

## Assignment – 4

### SMS SPAM Classification

Assignment Date	11 November 2022
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Maximum Marks	2 Marks

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_7\HX5001-HX6001\Assignment\Assignment_4\spam.csv", encoding='latin-1')
```

```
df.head()
```

	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

```
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()
```

```
403
```

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

```
0
```

```
df.shape
```

```
(5169, 2)
```

```
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 already 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

```
epochs=5
history = model.fit(x_train, y_train, epochs = epochs, validation_steps=0.18, batch_size=10)
```

Epoch 1/5

424/424 [=====] - 16s 14ms/step - loss: 0.1346 - accuracy: 0.9552

Epoch 2/5

424/424 [=====] - 6s 15ms/step - loss: 0.0356 - accuracy: 0.9887

Epoch 3/5

424/424 [=====] - 6s 15ms/step - loss: 0.0203 - accuracy: 0.9941

Epoch 4/5

424/424 [=====] - 6s 14ms/step - loss: 0.0096 - accuracy: 0.9969

Epoch 5/5

424/424 [=====] - 6s 15ms/step - loss: 0.0043 - accuracy: 0.9988

## 8. Save the model

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

## 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")
    else:
        print("Harm")
```

```
review1 = ["think he goes"]
predict(review1)
```

1/1 [=====] - 1s 1s/step  
Harm

```
review2 = ["Go until jurong point"]
predict(review2)
```

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