**Assignment - 4**

**LSTM for Text Classification**

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| Assignment submission | 4 November 2022 |
| Student Name | Kameshwari R |
| Student Roll Number | 951920LCS02 |
| Maximum Marks | 2 Marks |

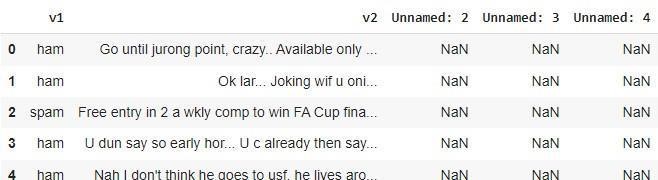
**1. Download the Dataset 2. Import required library** import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns from sklearn.model\_selection import train\_test\_split from sklearn.preprocessing import LabelEncoder from keras.models import Model from keras.layers import LSTM, Activation, Dense, Dropout, Input, Embedding from keras.optimizers import RMSprop from keras.preprocessing.text import Tokenizer from keras.utils import pad\_sequences from keras.utils import

to\_categorical from keras.callbacks import EarlyStopping

%matplotlib inline

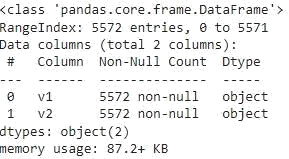
**3. Read dataset and do pre-processing** Load the data into Pandas dataframe df = pd.read\_csv('/content/spam.csv',delimiter=',',encoding='latin-1')

df.head()



Drop the columns that are not required for the neural network.

df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'],axis=1,inplace=True) df.info() pH

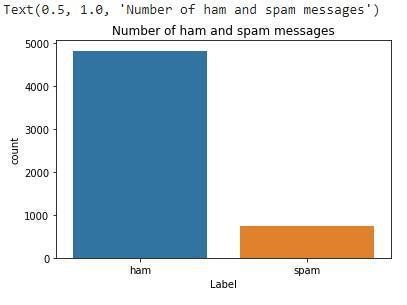


Understand the distribution better.

sns.countplot(df.v1) plt.xlabel('Label')

plt.title('Number of ham and spam messages')

Create input and output vectors.



Process the labels.

1. = df.v2 Y = df.v1 le = LabelEncoder()

1. = le.fit\_transform(Y)

Y = Y.reshape(-1,1)

Split into training and test data.

X\_train,X\_test,Y\_train,Y\_test = train\_test\_split(X,Y,test\_size=0.15)

Process the data

* Tokenize the data and convert the text to sequences.
* Add padding to ensure that all the sequences have the same shape.
* There are many ways of taking the \*max\_len\* and here an arbitrary length of 150 is chosen.

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)

**5. Create Model**

● **Add Layers (LSTM, Dense-(Hidden Layers), Output)**

Define the RNN structure. def

RNN():

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 =

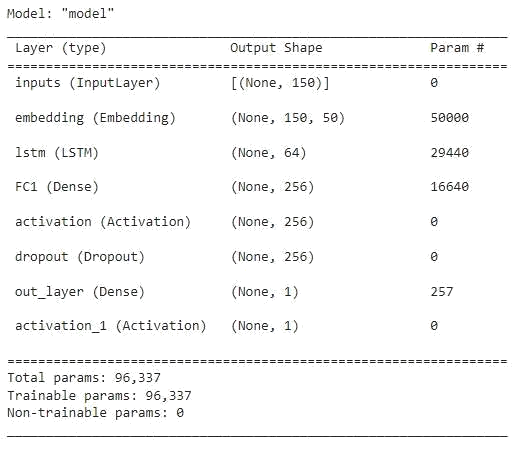
Activation('sigmoid')(layer) model =

Model(inputs=inputs,outputs=layer) return model

Call the function and compile the model.

model = RNN() model.summary() **6. Compile the Model**

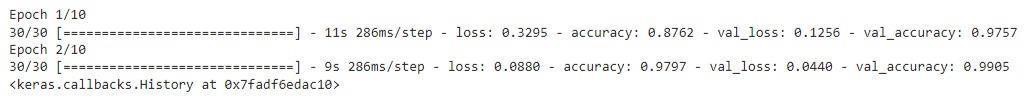
model.compile(loss='binary\_crossentropy',optimizer=RMSprop(),metrics=['accuracy'])



1. **Fit the Model**

model.fit(sequences\_matrix,Y\_train,batch\_size=128,epochs=10, validation\_split=0.2,callbacks=[EarlyStopping(monitor='val\_loss',min\_d

elta=0.0001)])



The model performs well on the validation set and this configuration is chosen as the final model.

1. **Save The Model** lstm\_model.save('text\_model.h5')

1. **Test The Model** test\_sequences =

tok.texts\_to\_sequences(X\_test) test\_sequences\_matrix

=pad\_sequences(test\_sequences,maxlen=max\_len)

Evaluate the model on the test set.

accr = model.evaluate(test\_sequences\_matrix,Y\_test)



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

