Assignment – 4

SMS SPAM Classification

Assignment Date	26 October 2022
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Maximum Marks	2 Marks

TASKS:

- 1. Download the dataset
- 2. Import required library

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import tensorflow
import nltk
from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
import string
from tensorflow.keras.preprocessing import sequence
from keras.models import Model, Sequential
from keras.preprocessing.text import Tokenizer
from keras.optimizers import Adam, RMSprop
from keras.layers import Input, Embedding, LSTM, Dense, Flatten, Dropout
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split
```

3. Read dataset and do Pre-processing

Read Dataset

```
df = pd.read_csv(r"C:\Users\manok\Documents\Sem_\HX5001-HX6001\Assignment\Assignment_4\spam.csv", encoding='latin-1')
df.head()
                                               v2 Unnamed: 2 Unnamed: 3 Unnamed: 4
0 ham
           Go until jurong point, crazy.. Available only ...
                                                          NaN
                                                                       NaN
                                                                                    NaN
                            Ok lar... Joking wif u oni...
                                                                       NaN
                                                                                    NaN
1 ham
                                                          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
         Nah I don't think he goes to usf, he lives aro...
                                                                       NaN
                                                                                    NaN
```

```
df.shape
(5572, 5)
```

Drop Unwanted Column

```
df = df.drop(["Unnamed: 2", "Unnamed: 3", "Unnamed: 4"], axis=1)
df = df.rename(columns={"v2" : "Text", "v1":"Label"})
```

```
df.head()
```

Label		Text
0	ham	Go until jurong point, crazy Available only
1	ham	Ok lar Joking wif u oni
2	spam	Free entry in 2 a wkly comp to win FA Cup fina
3	ham	U dun say so early hor U c already then say
4	ham	Nah I don't think he goes to usf, he lives aro

Remove Duplicate and Null Data

```
df.isnull().sum()

Label  0
Text  0
dtype: int64

df.duplicated().sum()
```

```
df = df.drop_duplicates(keep='first')
df.duplicated().sum()
```

0

```
df.shape
(5169, 2)
```

Normalizing the case, Removing the unwanted punctuations, Remove Stopwords

```
ps = PorterStemmer()
```

```
def transform_text(text):
  text = text.lower()
   text = nltk.word_tokenize(text)
   y = []
   for i in text:
       if i.isalnum():
           y.append(i)
   text = y[:]
   y.clear()
   for i in text:
       if i not in stopwords.words('english') and i not in string.punctuation:
           y.append(i)
   text = y[:]
   y.clear()
   for i in text:
       y.append(ps.stem(i))
   return " ".join(y)
```

```
df['Transformed_Text'] = df['Text'].apply(transform_text)
```

```
df.head()
```

	Label	Text	Transformed_Text
0	ham	Go until jurong point, crazy Available only	go jurong point crazi avail bugi n great world
1	ham	Ok lar Joking wif u oni	ok lar joke wif u oni
2	spam	Free entry in 2 a wkly comp to win FA Cup fina	free entri 2 wkli comp win fa cup final tkt 21
3	ham	U dun say so early hor U c already then say	u dun say earli hor u c alreadi say
4	ham	Nah I don't think he goes to usf, he lives aro	nah think goe usf live around though

Counting Words

```
avg_words_len=round(sum([len(i.split()) for i in df['Text']])/len(df['Text']))
print(avg_words_len)
# avg_words_len=200
```

15

```
s = set()
for sent in df['Transformed_Text']:
    for word in sent.split():
        s.add(word)
total_words_length=len(s)
print(total_words_length)
# total_words_length=2000
```

4. Create model

```
x = df.Transformed_Text
y = df.Label
le = LabelEncoder()
y = le.fit_transform(y)
y = y.reshape(-1,1)

# y = df['Label'].values

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.18, random_state=10)
x_train.shape, y_train.shape, x_test.shape, y_test.shape

((4238,), (4238, 1), (931,), (931, 1))

model = Sequential()
```

5. Add layers

```
tokenizer = Tokenizer(num_words = total_words_length, lower = True)
tokenizer.fit_on_texts(x_train)
sequences = tokenizer.texts_to_sequences(x_train)
x_train = sequence.pad_sequences(sequences, maxlen = avg_words_len)
```

Input Layer

```
# model.add(Input(shape=(1), dtype=tf.string))
# model.add(Input(name='inputs',shape=[avg_words_len]))

model.add(Embedding(total_words_length, 50, input_length = avg_words_len))
```

LSTM Layer

```
model.add(LSTM(64))
```

Hidden Layer

```
model.add(Dense(64, activation = "relu"))

model.add(Flatten())

model.add(Dropout(0.2))

model.add(Dense(32, activation = "relu"))
```

Output Layer

```
model.add(Dense(1, activation = 'sigmoid'))
```

Model Summary

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 15, 50)	336800
lstm (LSTM)	(None, 64)	29440
dense (Dense)	(None, 64)	4160
flatten (Flatten)	(None, 64)	0
dropout (Dropout)	(None, 64)	0
dense_1 (Dense)	(None, 32)	2080
dense_2 (Dense)	(None, 1)	33

Total params: 372,513 Trainable params: 372,513 Non-trainable params: 0

6. Compile the model

```
adam = Adam(learning_rate = 0.001, beta_1 = 0.85, beta_2 = 0.97, epsilon = 1e-07)
model.compile(loss = "binary_crossentropy", optimizer = adam, metrics = ["accuracy"])
```

7. Fit the model

```
history = model.fit(x_train, y_train, epochs = epochs, validation_steps=0.18, batch_size=10)
Epoch 1/5
ccuracy: 0.9552
Epoch 2/5
curacy: 0.9887
Epoch 3/5
curacy: 0.9941
Epoch 4/5
       424/424 [=====
curacy: 0.9969
Epoch 5/5
curacy: 0.9988
```

8. Save the model

```
model.save("spam_analysis.h5")
```

review2 = ["Go until jurong point"]

1/1 [=======] - 0s 46ms/step

predict(review2)

Harm

```
9. Test the model
  test_sequences = tokenizer.texts_to_sequences(x_test)
  x_test = sequence.pad_sequences(test_sequences, maxlen=avg_words_len)
  accuracy = model.evaluate(x_test, y_test)
 30/30 [============== ] - 2s 10ms/step - loss: 0.2072 - accuracy: 0.9731
  def predict(message):
     txt = tokenizer.texts_to_sequences(message)
     txt = sequence.pad_sequences(txt, maxlen=avg_words_len)
     pred = model.predict(txt)
     if pred>0.5:
         print("spam")
         print("Harm")
  review1 = ["think he goes"]
  predict(review1)
 1/1 [======] - 1s 1s/step
 Harm
```