Assignment-4 SMS SPAM Classification

Assignment Date	31 oct 2022
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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.

1. Download the dataset:

https://www.kaggle.com/code/kredy10/simple-lstm-for-text-classification/data

2. Importe Required libararies

```
In [9]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import keras
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, pad_sequences
from keras.callbacks import EarlyStopping
%matplotlib inline
```

3. Read dataset and pre-processing

```
In [2]:
```

```
df = pd.read_csv('spam.csv', delimiter=',', encoding='latin-1')
df.head()
```

Out[2]:

	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

```
In [3]:
```

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

```
In [4]:
```

```
df.shape
Out[4]:
(5572, 2)
```

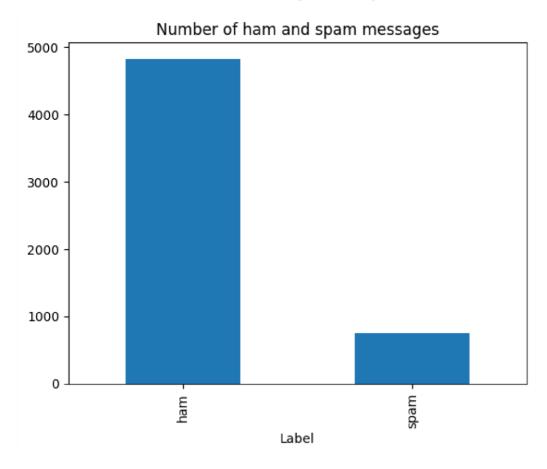
In [5]:

```
#plot the ham and spam messages to understand the distribution
df['v1'].value_counts().plot(kind='bar')
plt.xlabel('Label')
```

plt.title('Number of ham and spam messages')

Out[5]:

Text(0.5, 1.0, 'Number of ham and spam messages')



In [6]:

```
X = df.v2
Y = df.v1
#label encoding for Y
le = LabelEncoder()
Y = le.fit_transform(Y)
Y = Y.reshape(-1,1)
```

Train-test split

In [7]:

```
#split into train and test sets
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.20)
```

Tokenizer

In [10]:

```
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 = keras.utils.pad_sequences(sequences, maxlen=max_len)
```

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

In [11]:

```
inputs = Input(name='inputs', shape=[max_len])
layer = Embedding(max_words, 50, input_length=max_len) (inputs)
```

```
layer = LSTM(64)(layer)
layer = Dense(256, name='FC1')(layer)
layer = Activation('relu')(layer)
layer = Dropout(0.5)(layer)
layer = Dense(1, name='out_layer')(layer)
layer = Activation('sigmoid')(layer)
```

5 Create Model

```
In [12]:
```

```
model = Model(inputs=inputs,outputs=layer)
```

6 Compile the Model

In [13]:

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

Model: "model"

Layer (type)	Output Shape	Param #
inputs (InputLayer)	[(None, 150)]	0
embedding (Embedding)	(None, 150, 50)	50000
lstm (LSTM)	(None, 64)	29440
FC1 (Dense)	(None, 256)	16640
activation (Activation)	(None, 256)	0
dropout (Dropout)	(None, 256)	0
out_layer (Dense)	(None, 1)	257
activation_1 (Activation)	(None, 1)	0

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

7 Fit the Model

In [15]:

8 Save the Model

```
In [16]:
model.save('spam_lstm_model.h5')
```

9 Test the Model

In []:

Loss: 0.091 Accuracy: 0.978