utils import pad_sequences
       from keras.callbacks import EarlyStopping
       max\_words = 1000
       max_len = 150
       tok = Tokenizer(num_words=max_words)
       seq = tok.fit_on_texts(X_train)
       sequences = tok.texts_to_sequences(X_train)
       sequences_matrix = pad_sequences(sequences,maxlen=max_len)
```

Question-2:

Create Model

Solution:

from tensorflow.keras.models import Sequential from keras.layers import LSTM, Activation, Dense, Dropout, Input, Embeddir from keras.models import Model from keras.optimizers import RMSprop

model = Sequential()

create model

Question-3:

Add Layers (LSTM, Dense-(Hidden Layers), Output)

Solution:

```
def lstm():
    inputs = Input(name='inputs',shape=[max_len])
    layer = Embedding(max_words,50,input_length=max_len)(inputs)
    layer = LSTM(64)(layer) # LSTM
    layer = Dense(256,name='FC1')(layer) # DENSE
    layer = Activation('relu')(layer)
    layer = Dropout(0.5)(layer)
    layer = Dense(1,name='out_layer')(layer)
    layer = Activation('sigmoid')(layer)
    model = Model(inputs=inputs,outputs=layer) # OUTPUT
    return model

model = Istm()
```

Add Layers (LSTM, Dense-(Hidden Layers), Output)

Question-4:

Compile The Model

Solution:

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

→ Compile the Model

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

Question-5:

Fit The Model

Solution:

model.fit(sequences_matrix,Y_train,batch_size=128,epochs=10, validation_split=0.2,callbacks=[EarlyStopping(monitor='val_loss', min_delta=0.0001)])

→ Fit the Model

Question-6:

Save The Model

Solution:

model.save('lstm.h5')

▼ Save The Model

```
model.save('lstm.h5')
```

Question-7:

Test The Model

Solution:

```
test_sequences = tok.texts_to_sequences(X_test)
test_sequences_matrix = pad_sequences(test_sequences,maxlen=max_len)
test_pred = model.predict(test_sequences_matrix)
test_pred = np.round(test_pred,0)
accr = model.evaluate(test_sequences_matrix,Y_test)
```

→ Test The Model