# **TextBlob**

### Loading the Data

```
!gdown --id 15LVlv7K31SzBs_IyZE_md0u6qEg6rby6
!unzip dataminingmt1782.zip
```

## Importing Required Libraries

```
!pip install -q wordcloud
import wordcloud
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from nltk.stem.wordnet import WordNetLemmatizer
import pickle
import cv2
import re
from pylab import rcParams
from matplotlib import rc
from nltk.tokenize import word_tokenize
from textblob import TextBlob
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, plot_confusion_matrix
from tqdm import tqdm
import torch
from torch import nn, optim
from torch.utils import data
%matplotlib inline
%config InlineBackend.figure_formate = 'retina'
sns.set(style = 'whitegrid', palette = 'muted', font_scale = 1.2)
HAPPY_COLOURS_PALETTE = ['#01BEFE', '#FFDD00', '#FF7D00', '#FF006D', '#ADFF02', '#8F00FF']
\verb|sns.set_palette(sns.color_palette(HAPPY_COLOURS_PALETTE))| \\
rcParams['figure.figsize'] = 8,8
RANDOM\_SEED = 42
np.random.seed(RANDOM SEED)
torch.manual_seed(RANDOM_SEED)
```

### Loading the file Data-frames

```
train_df = pd.read_csv('train.csv')
test_df = pd.read_csv('test.csv')
train_df
```

### **Setting Hyper Parameters**

```
IMG_SCALING = (3, 3)
BATCH_SIZE = 48
GAUSSIAN_NOISE = 0.1
N_CLASSES = 3
dim = (192,192)
MAIN_PATH = ''
```

### Exploratory Data Analysis and Text pre-processing

```
print(train_df['image id'].duplicated().any()) # Check if there are any duplicated images in the
data
train_df['label'].value_counts().plot(kind = 'barh')# Plot the counts of each label
train_df['label_num'].value_counts().plot(kind = 'barh') # Plot the counts of each label_num
```

```
sns.countplot(train_df.label_num)
plt.xlabel('Sentiment')
```

```
train_df['text'] = train_df['text'].apply(lambda x:x.lower())
test_df['text'] = test_df['text'].apply(lambda x: x.lower())# Convert the text in lower case
print(train_df[train_df['text']=='#name?']) # Check if there are any empty text images in the
dataset
print(test_df[test_df['text'] == '#name?'])
train_df['text']
```

```
train_df.isnull().sum() #Check number of null entries in the dataframe
```

```
train_df.drop(858,inplace=True)# Remove the corrupted Image
train_df[ train_df['image id']=='image_6357.jpg' ]
```

```
num = 858#Check that the corrupted image has been removed
train_df['label'].iloc[num]
img = cv2.imread(MAIN_PATH +'train_images/train_images/'+train_df['image id'].iloc[num])
plt.figure(figsize = (12*1.2,8*1.2))
plt.imshow(img)
plt.axis('off')
```

```
for i,id in enumerate(train_df['image id']):
    try:
        img = cv2.imread(MAIN_PATH +'train_images/train_images/'+id)
        if(img.shape[2]!=3):
            print(id + ' ' + img.shape)
    except:
        print(id+' '+ str(i))
# Verify that all the images that are read are of dimension(None,None,3)
```

```
for text,i in zip(train_df['text'],train_df['ID']):
    print(text)
# Go through the text and see what kind of processing is needed
```

```
# Text cleaning very important
def clean_str(string):
    string = re.sub(r"\n", " ", string)
    string = re.sub(r"\r", " ", string)
    string = re.sub(r"[0-9]", "digit", string)# Replace digits with the word digit
    string = re.sub(r"\'", " ", string)
    string = re.sub(r"\"", " ", string)
    string = re.sub(r"\?", " ", string)
    string = re.sub(r"\!", " ", string)
    string = re.sub(r"\/", " ", string)
    string = re.sub(r"\\", " ", string)
    string = re.sub(r"\.", " ", string)
    sample = re.sub(r'''(?i))b((?:https?://|www\d{0,3}[.]|[a-z0-9.\-]+[.][a-z]{2,4}/)(?:[^\s())
<>]+|\(([^\s()<>]+|(\([^\s()<>]+\)))*\))+(?:\(([^\s()<>]+\)))*\)|[^\s`!()\[\]
{};:'".,<>?«»""']))''', " ", string)# Remove hyperlinks
    sample = re.sub(r"http\S+", "", sample)
    sample = re.sub(r"www.[a-zA-Z0-9_@]+.COM","", sample)# Remove the hyperlinks with www and
all the possible missing combintations of .
    sample = re.sub(r"WWW.[a-zA-Z0-9_@]+.com","", sample)
    sample = re.sub(r"www.[a-zA-Z0-9_@]+.com","", sample)
    sample = re.sub(r"WWW.[a-zA-Z0-9_@]+.COM","", sample)
    sample = re.sub(r"www [a-zA-Z0-9_@]+ COM","", sample)
    sample = re.sub(r"WWW [a-zA-Z0-9_@]+ com","", sample)
    sample = re.sub(r"www [a-zA-Z0-9 @]+ com","", sample)
    sample = re.sub(r"WWW [a-zA-Z0-9_@]+ COM","", sample)
    sample = re.sub(r"www.[a-zA-Z0-9_@]+ COM","", sample)
    sample = re.sub(r"WWW.[a-zA-Z0-9_@]+ com","", sample)
    sample = re.sub(r"www.[a-zA-Z0-9_@]+ com","", sample)
    sample = re.sub(r"WWW.[a-zA-Z0-9 @]+ COM","", sample)
    sample = re.sub(r"www [a-zA-Z0-9_@]+.COM","", sample)
    sample = re.sub(r"www.[a-zA-Z0-9_@]+.COM","", sample)
    sample = re.sub(r"WWW[a-zA-Z0-9 @]+com","", sample)
    sample = re.sub(r"www[a-zA-Z0-9_@]+com","", sample)
    sample = re.sub(r"WWW[a-zA-Z0-9_@]+COM","", sample)
    sample = re.sub(r"www[a-zA-Z0-9_@]+.COM","", sample)
    sample = re.sub(r"WWW[a-zA-Z0-9_@]+.com","", sample)
    sample = re.sub(r"www[a-zA-Z0-9 @]+.com","", sample)
    sample = re.sub(r"WWW[a-zA-Z0-9_@]+.COM","", sample)
    sample = re.sub(r"www.[a-zA-Z0-9_@]+COM","", sample)
    sample = re.sub(r"WWW.[a-zA-Z0-9_@]+com","", sample)
    sample = re.sub(r"www.[a-zA-Z0-9_@]+com","", sample)
    sample = re.sub(r"WWW.[a-zA-Z0-9_@]+COM","", sample)
```

```
sample = re.sub(r"http\S+", "", sample)
sample = re.sub(r"[a-zA-Z0-9_@]+.COM","", sample)
return sample
# string = re.sub(r"^https?:\/\/.*[\r\n]*', '', string, flags=re.MULTILINE) # WILL REMOVE
HYPERLINKS!!!!
return string.strip().lower()
train_df['text'] = train_df['text'].apply(clean_str) # Should be very very clean
```

```
import nltk
nltk.download('punkt')
def TOK(text):
    return word_tokenize(text.lower())
train_df['tokens'] = train_df.text.apply(TOK)

train_df
```

```
test_df['tokens'] = test_df.text.apply(TOK)
test_df
```

```
nltk.download('stopwords')
from nltk.corpus import stopwords
stop_words = [word.lower() for word in stopwords.words('english')]
def remove_stopwords(tokens):
  ans = []
  for tok in tokens:
    if tok in stop_words:
     continue
    pas = True
    for s in '@!`~\'\\*&^\$#-=+[].,<>?':
     if s in tok:
        pas = False
        break
    if not pas:
      continue
    if 'meme' in tok:
     continue
    ans.append(tok)
  return ans
# print(train_df.tokens.iloc[599],remove_stopwords(train_df.tokens.iloc[599]))
```

```
train_df.tokens = train_df.tokens.apply(remove_stopwords)
test_df.tokens = test_df.tokens.apply(remove_stopwords)
```

```
def Tokens_to_text(tokens):
    s = ''
    for token in tokens:
        s+=token+' '
    return s.lower().strip()

def maxLen(df):
    x = 0
    for i, tokens in enumerate(df.tokens):
        x = max(x, len(tokens))
    return x

max_words = maxLen(train_df)
```

```
train_df.text = train_df.tokens.apply(Tokens_to_text)
test_df.text = test_df.tokens.apply(Tokens_to_text)
```

#### Before processing vs after processing

```
Before: 3...2...1..dont do it timmy!

After: digitdigitdigit dont do it timmy

Before: a vote for trump is a vote for putin 2009 www.protectourelections.com

After a vote for trump is a vote for putin 2009

Before: "i start where the last man left off." thomas edison visit: www.cettechnology.com/memes for more quotes @ techsolmarketing.com - free for use without modification

After: i start where the last man left off thomas edison visit: memes for more quotes @ free for use without modification
```

### Using textblob to get Subjectivity and Polarity

```
def getPolarityfromBlob(text):
    return TextBlob(text).sentiment.polarity
def getSubjectivityfromBlob(text):
    return TextBlob(text).sentiment.subjectivity
```

```
train_df['Subjectivity'] = train_df['text'].apply(lambda x: getSubjectivityfromBlob(x))
train_df['Polarity'] = train_df['text'].apply(lambda x: getPolarityfromBlob(x))
train_df
```

```
test_df['Subjectivity'] = test_df['text'].apply(lambda x: getSubjectivityfromBlob(x))
test_df['Polarity'] = test_df['text'].apply(lambda x: getPolarityfromBlob(x))
test_df
```

```
from sklearn.model_selection import train_test_split
X = train_df[['Subjectivity', 'Polarity'] ].values
y = train_df.label_num.values
train_X, test_X, train_y, test_y = train_test_split(X,y, test_size = 0.10, random_state = 42)
print(train_X.shape, test_X.shape)
```

## Training a RandomForestClassifier

```
from sklearn.ensemble import RandomForestClassifier
clf = RandomForestClassifier(max_depth=2, random_state=0)
clf.fit(train_X, train_y)
```

## Checking accuracy on validation dataset

```
from sklearn.metrics import accuracy_score
pred = clf.predict(test_X)
print(accuracy_score(test_y, pred))
```

## Making Submission

```
pred = clf.predict(test_df[['Subjectivity', 'Polarity']].values)

IDs = test_df.ID.values

labels = pred

Submission = pd.DataFrame({"ID":IDs, "label_num":labels})

Submission.to_csv("Submission.csv", index = False)
```