Assignment_4

November 17, 2022

Assignment Date	08 October 2022		
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Maximum Marks	2 Marks		

1.Download the dataset

2.Import required library

[4]: import pandas as pd import numpy as np 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 pad_sequences from keras.utils import to categorical from keras.callbacks import EarlyStopping J. Neua Dataset and do preprocessing

[6]: data=pd.read_csv('/content/spam.csv',encoding='latin')

[7]: df =pd.read_csv('/content/spam.csv',delimiter=',',encoding='latin-1') df.head()

[7]:		v1		v2 Unnamed: 2	\
	0	ham	Go until jurong point, crazy Available only	Na	
				N	
	1	ham	Ok lar Joking wif u oni	NaN	
	2	spam	Free entry in 2 a wkly comp to win FA Cup fina	Na	
				N	
	3	ham	U dun say so early hor U c already then say	NaN	
	4	ham	Nah I don't think he goes to usf, he lives aro	Na	
				N	

```
0
                                NaN
                   Na
                   Ν
       1
                   Na
                                NaN
                   Ν
       2
                   Na
                                NaN
                   Ν
       3
                   Na
                                NaN
                    Ν
       4
                   Na
                                NaN
                    Ν
       df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'],axis=1,inplace=True) df.info()
 [8]:
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 5572 entries, 0 to 5571 Data
      columns (total 2 columns):
             Column Non-Null Count
                                          Dtype
        0
                       5572 non-null
             v1
                                          object
        1
             v2
                       5572 non-null
                                          object
      dtypes: object(2) memory
      usage: 87.2+ KB
       # Count of Spam and Ham values
 [9]:
       df.groupby(['v1']).size()
 [9]: v1
                 4825
       ham
                  747
       spam
       dtype: int64
[10]:
       # Label Encoding target column
       X = df.v2 Y =
       df.v1
       le = LabelEncoder()
       Y = le.fit transform(Y) Y =
       Y.reshape(-1,1)
[11]:
       # Test and train split
       X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.15)
[12]:
       # Tokenisation function
       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 =
       pad_sequences(sequences,maxlen=max_len)
```

Unnamed: 3 Unnamed: 4

4. Create Model and 5. Add Layers (LSTM, Dense-(Hidden Layers), Output)

[13]: # Creating LSTM model 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) model = Model(inputs=inputs,outputs=layer)

6.Compile the model & 7.Fit the Model

[14]: model.summary()

model.

Model: "model"

Layer (type)	Output Shape	Param #
inputs (InputLayer)	[(None, 150)]	0
embedding (Embedding)	(None, 150, 50)	50000
Istm (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

Epoch 1/10

30/30 [===========] - 12s 311ms/step - loss: 0.3269 -

accuracy: 0.8746 - val_loss: 0.1554 - val_accuracy: 0.9778

```
Epoch 2/10
     30/30 [============] - 12s 380ms/step - loss: 0.0819 -
     accuracy: 0.9794 - val loss: 0.0461 - val accuracy: 0.9831 Epoch 3/10
     accuracy: 0.9873 - val loss: 0.0343 - val accuracy: 0.9905 Epoch 4/10
     30/30 [============= ] - 8s 261ms/step - loss: 0.0370 -
     accuracy: 0.9894 - val loss: 0.0447 - val accuracy: 0.9895 Epoch 5/10
     30/30 [============ - - 9s 310ms/step - loss: 0.0276 -
     accuracy: 0.9918 - val loss: 0.0340 - val accuracy: 0.9905 Epoch 6/10
     30/30 [============ - - 8s 260ms/step - loss: 0.0224 -
     accuracy: 0.9929 - val loss: 0.0402 - val accuracy: 0.9905 Epoch 7/10
     30/30 [============== ] - 8s 260ms/step - loss: 0.0159 -
     accuracy: 0.9958 - val_loss: 0.0442 - val_accuracy: 0.9916 Epoch 8/10
     accuracy: 0.9960 - val loss: 0.0433 - val accuracy: 0.9905 Epoch 9/10
     30/30 [============== ] - 8s 261ms/step - loss: 0.0108 -
     accuracy: 0.9974 - val loss: 0.0952 - val accuracy: 0.9736 Epoch 10/10
     accuracy: 0.9979 - val loss: 0.0607 - val accuracy: 0.9884 [14]:
<keras.callbacks.History at 0x7f823de6acd0>
     8. Save the Model
     model.save('sms classifier.h5')
     9. Test the model
     test sequences = tok.texts to sequences(X test) test sequences matrix =
     pad sequences(test sequences,maxlen=max len)
     accr = model.evaluate(test sequences matrix,Y test)
     0.9856
     print('Test set\n
                        Loss: {:0.3f}\n
                                        Accuracy:{:0.3f}'.format(accr[0],accr[1]))
     Test set Loss:
       0.104
```

Accuracy: 0.986

[15]:

[16]:

[17]:

[18]: