Project - Sarcasm Detection

Introduction: In this project, I predicted sarcasm in the news headlines from the dataset given with the help of Machine Learning algorithms. I choose two algorithms to detect sarcasm: Logistic Regression and Random Forest. I made this project on Jupyter Notebook. Below are the steps of this project:

Technologies Used: Python, Machine Learning, Logistic Regression, Random Forest, MongoDB, Word Cloud, Matplotlib, Text Mining, NLTK, json, Joblib, etc.

Instructions: There are two files:

- Sarcasm Detection.ipynb (main file)
- userPrediction.ipynb (for inputting new single headline)

After saving the folder as it is, just need to run the files and all the code and it's output/result can be seen in the main file once running. It will take some time to upload the TermDocumentFrequency data file as it is very large and required high RAM and SSD. But except it, everything will work fast and efficiently. Also, the "userPrediction.ipynb" file can be run from the main file(in the last cell).

Code

Word Cloud for top 200 frequent words:

sarcastic = []
notSarcastic = []

#splitting sarcastic and not sarcastic headlines for visualizing wordcloud

if dataset['is sarcastic'][i] == 1:

for i in range (0,len(dataset)):

sarcastic.append(dataset ['headline'][i])

else:

notSarcastic.append(dataset['headline'][i])

#Visualizing sarcastic headlines in wordcloud

wordcloud sarcastic=WordCloud(width=1000,height=500,stopwords=STOPWORDS

```
,background color = 'white').generate(''.join(sarcastic)[0:200])
plt.figure(figsize = (15,8))
plt.imshow(wordcloud sarcastic)
plt.axis('off')
plt.title("Sarcastic", fontsize = 50, fontweight='bold')
plt.show()
1. (Filename – Sarcasm Detection.ipynb)
Step 1: Importing Libraries
#Loading libraries
import pymongo as pymongo
from pymongo import MongoClient
import pprint
import json
import csv
from textblob import TextBlob
import os
import pandas as pd
from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
from textblob.sentiments import NaiveBayesAnalyzer
from nltk.corpus import stopwords
import string
import nltk
import textmining
import matplotlib.pyplot as plt
from wordcloud import WordCloud
```

from wordcloud import WordCloud, STOPWORDS

```
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
import seaborn as sns
from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.ensemble import RandomForestClassifier
import joblib
```

Step 2: Creating database and fetching dataset from the database(MongoDB)

#Saving the dataset in the database, we can also read csv file directly or can fetch the dataset from database(MongoDB)

```
def Database():

# Connect to the MongoDB, change the connection string per your MongoDB environment

Channel = MongoClient(port=27017)

# Set the db object to point to the business database

Select_db = Channel.SarcasmDetection

# Set collection name

Select_collection = Select_db.SarcasmHeadlinesDataset

# Extract data from mongodb

# Create json file

cursor = Select_collection.find()

# Convert json into pandas

result = pd.DataFrame(list(cursor))

return result
```

dataset = Database()

Step 3: Deleting Unnecessary columns from the dataset, checking null values in the dataset if any

```
dataset = dataset.drop(['_id', 'article_link'], axis = 1)
#Checking is there is any null value
dataset.isnull().sum()
#Counting all the values.
dataset.count()
```

Step 4: Pre-processing

```
#stop words and punctuations
stop = set(stopwords.words("english"))
exclude = set(string.punctuation)
#removing stop words, punctuations, numbers from our dataset
def clean(doc):
    stop_free = ''.join([i for i in doc.lower().split() if i not in stop])
    punc_free = ".join([i for i in stop_free if i not in exclude])
    num_free = ".join(i for i in punc_free if not i.isdigit())
    return (num_free)

post_corpus = [clean(dataset.iloc[i,1]) for i in range(0, dataset.shape[0])]
```

Step 5: Visualization(WordCloud)

```
sarcastic.append(dataset['headline'][i])
         else:
              notSarcastic.append(dataset['headline'][i])
#Visualizing sarcastic headlines in wordcloud
wordcloud sarcastic = WordCloud(width = 1000, height = 500, stopwords =
       STOPWORDS, background color = 'white').generate(".join(sarcastic))
plt.figure(figsize = (15,8))
plt.imshow(wordcloud sarcastic)
plt.axis('off')
plt.title("Sarcastic", fontsize = 50, fontweight='bold')
plt.show()
#Visualizing non-sarcastic headlines in wordcloud
wordcloud notSarcastic = WordCloud(width = 1000, height = 500, stopwords =
       STOPWORDS, background color = 'white').generate(".join(notSarcastic))
plt.figure(figsize = (15,8))
plt.imshow(wordcloud notSarcastic)
plt.axis('off')
plt.title("Non-Sarcastic", fontsize = 50, fontweight='bold')
plt.show()
Step 6: Creating Document Term Matrix
# Create document term matrix
tdm = textmining.TermDocumentMatrix()
for i in post corpus:
```

```
tdm.add doc(i)
# Write tdm - saving matrix in csv format in system.
tdm.write csv("TDM DataFRame.csv", cutoff = 1)
# Load dataframe from system.
df = pd.read csv("TDM DataFRame.csv")
Step 7: Prediction
x = df
y = dataset['is sarcastic']
#Splitting data
x train, x test, y train, y test = train test split(x,y, stratify = y, test size = 0.3,
       random state = 101)
#Normalizing the values
y train.value counts(normalize = True)
y test.value counts(normalize = True)
Step 7(a): Logistic Regression
def LR(dataset):
       #SAVING THE MODEL BECAUSE WE DO NOT NEED TO TRAIN THE
       MODEL AGAIN & AGAIN. SO, AFTER TRAINING, SAVE THE MODEL with
       joblib library.
       #TRAINING & PREDICTING
       #logmodel = LogisticRegression()
       #logmodel.fit(x train, y train)
       #joblib.dump(logmodel, 'lr.pkl')
       lr model = joblib.load('lr.pkl')
       lr predictions = lr model.predict(x test)
       return lr predictions
```

```
lr predictions = LR(dataset)
```

```
Accuracy:
```

```
#ACCURACY
def LR accuracy():
         accuracy = accuracy score(y test, lr predictions)
         print("Accuracy:",int(accuracy*100),"%\n")
         #CONFUSION MATRIX
         print("Confusion Matrix:\n", confusion matrix(y test, lr predictions),"\n")
         print("Classification Report:\n",classification report(y test, lr predictions))
         print("HeatMap:\n", sns.heatmap(confusion matrix(y test, lr predictions)))
LR accuracy()
Step 7(b): Random Forest
def RF(dataset):
         \#rf \mod = RandomForestClassifier(n estimators = 20).fit(x train, y train)
         #rf predictions = RF model.predict(x test)
         #SAVING MODEL
         #joblib.dump(rf model, 'rf.pkl')
         rf model = joblib.load('rf.pkl')
         rf predictions = rf model.predict(x test)
         return rf predictions
rf predictions = RF(dataset)
Accuracy:
def RF accuracy():
         #ACCURACY
         RF_accuracy = accuracy_score(y_test, rf_predictions)
```

```
print("Accuracy:",int(RF_accuracy*100),"%\n")
#CONFUSION MATRIX
print("Confusion Matrix:\n", confusion_matrix(y_test, rf_predictions),"\n")
print("Classification Report:\n",classification_report(y_test, rf_predictions)))
print("HeatMap:\n", sns.heatmap(confusion_matrix(y_test, rf_predictions))))
```

RF accuracy()

After applying two algorithms: Logistic Regression & Random Forest, we got predictions from both of the algorithm and found that Logistic Regression is more accurate than Random Forest as:

Logistic Regression: 79%

Random Forest: 74%

Step 8: Running another file to detect sarcasm for single headline entered by the user

%run userPrediction.ipynb

2. (Filename – userPrediction.ipynb)

Step 1: Importing Libraries

#Loading libraries

import pymongo as pymongo

from pymongo import MongoClient

import pprint

import json

import csv

from textblob import TextBlob

import os

```
import pandas as pd

from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer

from textblob.sentiments import NaiveBayesAnalyzer

from nltk.corpus import stopwords

import string

import nltk

import textmining

from sklearn.model_selection import train_test_split

from sklearn.linear_model import LogisticRegression

import joblib
```

Step 2: Connecting Dataset to fetch the dataset stored in it(MongoDB)

#saving the dataset in the database, we can also read csv file directly or can fetch the dataset from database(MongoDB)

```
def Database():

# Connect to the MongoDB, change the connection string per your MongoDB environment
Channel = MongoClient(port=27017)

# Set the db object to point to the business database
Select_db = Channel.SarcasmDetection

# Set collection name
Select_collection = Select_db.SarcasmHeadlinesDataset

# Extract data from mongodb

# Create json file
cursor = Select_collection.find()

# Convert json into pandas
result = pd.DataFrame(list(cursor))
return result
```

```
dataset = Database()
dataset
```

Step 3: Removing unnecessary columns

```
dataset = dataset.drop(['_id', 'article_link'], axis = 1)
dataset
```

Step 4: Selecting subset of the data as data is very large to train and load again & again

```
dataset = dataset.iloc[:2000]
dataset
```

Step 5: Get input from the user to get sarcasm detected for it.

```
def giveInput():
    userInput = input("Text: ")
    return userInput
obj = giveInput()
obj
```

Step 6: PreProcessing – removinf stop words, punctuations & numbers)

```
#stop words and punctuations
stop = set(stopwords.words("english"))
exclude = set(string.punctuation)
#removing stop words, punctuations, numbers from our dataset
def clean(doc):
    stop_free = ''.join([i for i in doc.lower().split() if i not in stop])
    punc_free = ".join([i for i in stop_free if i not in exclude])
```

```
num_free = ".join(i for i in punc_free if not i.isdigit())
return (num_free)

post_corpus = [clean(dataset.iloc[i,1]) for i in range(0, dataset.shape[0])]
```

Step 6: PreProcessing – for single sentence(input) entered by user. And, append it in the post_corpus

```
userInput = clean(obj)
post_corpus.append(userInput)
post_corpus
```

Step 7: Creating Document Term Matrix

```
# Create document term matrix

tdm = textmining.TermDocumentMatrix()

for i in post_corpus:
        tdm.add_doc(i)

# Write tdm - saving matrix in csv format in system.

tdm.write_csv("userPrediction\\userInput.csv", cutoff = 1)

# Load dataframe from system.

df = pd.read_csv("userPrediction\\userInput.csv")

userInput = df.iloc[len(df)-1]

df = df.drop(len(df)-1)
```

Step 8: Spliting Dataset (test, train)

```
x = df
y = dataset['is sarcastic']
```

```
#Splitting data
x train, x test, y train, y test = train test split(x,y, stratify = y, test size = 0.3, random state
= 101)
#Normalizing the values
y train.value counts(normalize = True)
y test.value counts(normalize = True)
Step 9: Prediction – using Logistic Regression
def LR(dataset):
      #SAVING THE MODEL BECAUSE WE DO NOT NEED TO TRAIN THE MODEL
       AGAIN & AGAIN. SO, AFTER TRAINING SAVE THE MODEL.
         #TRAINING & PREDICTING
         logmodel = LogisticRegression()
         logmodel.fit(x train, y train)
         joblib.dump(logmodel, 'userPrediction\\lr.pkl')
         lr model = joblib.load('userPrediction\\lr.pkl')
         lr predictions = lr model.predict(x test)
         return lr predictions
lr predictions = LR(dataset)
Step 10: Getting output (Sarcasm or Not)
def Prediction(lr predictions):
         if lr predictions[-1] == 0:
                  prediction = "Sarcasm is Not Present"
                  print(prediction)
         elif lr predictions[-1] == 1:
                  prediction = "Sarcasm is Present"
```

print(prediction)

else:

print("There is some error!")

Prediction(lr_predictions)