

## Assignment-4

### SMS SPAM Classification

Assignment Date	:	28 October 2022
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Maximum Marks	:	2 Marks

#### Task 1:

Download the dataset

[Download Dataset](#)

#### Task 2:

##### Question- 1:

**Import the necessary libraries**

Solution:

```
import pandas as pd
import numpy as np
from keras import utils
import matplotlib.pyplot as plt
import seaborn as sns
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
import matplotlib
import inline
```

---

Output:

1. Import the necessary libraries

```
In [1]: import pandas as pd
import numpy as np
from keras import utils
import matplotlib.pyplot as plt
import seaborn as sns
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
%matplotlib inline
```

Solution:

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

dataframe

Output:

```
In [2]: dataframe = pd.read_csv('spam.csv', delimiter=',', encoding='latin-1')
dataframe
```

```
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
...	...	...	...	...	...
5567	spam	This is the 2nd time we have tried 2 contact u...	NaN	NaN	NaN
5568	ham	Will l_b going to esplanade fr home?	NaN	NaN	NaN
5569	ham	Pity, * was in mood for that. So...any other s...	NaN	NaN	NaN
5570	ham	The guy did some bitching but I acted like i'd...	NaN	NaN	NaN
5571	ham	Rofl. Its true to its name	NaN	NaN	NaN

5572 rows × 5 columns

## Task 3:

### Question- 2:

## Preprocessing

Solution:

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

```
dataframe
```

Output:

```
2)Preprocessing

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

Out[3]:
```

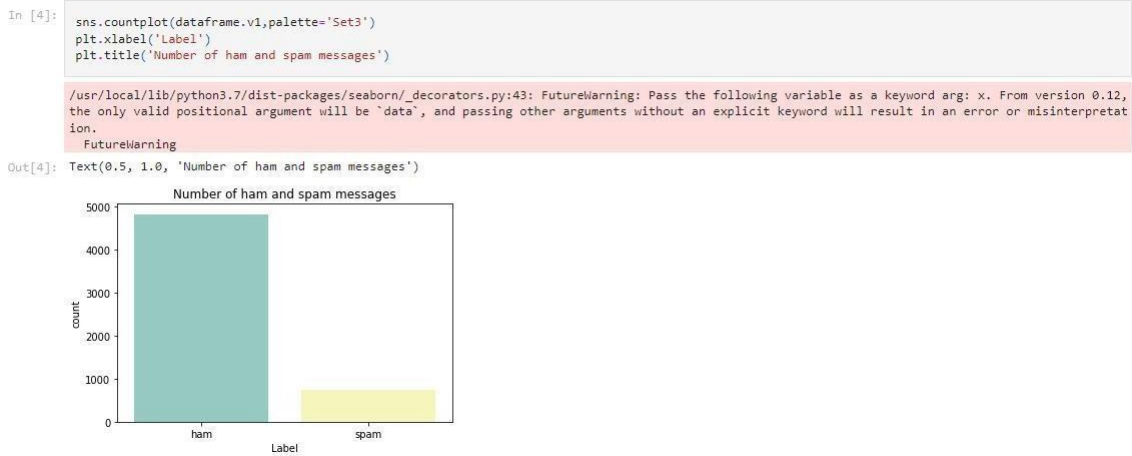
	v1	v2
0	ham	Go until jurong point, crazy.. Available only ...
1	ham	Ok lar... Joking wif u oni...
2	spam	Free entry in 2 a wkly comp to win FA Cup fina...
3	ham	U dun say so early hor... U c already then say...
4	ham	Nah I don't think he goes to usf, he lives aro...
...	...	...
5567	spam	This is the 2nd time we have tried 2 contact u...
5568	ham	Will _b going to esplanade fr home?
5569	ham	Pity, * was in mood for that. So...any other s...
5570	ham	The guy did some bitching but I acted like i'd...
5571	ham	Rofl. Its true to its name

5572 rows × 2 columns

Solution:

```
sns.countplot(dataframe.v1,palette='Set3')  
plt.xlabel('Label')  
plt.title('Number of ham and spam messages')
```

Output:



Solution:

```
X = dataframe.v2
Y = dataframe.v1
le = LabelEncoder()
Y = le.fit_transform(Y)
Y = Y.reshape(-1,1)
```

Output:

```
In [5]: X = dataframe.v2
Y = dataframe.v1
le = LabelEncoder()
Y = le.fit_transform(Y)
Y = Y.reshape(-1,1)
```

## Task 4:

### Question- 3:

### Split into training and test data

Solution:

```
X_train,X_test,Y_train,Y_test =  
train_test_split(X,Y,test_size=0.15)
```

Output:

3)Split into training and test data.

```
In [6]: X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.15)
```

Solution:

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

Output:

```
In [7]: 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 = utils.pad_sequences(sequences,maxlen=max_len)
```

Solution:

```
sequences_matrix.shape
```

---

Output:

```
In [8]: sequences_matrix.shape
```

```
Out[8]: (4736, 150)
```

Solution:

```
sequences_matrix.ndim
```

Output:

```
In [9]: sequences_matrix.ndim
```

```
Out[9]: 2
```

Solution:

```
sequences_matrix =  
    np.reshape(sequences_matrix, (4736, 150, 1))
```

Output:

```
In [10]: sequences_matrix = np.reshape(sequences_matrix, (4736, 150, 1))
```

Solution:

```
sequences_matrix.ndim #3d shape verification to  
    proceed to RNN LSTM
```

Output:

```
In [11]: sequences_matrix.ndim #3d shape verification to proceed to RNN LSTM
```

```
Out[11]: 3
```

---

## Task 5:

### Question- 4:

## Create model for RNN

Solution:

```
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import LSTM
from keras.layers import Embedding
```

Output:

4)Create model for RNN

```
In [12]: from keras.models import Sequential
from keras.layers import Dense
from keras.layers import LSTM
from keras.layers import Embedding
```

Solution:

```
model = Sequential()
```

Output:

```
In [13]: model = Sequential()
```

---

## Task 6:

### Question- 5:

## Add Layers

Solution:

```
model.add(Embedding(max_words, 50, input_length=max_len))
```

Output:

5)Add Layers

```
In [14]: model.add(Embedding(max_words, 50, input_length=max_len))
```

Solution:

```
model.add(LSTM(units=64, input_shape =  
              (sequences_matrix.shape[1], 1), return_sequences=True))
```

Output:

```
In [15]: model.add(LSTM(units=64, input_shape = (sequences_matrix.shape[1], 1), return_sequences=True))
```

Solution:

```
model.add(LSTM(units=64, return_sequences=True))
```

Output:

```
In [16]: model.add(LSTM(units=64, return_sequences=True))
```

---



Solution:

```
model.add(LSTM(units=64, return_sequences=True))
```

Output:

```
In [17]: model.add(LSTM(units=64, return_sequences=True))
```

Solution:

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

Output:

```
In [18]: model.add(LSTM(units=64))
```

Solution:

```
model.add(Dense(units = 256, activation = 'relu'))
```

Output:

```
In [19]: model.add(Dense(units = 256, activation = 'relu'))
```

Solution:

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

Output:

```
In [20]: model.add(Dense(units = 1, activation = 'sigmoid'))
```

---

## Task 7:

### Question- 6:

## Compile the model

Solution:

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

Output:

6) Compile the model

In [21]:

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

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
embedding (Embedding)	(None, 150, 50)	50000
lstm (LSTM)	(None, 150, 64)	29440
lstm_1 (LSTM)	(None, 150, 64)	33024
lstm_2 (LSTM)	(None, 150, 64)	33024
lstm_3 (LSTM)	(None, 64)	33024
dense (Dense)	(None, 256)	16640
dense_1 (Dense)	(None, 1)	257
=====		
Total params: 195,409		
Trainable params: 195,409		
Non-trainable params: 0		

Fit the model

Solution:

```
modelf =  
    model.fit(sequences_matrix, Y_train, batch_size=128,  
e  
    pochs=10, validation_split=0.2) modelf
```

Output:

```
In [22]: modelf = model.fit(sequences_matrix, Y_train, batch_size=128, epochs=10, validation_split=0.2)  
modelf  
  
Epoch 1/10  
30/30 [=====] - 45s 1s/step - loss: 0.4619 - accuracy: 0.8456 - val_loss: 0.4486 - val_accuracy: 0.8460  
Epoch 2/10  
30/30 [=====] - 35s 1s/step - loss: 0.3779 - accuracy: 0.8728 - val_loss: 0.4079 - val_accuracy: 0.8460  
Epoch 3/10  
30/30 [=====] - 34s 1s/step - loss: 0.2314 - accuracy: 0.9116 - val_loss: 0.1126 - val_accuracy: 0.9662  
Epoch 4/10  
30/30 [=====] - 39s 1s/step - loss: 0.0753 - accuracy: 0.9794 - val_loss: 0.0773 - val_accuracy: 0.9778  
Epoch 5/10  
30/30 [=====] - 37s 1s/step - loss: 0.0471 - accuracy: 0.9879 - val_loss: 0.0545 - val_accuracy: 0.9842  
Epoch 6/10  
30/30 [=====] - 34s 1s/step - loss: 0.0331 - accuracy: 0.9913 - val_loss: 0.0506 - val_accuracy: 0.9863  
Epoch 7/10  
30/30 [=====] - 34s 1s/step - loss: 0.0253 - accuracy: 0.9939 - val_loss: 0.0446 - val_accuracy: 0.9916  
Epoch 8/10  
30/30 [=====] - 34s 1s/step - loss: 0.0210 - accuracy: 0.9950 - val_loss: 0.0572 - val_accuracy: 0.9852  
Epoch 9/10  
30/30 [=====] - 34s 1s/step - loss: 0.0171 - accuracy: 0.9968 - val_loss: 0.0495 - val_accuracy: 0.9895  
Epoch 10/10  
30/30 [=====] - 34s 1s/step - loss: 0.0124 - accuracy: 0.9976 - val_loss: 0.0536 - val_accuracy: 0.9916  
  
Out[22]:
```

## Task 8:

Question- 7:

### Save the model

Solution:

```
model.save
```

Output:

7) Save the model

```
In [23]: model.save
```

```
Out[23]: >
```

## Task 9:

Question- 8:

### Testing the model

Solution:

```
test_sequences = tok.texts_to_sequences(X_test)
test_sequences_matrix =
utils.pad_sequences(test_sequences, maxlen=max_len)
```

Output:

8. Testing the model

```
In [24]: test_sequences = tok.texts_to_sequences(X_test)
test_sequences_matrix = utils.pad_sequences(test_sequences, maxlen=max_len)
```

Solution:

```
accr = model.evaluate(test_sequences_matrix, Y_test)
```

Output:

```
In [25]: accr = model.evaluate(test_sequences_matrix, Y_test)
27/27 [=====] - 6s 176ms/step - loss: 0.1019 - accuracy: 0.9809
```

Solution:

```
l = accr[0] a =accr[1] print('Test set\n Loss:
{:0.3f}\n Accuracy:
{:0.3f}'.format(l, a))
```

Output:

```
In [26]: l = accr[0]
a = accr[1]
print('Test set\n Loss: {:.3f}\n Accuracy: {:.3f}'.format(l,a))

Test set
Loss: 0.102
Accuracy: 0.981

Accuracy and Loss Graph
```

Solution:

```
Results = pd.DataFrame({"Train Loss":
                        model.history['loss'], "Validation Loss":
                        model.history['val_loss'],
                        "Train Accuracy":
                        model.history['accuracy'], "Validation
                        Accuracy": model.history['val_accuracy']
                        }) fig, ax = plt.subplots(nrows=2,
figsize=(16, 9)) results[["Train Loss", "Validation
Loss"]].plot(ax=ax[0]) results[["Train Accuracy",
"Validation
Accuracy"]].plot(ax=ax[1])
ax[0].set_xlabel("Epoch")

ax[1].set_xlabel("Epoch")

plt.show()
```

Output:

```
In [27]: results = pd.DataFrame({"Train Loss": model.history['loss'], "Validation Loss": model.history['val_loss'],
                                "Train Accuracy": model.history['accuracy'], "Validation Accuracy": model.history['val_accuracy']
                                })
fig, ax = plt.subplots(nrows=2, figsize=(16, 9))
results[["Train Loss", "Validation Loss"]].plot(ax=ax[0])
results[["Train Accuracy", "Validation Accuracy"]].plot(ax=ax[1])
ax[0].set_xlabel("Epoch")
ax[1].set_xlabel("Epoch")
plt.show()
```

