ASSIGNMENT 4

Date	31 October 2022
Team ID	PNT2022TMID43225
Project Name	Deep learning fundus image analysis for early detection of diabetic retinopathy.

```
Automatically generated by Colaboratory.
Original file is located at
    https://colab.research.google.com/drive/1BuzOmo9BAXkrsPk9gOKeCV1JA rd4 Vy
**Import the required libraries**
# Commented out IPython magic to ensure Python compatibility.
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
sns.set style("darkgrid")
# %matplotlib inline
import string
import nltk
from nltk.corpus import stopwords
from wordcloud import WordCloud
from sklearn.feature extraction.text import CountVectorizer
from nltk.stem import WordNetLemmatizer
from sklearn.model selection import train test split
from sklearn import metrics
"""**Read** **the Dataset**
messages = pd.read_csv('../content/spam.csv',encoding = 'latin-1')
messages.head()
"""**Preprocessing the data**
** ** **
messages.isnull()
messages.isnull().sum()
messages.isnull().sum()/len(messages)
messages.dropna(axis = 1)
messages.fillna(0)
messages.mean()
messages.mode()
messages.median()
messages.std()
messages.describe()
```



```
messages.describe(include='all')
messages.drop(["Unnamed: 2", "Unnamed: 3", "Unnamed: 4"], axis=1, inplace=True)
```

```
messages.columns = ["SpamHam","Tweet"]
sns.countplot(messages["SpamHam"])
 import nltk
 nltk.download('stopwords')
 from sklearn.preprocessing import LabelEncoder
 lb enc = LabelEncoder()
 y = lb enc.fit transform(messages["SpamHam"])
 import pandas as pd
 import numpy as np
 import re
 import collections
 import contractions
 import seaborn as sns
 import matplotlib.pyplot as plt
 plt.style.use('dark background')
 import nltk
 from nltk.stem import WordNetLemmatizer
 from nltk.corpus import stopwords
 import warnings
 warnings.simplefilter(action='ignore', category=Warning)
 import keras
 from keras.layers import Dense, Embedding, LSTM, Dropout
 from keras.models import Sequential
 from keras.preprocessing.text import Tokenizer
 import pickle
 pip install contractions
 pip install pad sequences
 from sklearn.preprocessing import LabelEncoder
 lb enc = LabelEncoder()
 y = lb enc.fit transform(messages["SpamHam"])
 tokenizer = Tokenizer() #initializing the tokenizer
 tokenizer.fit on texts(messages) # fitting on the sms data
 text to sequence = tokenizer.texts to sequences(messages)
 import numpy as np # linear algebra
 import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
 import matplotlib.pyplot as plt
 # Input data files are available in the "../input/" directory.
 # For example, running this (by clicking run or pressing Shift+Enter) will list the files in
 the input directory
 # we need to fit model with sequence of tokens with specific lengt
 from keras.preprocessing.text import Tokenizer
 from keras.models import Sequential
 # normal LSTM/GRU and the Version with Cuda
 from keras.layers import Dense, Embedding, GRU, LSTM, Dropout, Bidirectional
 from keras.callbacks import TensorBoard, EarlyStopping, ModelCheckpoint
 # keras wrapper for k-fold cross-validation
 from keras.wrappers.scikit learn import KerasClassifier
 # normsl cross validation
 from sklearn.model selection import cross val score, train test split
 # cross validation for hyperparameter tuning
 from sklearn.model selection import GridSearchCV
```

```
x raw = []
y_raw = []
with open ("spam.csv", encoding = "ISO-8859-1") as f:
    for line in f:
        y raw.append(line.split()[0])
        x raw.append(' '.join(i for i in line.split()[1:]))
y = [1 if i=='ham' else 0 for i in y raw]
print(max(len(s) for s in x raw))
print(min(len(s) for s in x raw))
sorted_X = sorted(len(s) for s in x_raw)
print(sorted X[len(sorted X) // 2])
tokenizer = Tokenizer()
tokenizer.fit on texts(x raw)
sequences = tokenizer.texts to sequences(x raw)
vocab size = len(tokenizer.word index)+1
print(vocab_size)
sum([len(x) for x in sequences])
X train, X test, y train, y test = train test split(sequences, y, test size = 0.2,
random state= 0)
"""**Create** **model**"""
model = Sequential()
model.add(Embedding(input dim=vocab size, output dim=embedding size, input length=max len))
model.add(Dropout(0.8))
model.add(LSTM(140, return sequences=False))
model.add(Dropout(0.8))
model.add(Dense(1, activation='sigmoid', name='Classification'))
model.summary()
"""**Add Layers (LSTM, Dense-(Hidden Layers), Output)**""
#LSTM hyperparameters
n lstm = 20
drop_lstm =0.2
#LSTM Spam detection architecture
model1 = Sequential()
model1.add(LSTM(n_lstm, dropout=drop_lstm, return_sequences=True))
model1.add(LSTM(n lstm, dropout=drop lstm, return sequences=True))
model1.add(Dense(1, activation='sigmoid'))
model1.compile(loss = 'binary_crossentropy', optimizer = 'adam', metrics=['accuracy'])
"""**Hidden Layer**""
model.add(Dense(300, activation='relu'))
model.add(Dense(150, activation='relu'))
"""**Output Layer**""
model.add(Dense(4,activation='softmax'))
model.compile(loss='categorical crossentropy',optimizer='adam',metrics=['accuracy'])
len(X_train)
"""Text Preprocessing
messages.head()
```

```
messages.keys
def remove url(text):
    re url = re.compile('https?://\S+|www\.\S+')
    return re url.sub('', text)
messages['SpamHam'] = messages['SpamHam'].apply(remove url)
def remove url(text):
    re url = re.compile('https?://\S+|www\.\S+')
    return re url.sub('', text)
messages['SpamHam'] = messages['SpamHam'].apply(remove url)
exclude = string.punctuation
def remove punc(text):
    return text.translate(str.maketrans('', '', exclude))
messages['SpamHam'] = messages['SpamHam'].apply(remove punc)
X = messages["SpamHam"]
y = messages['SpamHam'].values
X train, X test, y train, y test = train test split(X, y, test size= 0.2, random state= 42,
stratify = y)
"""**Compile The Model**"""
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
"""**Fit the model**""
def train model(model):
   model.fit(X_train, y_train)
    y pred = model.predict(X test)
    y prob = model.predict proba(X test)
    accuracy = round(accuracy_score(y_test, y_pred), 3)
    precision score = round(precision score(y test, y pred), 3)
    recall = round(recall_score(y_test, y_pred), 3)
   print(f'Accuracy of the model: {accuracy}')
   print(f'Precision Score of the model: {precision}')
   print(f'Recall Score of the model: {recall}')
    sns.set_context('notebook', font_scale= 1.3)
    fig, ax = plt.subplots(1, 2, figsize = (25, 8))
    ax1 = plot_confusion_matrix(y_test, y_pred, ax= ax[0], cmap= 'YlGnBu')
    ax2 = plot roc(y test, y prob, ax= ax[1], plot macro= False, plot micro= False, cmap=
'summer')
from keras.models import Sequential
from keras.layers import Dense
from sklearn.datasets import make blobs
from sklearn.preprocessing import MinMaxScaler
# generate 2d classification dataset
x, y = make blobs(n samples=100, centers=2, n features=2, random state=1)
scalar = MinMaxScaler()
scalar.fit(x)
x = scalar.transform(x)
# define and fit the final model
model = Sequential()
model.add(Dense(4, input shape=(2,), activation='relu'))
model.add(Dense(4, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
model.compile(loss='binary_crossentropy', optimizer='adam')
model.fit(x, y, epochs=10, verbose=5)
```

```
ls model.save('spam.h5')
"""**Test the model**"""
ls

clf = MultinomialNB()
    clf.fit(X_train, y_train)
    print("Accuracy of Model", clf.score(X_test, y_test)*100,"%")
```

"""**Save the model**""