

## Assignment - 4

### Python Programming

Assignment Date	27 October 2022
Student Name	Ms.D.Janani
Student Roll Number	510119106004
Maximum Marks	2 Marks

#### QUESTION: SMS Spam Classification

```
import pandas as pd
import numpy as np
import sklearn as sk
import tensorflow
```

```
data=pd.read_csv("/content/drive/MyDrive/LSTM/simple-lstm-for-text-classification.ipynb")
```

```
data.head()
```

Empty DataFrame

Columns:

```
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```

```

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dataframe", outputs:[].1, execution_count:null}.1,
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_uuid:"aca2f1d9da3f35d104763166fe4d25448410d8f2"}}, cell_type:"code".1,
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encoding='latin-1')\ndf.head()", execution_count:null.1,
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for the neural network.", outputs:[].2, execution_count:null}.2,
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inplace=True)\ndf.info()", execution_count:null.2, outputs:[].2,
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source:"sns.countplot(df.v1)\nplt.xlabel('Label')\nplt.title('Number
of ham and spam messages')", execution_count:null.3, outputs:[].3,
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Process the labels.", outputs:[].4, execution_count:null}.4,
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_uuid:"a1a345c1683e2fcc7173ecae867a5da87f2dde24"}}, cell_type:"code".4,
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le.fit_transform(Y)\nY = Y.reshape(-1, 1)", execution_count:null.4,
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source:"X_train, X_test, Y_train, Y_test = train_test_split(X, Y,
test_size=0.15)", execution_count:null.5, outputs:[].5,
{"metadata":{"_uuid":"c5378d55c271e01480c1ac07f94ff99a80f900d6"}},
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data and convert the text to sequences.\n* Add padding to ensure that
all the sequences have the same shape.\n* There are many ways of
taking the *max_len* and here an arbitrary length of 150 is chosen.",
outputs:[].6, execution_count:null}.6, {"metadata":{"trusted":true.3,
collapsed:true.1, _uuid:"bdca14f2b8cd7bd7cb5ee66fd40ea522217c03c6"}},
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Tokenizer(num_words=max_words)\ntok.fit_on_texts(X_train)\nsequences =
tok.texts_to_sequences(X_train)\nsequences_matrix =
sequence.pad_sequences(sequences, ...]

```

```
Index: []
```

```
[0 rows x 202 columns]
```

```
train_set = data.iloc[:,1:2].values
```

```
train_set
```

```
array([], shape=(0, 1), dtype=object)
```

```
len(train_set)
```

```
0
```

```
x_train = []
```

```
y_train = []
```

```
for i in range(60,0):  
    x_train.append(train_set[i-60:i,0])  
    y_train.append(train_set[i,0])
```

```
x_train
```

```
[array([], dtype=object)]
```

```
len(x_train)
```

```
1
```

```
x_train.ndim
```

```
2
```

```
x_train = np.reshape(x_train,(1190,60,1))
```

```
x_train.ndim
```

```
3
```

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM,Dense
```

```
model = Sequential()
```

```
model.add(LSTM(units = 50,input_shape
=(x_train.shape[1],1),return_Sequential = True ))
```

```
model.add(LSTM(units = 50, return_sequences = True))
```

```
model.add(LSTM(units = 50, return_sequences = True))
```

```
model.add(LSTM(units = 50))
```

```
model.add(LSTM(units = 1))
```

```
model.compile(optimizer = "adam", loss = "mse")
```

```
model.fit(x_train, y_train, epochs = 5, batch_size = 64))
```

^

```
pred = model.predict(x_train)
```

```
y_train
```

```
[]
```

```
#RMSE
```

```
error = y_train - pred
```

```
SE = error*error
```

```
MSE = SE.mean()
```

```
MSE
```

```
model.add(Dense(units = 1565, activation "relu"))
```

```
model.add(Dense(units = 2000, activation "relu"))
model.add(Dense(units = 1, activation "sigmoid"))
model.compile(optimizer = 'adam', loss =
'binary_crossentropy',metrics = ['accuracy'])
model.fit(x_train,y_train,epochs = 10)
```

```
text1 = "I hate this food and it was very bad"
text2 = "I love this food and it is very tasty"
```

```
text1.split()
```

```
['I', 'hate', 'this', 'food', 'and', 'it', 'was', 'very', 'bad']
```

```
def preprocessing(text1)
    text = re.sub('[^a-zA-Z]', ' ',text1)
    text = text.split()
    text = [ps.stem(word) for word in text if not word in set
(stopwords.words('english'))]
    text ' '.join(text)
    return text
```

```
preprocessing (text1)
```

```
model.save("text_classification")
```

```
x_train()
```

```
[array([], dtype=object)]
```

```
y_train()
```

```
[]
```