Dawood Sarfraz

Sentimental Analysis

Support Vector Machine

Random Forest

Decision Tree

XGBoost

Logistic Regression

Performed Feature Engineering

About Dataset

Context This is the sentiment 140 dataset. It contains 1,600,000 tweets extracted using the twitter api. The tweets have been annotated (0 = negative, 4 = positive) and they can be used to detect sentiment.

Content

It contains the following 6 fields:

target:

The polarity of the tweet (0 = negative, 4 = positive)

ids:

The id of the tweet (2087)

date:

The date of the tweet (Sat May 16 23:58:44 UTC 2009)

flag:

The query (lyx). If there is no query, then this value is NO QUERY.

user:

The user that tweeted (robotickilldozr)

text:

The text of the tweet (Lyx is cool)

```
In [616... # This Library is Just for Calculte Ruuning Time of Notebook
         from datetime import datetime
         start time = datetime.now()
         end time = datetime.now()
         total time = end time - start time
In [617... ls
         nltk data/
                                                     trained Model.sav
         'Screenshot from 2024-06-30 01-59-16.png'
                                                     tweets.csv
         'Sentimental Analysis.ipynb'
                                                     tweets.zip
         trained Model logistic regression.sav
In [618... # mostly we get dataset into Zip file Here is code to extact
         from zipfile import ZipFile as zf
         dataset file = "tweets.zip"
         with zf(dataset_file, "r") as zip:
```

```
zip.extractall()
             print(f"{dataset file} has extacted start your project.")
        tweets.zip has extacted start your project.
In [619... ls
                                                     trained Model.sav
         nltk data/
         'Screenshot from 2024-06-30 01-59-16.png'
                                                     tweets.csv
         'Sentimental Analysis.ipynb'
                                                     tweets.zip
         trained Model logistic regression.sav
In [620... # Importing Useful Libraries
In [621... import re
         import numpy as np
         import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         # Import Natural Language Tool Kit library for text processing
         from nltk.stem.porter import PorterStemmer
         from nltk.corpus import stopwords
         # Downloading "Stopworks"
         import nltk
         nltk.download("stopwords")
         # Import Sklean library
         import sklearn
         from sklearn.model selection import train test split
         from sklearn.feature extraction.text import TfidfVectorizer
         from sklearn.linear model import LogisticRegression
         from sklearn.svm import SVC
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier
         from xgboost import XGBClassifier
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.metrics import accuracy score
```

```
print('The nltk version using {}.'.format(nltk. version ))
         print('The scikit-learn version using {}.'.format(sklearn. version ))
        The nltk version using 3.8.1.
        The scikit-learn version using 1.4.2.
        [nltk data] Downloading package stopwords to /home/ubuntu/nltk data...
        [nltk data] Package stopwords is already up-to-date!
In [622... print(stopwords.words('english')) # As my data set is in English so im using "english"
        ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've", "you'll", "you'd", 'your',
        'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "i
        t's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'tha
        t', "that'll", 'these', 'those', 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'hav
        ing', 'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while',
        'of', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'after', 'abo
        ve', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', 'further', 'then', 'on
        ce', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other',
        'some', 'such', 'no', 'nor', 'not', 'only', 'own', 'same', 'so', 'than', 'too', 'very', 's', 't', 'can', 'will', 'ju
        st', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', "aren't", 'c
        ouldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'i
        sn', "isn't", 'ma', 'mightn', "mightn't", 'mustn', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shou
        ldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'wouldn', "wouldn't"]
In [623... tweets data = pd.read csv("tweets.csv",encoding="ISO-8859-1")
         original data = pd.read csv("tweets.csv",encoding="ISO-8859-1")
In [624... rows, columns = tweets data.shape
In [625... print(f"{rows} are total rows in {dataset file}")
        1599999 are total rows in tweets.zip
In [626... print(f"{columns} are total columns in {dataset file}")
        6 are total columns in tweets.zip
In [627... tweets data
```

Out[627	Target		ids	date	flag	user	text
	0	0	1467810672	Mon Apr 06 22:19:49 PDT 2009	NO_QUERY	scotthamilton	is upset that he can't update his Facebook by
	1	0	1467810917	Mon Apr 06 22:19:53 PDT 2009	NO_QUERY	mattycus	@Kenichan I dived many times for the ball. Man
	2	0	1467811184	Mon Apr 06 22:19:57 PDT 2009	NO_QUERY	ElleCTF	my whole body feels itchy and like its on fire
	3	0	1467811193	Mon Apr 06 22:19:57 PDT 2009	NO_QUERY	Karoli	@nationwideclass no, it's not behaving at all
	4	0	1467811372	Mon Apr 06 22:20:00 PDT 2009	NO_QUERY	joy_wolf	@Kwesidei not the whole crew
	•••		•••		•••		
	1599994	4	2193601966	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	AmandaMarie1028	Just woke up. Having no school is the best fee
	1599995	4	2193601969	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	The WDB oards	TheWDB.com - Very cool to hear old Walt interv
	1599996	4	2193601991	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	bpbabe	Are you ready for your MoJo Makeover? Ask me f
	1599997	4	2193602064	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	tinydiamondz	Happy 38th Birthday to my boo of alll time!!!
	1599998	4	2193602129	Tue Jun 16 08:40:50 PDT 2009	NO_QUERY	RyanTrevMorris	happy #charitytuesday @theNSPCC @SparksCharity

1599999 rows × 6 columns

If no lables so Assigning column names

```
In [629... #column_names = ["Target", "ids", "date", "flag", "user", "text"]
In [630... tweets_data.sample(10)
```

text	user	flag	date	ids	Target		Out[630	
Im back soo tweet me	official_roxy_j	NO_QUERY	Sat Jun 06 16:49:06 PDT 2009	2059381010	4	1427351		
@Evelyncharmz	m4s	NO_QUERY	Sun May 17 09:50:35 PDT 2009	1827030251	4	961269		
The lil lion cried after I dropped him off in	xoxWENDYxox	NO_QUERY	Thu Jun 25 06:04:48 PDT 2009	2325798140	0	790898		
@Uk_joedan_fan Thanks. She's seeing the neurol	Pixie_Anna	NO_QUERY	Sat Jun 06 10:48:45 PDT 2009	2056256111	0	396033		
think its very hard to learn the little one, t	LeaAndLuas	NO_QUERY	Mon May 18 03:36:56 PDT 2009	1834618006	0	126930		
raining today !	Peyton_Irving	NO_QUERY	Fri Jun 05 09:11:24 PDT 2009	2044323024	0	356585		
@caseylynnm weird, that does suck. I still h	mollier	NO_QUERY	Wed Jun 17 22:38:52 PDT 2009	2218528002	0	594501		
Its not faire !! Why cant Bon Jovi come to Mon	TickleMeStefy	NO_QUERY	Fri Jun 05 18:47:15 PDT 2009	2050504082	0	371730		
@vagrantdead ty for the #followfriday playoff 	_LostSoul_	NO_QUERY	Fri Jun 05 16:35:06 PDT 2009	2049267461	4	1362234		
I'm playing pet society! hahaha It kills bored	hannnahismyname	NO_QUERY	Sun May 31 03:05:52 PDT 2009	1979449960	4	1160689		

In [631... tweets_data.tail(10)

Out[631		Target	ids	date	flag	user	text				
	1599989	4	2193579249	Tue Jun 16 08:38:59 PDT 2009	NO_QUERY	razzberry5594	WOOOO! Xbox is back				
	1599990	4	2193579284	Tue Jun 16 08:38:59 PDT 2009	NO_QUERY	AgustinaP	@rmedina @LaTati Mmmm That sounds absolutely				
	1599991	4	2193579434	Tue Jun 16 08:39:00 PDT 2009	NO_QUERY	sdancingsteph	ReCoVeRiNg FrOm ThE lOnG wEeKeNd				
	1599992	4	2193579477	Tue Jun 16 08:39:00 PDT 2009	NO_QUERY	ChloeAmisha	@SCOOBY_GRITBOYS				
	1599993	4	2193579489	Tue Jun 16 08:39:00 PDT 2009	NO_QUERY	EvolveTom	@Cliff_Forster Yeah, that does work better tha				
	1599994	4	2193601966	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	AmandaMarie1028	Just woke up. Having no school is the best fee				
	1599995	4	2193601969	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	TheWDBoards	TheWDB.com - Very cool to hear old Walt interv				
	1599996	4	2193601991	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	bpbabe	Are you ready for your MoJo Makeover? Ask me f				
	1599997	4	2193602064	Tue Jun 16 08:40:49 PDT 2009	NO_QUERY	tinydiamondz	Happy 38th Birthday to my boo of alll time!!!				
	1599998	4	2193602129	Tue Jun 16 08:40:50 PDT 2009	NO_QUERY	RyanTrevMorris	happy #charitytuesday @theNSPCC @SparksCharity				
In [632	<pre>tweets_data.replace({"Target":{4:1}}, inplace = True) # Making Target 04 to 01</pre>										

```
In [632... tweets_data.replace({"Target":{4:1}}, inplace = True) # Making Target 0----4 to 0----1
In [633... tweets_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1599999 entries, 0 to 1599998
        Data columns (total 6 columns):
             Column Non-Null Count
                                        Dtype
             Target 1599999 non-null int64
                     1599999 non-null int64
             ids
                     1599999 non-null object
             date
                     1599999 non-null object
         3
             flag
                     1599999 non-null object
             user
         5
                     1599999 non-null object
             text
        dtypes: int64(2), object(4)
        memory usage: 73.2+ MB
In [634... tweets data.duplicated().sum() # Check duplicates in Data set
Out[634... 0
In [635... tweets data.ids.duplicated().sum() # Check duplicate ids
Out[635... 1685
In [636... # Creating a new column for Length of Every tweet
         tweets data['text length'] = tweets data['text'].apply(len)
In [637... tweets data["text length"]
Out[637... 0
                     111
          1
                      89
          2
                      47
                     111
                      29
                     . . .
          1599994
                      56
          1599995
                      81
          1599996
                      57
          1599997
                      65
          1599998
                      62
          Name: text length, Length: 1599999, dtype: int64
```

```
In [638... max(tweets data.text length) # Maximum length of tweet
Out[638... 717
In [639... print(tweets data["text"])
         0
                    is upset that he can't update his Facebook by ...
         1
                    @Kenichan I dived many times for the ball. Man...
         2
                      my whole body feels itchy and like its on fire
         3
                    @nationwideclass no, it's not behaving at all....
         4
                                        @Kwesidei not the whole crew
                    Just woke up. Having no school is the best fee...
         1599994
         1599995
                   TheWDB.com - Very cool to hear old Walt interv...
                   Are you ready for your MoJo Makeover? Ask me f...
         1599996
                    Happy 38th Birthday to my boo of alll time!!! ...
         1599997
         1599998
                   happy #charitytuesday @theNSPCC @SparksCharity...
        Name: text, Length: 1599999, dtype: object
In [640... tweets data["text"].str.len().min() # minimum length tweet
Out[640... 6
In [641... tweets data["text"].str.len().idxmin() # minimum length tweet inex
Out[641... 4232
In [642... tweets data.text[tweets data["text"].str.len().idxmin()]
Out[642... ' just '
In [643... min value = tweets data.text[tweets data.text.str.len().idxmin()] # Same as above but different method
In [644... min value
Out[644... ' just '
In [645... len(min value)
```

7/1/24, 8:08 PM

```
Out[645... 6
In [646... max_value = tweets_data.text[tweets_data.text.str.len().idxmax()] # Same as above but different method

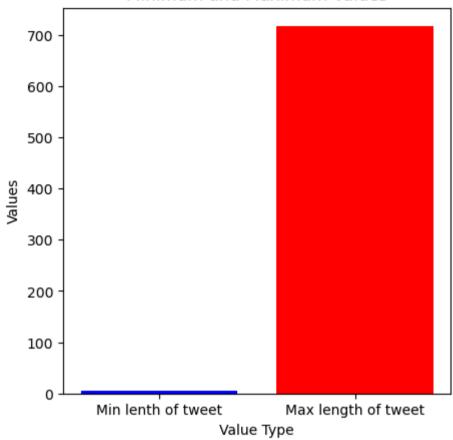
In [647... len(max_value)

Out[647... 717

In []:

In [648... plt.figure(figsize=(5,5))
    plt.bar(['Min lenth of tweet', 'Max length of tweet'], [len(min_value), len(max_value)], color=['blue', 'red'], )
    plt.xlabel('Value Type')
    plt.ylabel('Values')
    plt.title('Minimum and Maximum Values')
    plt.show()
```





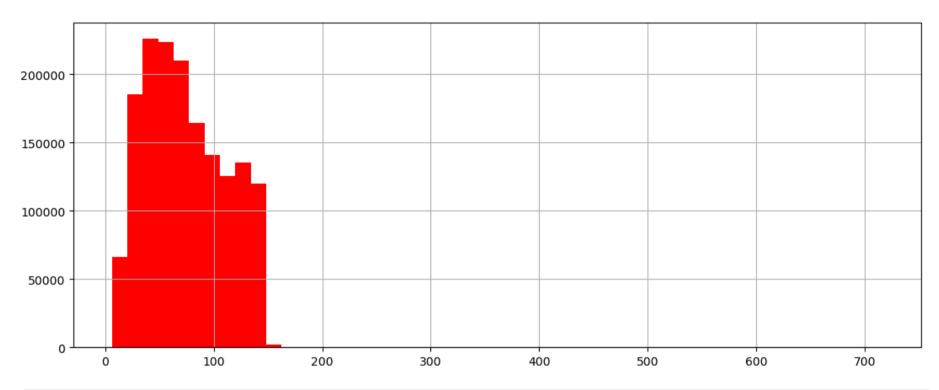
```
In [649... tweets_data["text"].str.len().max() # minimum length tweet

Out[649... 717

In [650... tweets_data['text_length'] = tweets_data['text'].apply(len)
# Find the index of the maximum length
max_length_index = tweets_data['text_length'].idxmax()
# Retrieve the tweet with the maximum length
max_length_tweet = tweets_data.loc[max_length_index, 'text']
print("Tweet with maximum length: ", len(max_length_tweet),"\n")
print(max_length_tweet)
```

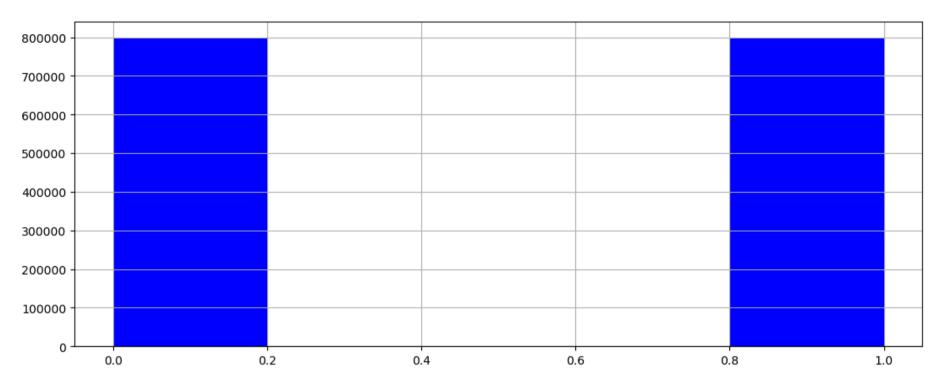
Tweet with maximum length: 717

@catfish_ohm \tilde{A} $\hat{A}^1\hat{A}^-\tilde{A}$ \hat{A} , $\hat{A}^1\hat{A}$, \hat{A} $\hat{A}^1\hat{A}$, \hat{A} $\hat{A}^1\hat{A}$, \hat{A} $\hat{A}^1\hat{A}$, \hat{A} , $\hat{A}^1\hat{A}$, \hat{A} , $\hat{A}^2\hat{A}$ $\hat{A}^1\hat{A}$, \hat{A} , $\hat{A}^2\hat{A}$ $\hat{A}^1\hat{A}$, \hat{A} , $\hat{A}^2\hat{A}$ $\hat{A}^1\hat{A}$, $\hat{A}^2\hat{A}$ $\hat{A}^2\hat{A}$ $\hat{A}^2\hat{A}$ $\hat{A}^2\hat{A}$, $\hat{A}^2\hat{A}$ $\hat{A}^2\hat{A}$, $\hat{A}^2\hat{A}$ $\hat{A}^2\hat{A}$, $\hat{A}^2\hat{A}$



In [653... tweets_data["Target"].hist(bins = 5, figsize = (13,5), color = 'b')

Out[653... <Axes: >



In [654... # My Dataset is very Large to perform Analysis I'm using just a subset form it

```
In [655...
tweets_data['tweet_created'] = pd.to_datetime(tweets_data['date']).dt.date
numberoftweets = tweets_data.groupby('tweet_created').size()
numberoftweets = numberoftweets.reset_index()
numberoftweets.columns = ['Date', 'Count']
numberoftweets
```

/tmp/ipykernel_5457/1907819345.py:1: FutureWarning: Parsed string "Mon Apr 06 22:19:49 PDT 2009" included an un-recognized timezone "PDT". Dropping unrecognized timezones is deprecated; in a future version this will raise. Instead p ass the string without the timezone, then use .tz_localize to convert to a recognized timezone.

tweets data['tweet created'] = pd.to datetime(tweets data['date']).dt.date

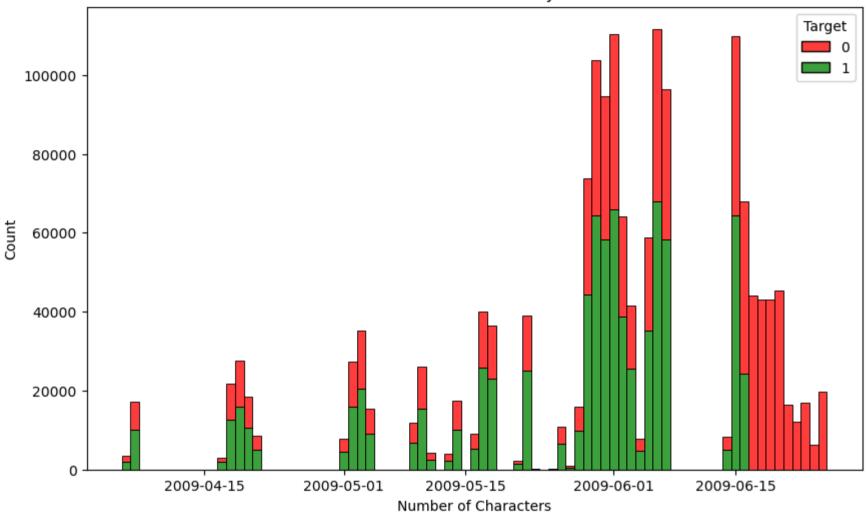
n		+	Γ	6	5	5	
U	u	L	L	U	J	J	

	Date	Count
0	2009-04-06	3359
1	2009-04-07	17311
2	2009-04-17	3084
3	2009-04-18	21754
4	2009-04-19	27469
5	2009-04-20	18460
6	2009-04-21	8587
7	2009-05-01	7716
8	2009-05-02	27434
9	2009-05-03	35333
10	2009-05-04	15481
11	2009-05-09	11739
12	2009-05-10	26029
13	2009-05-11	4186
14	2009-05-13	4066
15	2009-05-14	17460
16	2009-05-16	9146
17	2009-05-17	40154
18	2009-05-18	36469
19	2009-05-21	2132
20	2009-05-22	39074
21	2009-05-23	169
22	2009-05-25	169
23	2009-05-26	10778

	Date	Count
24	2009-05-27	841
25	2009-05-28	15903
26	2009-05-29	73827
27	2009-05-30	103673
28	2009-05-31	94588
29	2009-06-01	110290
30	2009-06-02	64192
31	2009-06-03	41588
32	2009-06-04	7842
33	2009-06-05	58757
34	2009-06-06	111676
35	2009-06-07	96350
36	2009-06-14	8272
37	2009-06-15	109781
38	2009-06-16	67980
39	2009-06-17	44012
40	2009-06-18	43004
41	2009-06-19	43136
42	2009-06-20	45364
43	2009-06-21	16360
44	2009-06-22	12009
45	2009-06-23	17002
46	2009-06-24	6299
47	2009-06-25	19694

```
In [656... numberoftweets.Count.max() # Maximum Tweets created in One Day
Out [656... 111676
In [657... numberoftweets.Count.min() # Minimum tweets created in One Day
Out[657... 169
In [658... # Find dates with the same number of tweets
         same tweets days = numberoftweets[numberoftweets['Count'].duplicated(keep=False)]['Date'].tolist()
         print("Days with the same number of tweets:")
         for day in same_tweets_days:
              print(day)
        Days with the same number of tweets:
         2009-05-23
        2009-05-25
In [659... tweets data['tweet created'].nunique()
Out[659... 48
In [660... # Plotting
         plt.figure(figsize=(10, 6))
         sns.histplot(data = tweets data, x = 'tweet created', hue='Target', multiple='stack', palette={0: 'red', 1: 'green'
         plt.xlabel('Number of Characters')
         plt.ylabel('Count')
         plt.title('Character Counts in Tweets by Sentiment')
         plt.show()
```

Character Counts in Tweets by Sentiment



```
In [661...
plt.figure(figsize=(16, 16))
    tweetcreated= tweets_data['tweet_created']
    color = sns.color_palette( "husl", len(numberoftweets))
    sns.barplot(x=numberoftweets.Count, y=numberoftweets.Date, data=numberoftweets, palette=color, legend=False)
    plt.title('Number of Tweets Created by Date')
    plt.ylabel("Counts of Tweets")
    plt.xlabel("Date of Tweets")
```

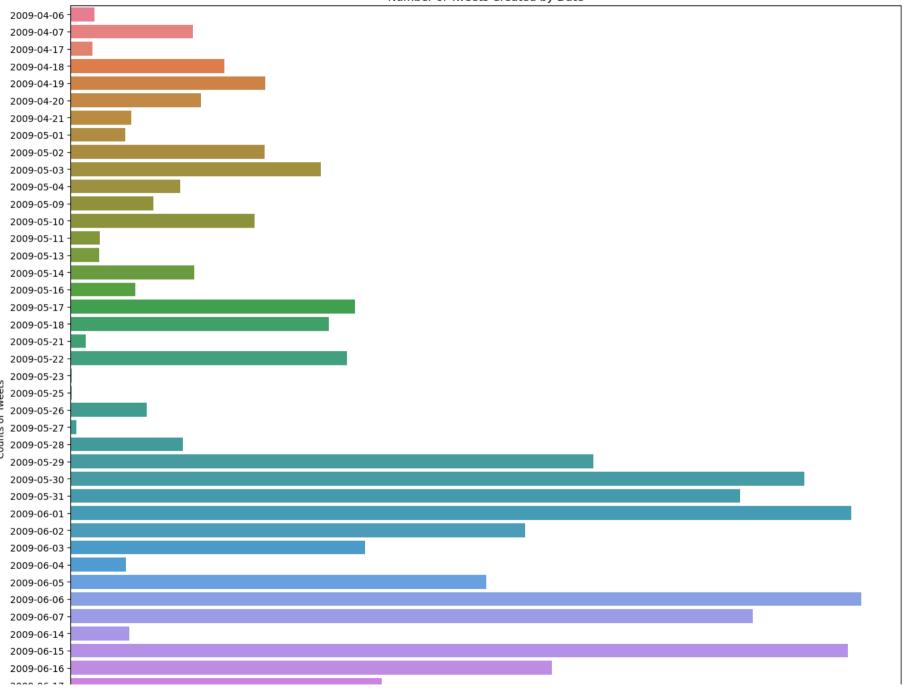
```
plt.xticks(rotation=90)
plt.show()
```

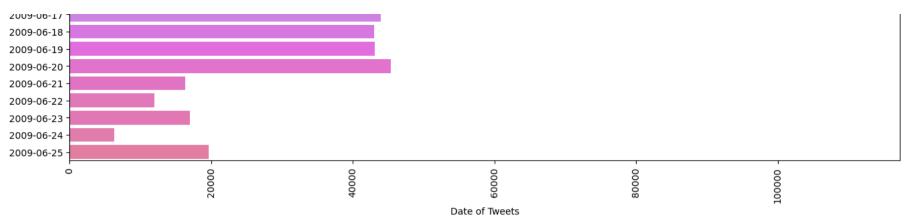
/tmp/ipykernel 5457/3614696218.py:4: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=numberoftweets.Count, y=numberoftweets.Date, data=numberoftweets, palette=color, legend=False)

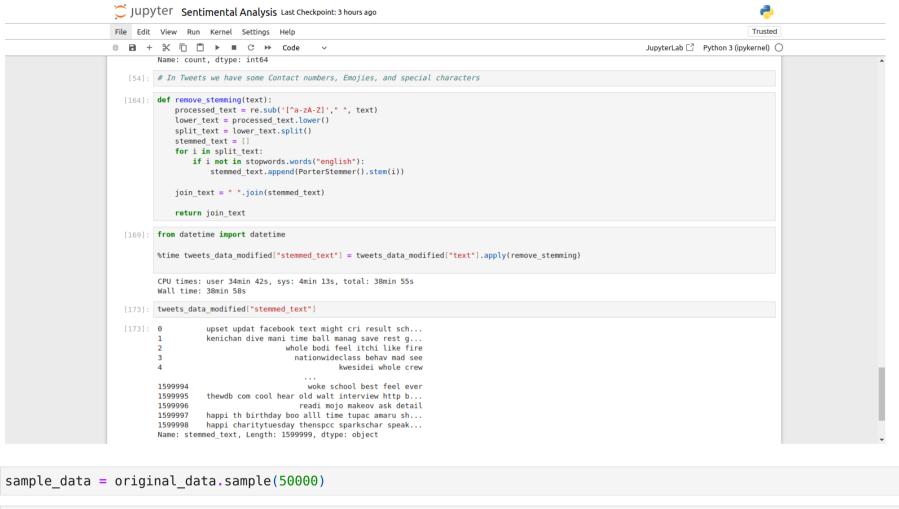






Sampling Data Set

Data set contain millions of tweets. So i'm take few samples bcz it will lots of time to train. I used whole Dataset it took almost 40 mints in text processing. You can try Whole Dataset.



```
In [664... sample_data = original_data.sample(50000)
In [665... sample data.head(10)
```

ut[665		Target	ids	date	flag	user	text
	1442958	4	2062053422	Sat Jun 06 22:13:19 PDT 2009	NO_QUERY	KellyG5	http://twitpic.com/1uoev - ha ha what are you
	34148	0	1564713481	Mon Apr 20 04:39:20 PDT 2009	NO_QUERY	cl4irethomas	@AnnieC1 could do with some sympathy she's
	789182	0	2325319004	Thu Jun 25 05:13:08 PDT 2009	NO_QUERY	acullenatheart	my iPhone is attempting to kill itself. Can't
	724967	0	2262207910	Sat Jun 20 22:04:45 PDT 2009	NO_QUERY	locn	Zion and Vegas was so much awesomeness. Too ba
	956955	4	1825348470	Sun May 17 05:21:54 PDT 2009	NO_QUERY	CarlaJaynee	Aww, 100 followers! I feel loved I love you g
	1302924	4	2008849072	Tue Jun 02 14:45:17 PDT 2009	NO_QUERY	salihagul	missing my daddio and laughing with him. i get
	1237897	4	1993141661	Mon Jun 01 09:53:57 PDT 2009	NO_QUERY	Janoita	Mmmm snickers ice cream
	1293666	4	2003248623	Tue Jun 02 06:10:33 PDT 2009	NO_QUERY	Tehlem0n	@jessicatsang Im on study leave. Only have to
	1057833	4	1962815334	Fri May 29 11:43:15 PDT 2009	NO_QUERY	katiePEACE	Ahahaha helping my uncle move hopfully hannah
	312714	0	2001586656	Tue Jun 02 01:22:51 PDT 2009	NO_QUERY	mealienprobeu	Another financial dispute with my mother that

In [666... sample_data.tail(10)

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Out[666		Target	ids	date	flag	user	text
	618693	0	2227618214	Thu Jun 18 13:25:24 PDT 2009	NO_QUERY	JGXO	Jus about ready, then off to IKEA w/ one of my
	477042	0	2178055311	Mon Jun 15 07:07:38 PDT 2009	NO_QUERY	itskb	the john lennon new york years exibit at the r
	1022957	4	1882912212	Fri May 22 07:41:48 PDT 2009	NO_QUERY	LeoWolfe	@xClureBearx thanks for the follow.
	771971	0	2302595827	Tue Jun 23 17:01:48 PDT 2009	NO_QUERY	bobbysgirl1972	I have had the hiccups for the past 4 1/2 hrs
	205948	0	1973005538	Sat May 30 10:17:46 PDT 2009	NO_QUERY	sethu_j	why do no shops do gorgeous, big bags nowadays
	1033361	4	1955775749	Thu May 28 20:36:19 PDT 2009	NO_QUERY	admance	@artfanatic411 Yes, I didTwitterholic is my
	1413067	4	2056772328	Sat Jun 06 11:45:54 PDT 2009	NO_QUERY	Irv25	@David_DB your welcome!!!
	416893	0	2061407079	Sat Jun 06 20:48:16 PDT 2009	NO_QUERY	kulex	Aww shit. Why did I decide to watch Grave of t
	812540	4	1548528643	Fri Apr 17 21:12:19 PDT 2009	NO_QUERY	xCAROxx	i love movies with sexy people in them. theyr
	496806	0	2185683159	Mon Jun 15 17:45:00 PDT 2009	NO_QUERY	oliviaistaken	Going to tweet all I can before "Downtime

In [667... sample_data.sample(10)

Out[667		Target	ids	date	flag	user	text
	32249	0	1564211797	Mon Apr 20 02:21:01 PDT 2009	NO_QUERY	vspijker	Back at work again tomorrow going to
	844138	4	1564005659	Mon Apr 20 01:17:35 PDT 2009	NO_QUERY	Rullet	@AMPRGROUP My street album free download http:
	93488	0	1770676922	Mon May 11 21:57:12 PDT 2009	NO_QUERY	sugarstar80	So fed up the Canucks are out of the playoffs
	1312056	4	2013576880	Tue Jun 02 23:01:16 PDT 2009	NO_QUERY	Lupeezy	My phone is Gona sleep before i do! Empress ke
	194912	0	1970391719	Sat May 30 03:40:50 PDT 2009	NO_QUERY	karmachord	@RedMummy I know what a fool I am!
	1079699	4	1968116955	Fri May 29 20:47:28 PDT 2009	NO_QUERY	MrsDawkins	@niyana where the link homie y do i feel we
	799299	0	2328936546	Thu Jun 25 10:09:37 PDT 2009	NO_QUERY	Tingbabe	not feeling good at all
	705462	0	2256369245	Sat Jun 20 12:40:16 PDT 2009	NO_QUERY	_Catalyst	@xanderkitty OMG you poor thing. I would have
	550302	0	2202875683	Tue Jun 16 22:34:55 PDT 2009	NO_QUERY	chekkacuomova	@sergeantkero are you using mac or windows? I
	69188	0	1693179792	Sun May 03 21:47:07 PDT 2009	NO_QUERY	eltah_08	@RealPaulJohnson it's always somethin wit u

```
In [668... sample_data.replace({"Target":{4:1}}, inplace = True)

In [669... # Removing the URl
    def rem_url(text):
        url = re.compile(r'https?://\S+|www\.\S+')
        return url.sub(r'',text)

# Removing the stepwords 'english'
def rem_stopwords(text):
        text = ' '.join([word for word in text.split() if word not in (stopwords.words('english'))])
        return text

# Removing the html
```

```
def rem html(text):
    html=re.compile(r'<.*?>')
    return html.sub(r'',text)
def remove at mentions(text):
    return re.sub(r'@\w+', '', text)
# Removing emojis by codes
def rem emoji(text):
    emoji pattern = re.compile("["
                           u"\U0001F600-\U0001F64F"
                           u"\U0001F300-\U0001F5FF"
                           u"\U000024C2-\U0001F251"
                           "]+", flags=re.UNICODE)
    return emoji pattern.sub(r'', text)
# Fixing the text
def fixing(text):
    text = re.sub(r"Mr\'s", " he is", text)
    text = re.sub(r"Mr\'.", " he is", text)
    text = re.sub(r"Ms\'.", " She ", text)
    text = re.sub(r"She\'s", " she is", text)
    text = re.sub(r"I\'m", " i am", text)
    text = re.sub(r"haven\'t", " have not", text)
    text = re.sub(r"you\'r", " you are", text)
    text = re.sub(r"won\'t", " will not", text)
    text = re.sub(r"won\'t've", " will not have", text)
    text = re.sub(r"can\'t", " can not", text)
    text = re.sub(r"don\'t", " do not", text)
    text = re.sub(r" plz", " please", text)
    text = re.sub(r"it\'s", " its", text)
    text = re.sub(r"can\'t've", " can not have", text)
    text = re.sub(r"ma\'am", " madam", text)
    text = re.sub(r"let\'s", " let us", text)
    text = re.sub(r"ain\'t", " am not", text)
    text = re.sub(r"shan\'t", " shall not", text)
    text = re.sub(r"sha\n't", " shall not", text)
    text = re.sub(r"o\'clock", " of the clock", text)
    text = re.sub(r"y\'all", " you all", text)
```

```
text = re.sub(r"n\'t", " not", text)
   text = re.sub(r"n\'t've", " not have", text)
   text = re.sub(r"\'re", " are", text)
   text = re.sub(r"\'s", " is", text)
   text = re.sub(r"\'d", " would", text)
   text = re.sub(r"\'d've", " would have", text)
   text = re.sub(r"\'ll", " will", text)
   text = re.sub(r"\'ll've", " will have", text)
   text = re.sub(r"\'t", " not", text)
   text = re.sub(r"\'ve", " have", text)
   text = re.sub(r"\'m", " am", text)
   text = re.sub(r"\'re", " are", text)
   text = re.sub(r"He\'s", " he is", text)
   text = re.sub(r"in\'s", " in", text)
    return text
# Seperate alphabets
def sep alphabets(text):
   words = text
   words = re.findall(r''[^\W\d]+|\d+'', words)
    return " ".join(words)
def tweet to words(tweet):
   letters only = re.sub("[^a-zA-Z]", " ",tweet)
   words = letters only.lower().split()
    stops = set(stopwords.words("english"))
    meaningful words = [w for w in words if not w in stops]
    return( " ".join( meaningful words ))
sample data['final tweet'] = sample_data['text']
sample data['final tweet'] = sample data['final tweet'].astype(str).apply(remove at mentions)
sample data['final tweet'] = sample data['final tweet'].apply(lambda x : rem url(x))
sample data['final tweet'] = sample data['final tweet'].apply(lambda x : rem stopwords(x))
sample data['final tweet'] = sample data['final tweet'].apply(lambda x : rem html(x))
sample data['final tweet'] = sample data['final tweet'].apply(lambda x : rem emoji(x))
sample data['final tweet'] = sample data['final tweet'].apply(lambda x : fixing(x))
sample data['final tweet'] = sample data['final tweet'].apply(lambda x : sep alphabets(x))
```

```
sample_data['final_tweet'] = sample_data['final_tweet'].apply(lambda x : tweet_to_words(x))
sample_data['final_tweet'] = sample_data['final_tweet'].astype(str).replace("plz", "please", regex=True)
In [670... sample_data
```

Out[670		Target	ids	date	flag	user	text	final_tweet
	1442958	1	2062053422	Sat Jun 06 22:13:19 PDT 2009	NO_QUERY	KellyG5	http://twitpic.com/1uoev - ha ha what are you	ha ha saying lol cute eh lol
	34148	0	1564713481	Mon Apr 20 04:39:20 PDT 2009	NO_QUERY	cl4irethomas	@AnnieC1 could do with some sympathy she's	could sympathy work sick today offers
	789182	0	2325319004	Thu Jun 25 05:13:08 PDT 2009	NO_QUERY	acullenatheart	my iPhone is attempting to kill itself. Can't	iphone attempting kill ca wait get home try fix
	724967	0	2262207910	Sat Jun 20 22:04:45 PDT 2009	NO_QUERY	locn	Zion and Vegas was so much awesomeness. Too ba	zion vegas much awesomeness bad stupid camera
	956955	1	1825348470	Sun May 17 05:21:54 PDT 2009	NO_QUERY	CarlaJaynee	Aww, 100 followers! I feel loved I love you g	aww followers feel loved love guys lt x
	•••		•••	•••	•••	•••		
	1033361	1	1955775749	Thu May 28 20:36:19 PDT 2009	NO_QUERY	admance	@artfanatic411 Yes, I didTwitterholic is my	yes twitterholic middle name
	1413067	1	2056772328	Sat Jun 06 11:45:54 PDT 2009	NO_QUERY	Irv25	@David_DB your welcome!!!	welcome
	416893	0	2061407079	Sat Jun 06 20:48:16 PDT 2009	NO_QUERY	kulex	Aww shit. Why did I decide to watch Grave of t	aww shit decide watch grave fireflies
	812540	1	1548528643	Fri Apr 17 21:12:19 PDT 2009	NO_QUERY	xCAROxx	i love movies with sexy people in them. theyr	love movies sexy people theyre quite delicious
	496806	0	2185683159	Mon Jun 15 17:45:00 PDT 2009	NO_QUERY	oliviaistaken	Going to tweet all I can before "Downtime	going tweet quot downtime quot

50000 rows × 7 columns

In [671... sample_data.head(10)

Out[671		Target	ids	date	flag	user	text	final_tweet
	1442958	1	2062053422	Sat Jun 06 22:13:19 PDT 2009	NO_QUERY	KellyG5	http://twitpic.com/1uoev - ha ha what are you	ha ha saying lol cute eh lol
	34148	0	1564713481	Mon Apr 20 04:39:20 PDT 2009	NO_QUERY	cl4irethomas	@AnnieC1 could do with some sympathy she's	could sympathy work sick today offers
	789182	0	2325319004	Thu Jun 25 05:13:08 PDT 2009	NO_QUERY	acullenatheart	my iPhone is attempting to kill itself. Can't	iphone attempting kill ca wait get home try fix
	724967	0	2262207910	Sat Jun 20 22:04:45 PDT 2009	NO_QUERY	locn	Zion and Vegas was so much awesomeness. Too ba	zion vegas much awesomeness bad stupid camera
	956955	1	1825348470	Sun May 17 05:21:54 PDT 2009	NO_QUERY	CarlaJaynee	Aww, 100 followers! I feel loved I love you g	aww followers feel loved love guys lt x
	1302924	1	2008849072	Tue Jun 02 14:45:17 PDT 2009	NO_QUERY	salihagul	missing my daddio and laughing with him. i get	missing daddio laughing get funny ness joke ma
	1237897	1	1993141661	Mon Jun 01 09:53:57 PDT 2009	NO_QUERY	Janoita	Mmmm snickers ice cream	mmmm snickers ice cream
	1293666	1	2003248623	Tue Jun 02 06:10:33 PDT 2009	NO_QUERY	Tehlem0n	@jessicatsang Im on study leave. Only have to	im study leave go school exams
	1057833	1	1962815334	Fri May 29 11:43:15 PDT 2009	NO_QUERY	katiePEACE	Ahahaha helping my uncle move hopfully hannah	ahahaha helping uncle move hopfully hannah hel
	312714	0	2001586656	Tue Jun 02 01:22:51 PDT 2009	NO_QUERY	mealienprobeu	Another financial dispute with my mother that	another financial dispute mother started much

In [672... sample_data.tail(10)

Out[672		Target	ids	date	flag	user	text	final_tweet
	618693	0	2227618214	Thu Jun 18 13:25:24 PDT 2009	NO_QUERY	JGXO	Jus about ready, then off to IKEA w/ one of my	jus ready ikea w one fav boys yayy ca much dam
	477042	0	2178055311	Mon Jun 15 07:07:38 PDT 2009	NO_QUERY	itskb	the john lennon new york years exibit at the r	john lennon new york years exibit rock n roll
	1022957	1	1882912212	Fri May 22 07:41:48 PDT 2009	NO_QUERY	LeoWolfe	@xClureBearx thanks for the follow.	thanks follow
	771971	0	2302595827	Tue Jun 23 17:01:48 PDT 2009	NO_QUERY	bobbysgirl1972	I have had the hiccups for the past 4 1/2 hrs	hiccups past hrs help whatta ugh
	205948	0	1973005538	Sat May 30 10:17:46 PDT 2009	NO_QUERY	sethu_j	why do no shops do gorgeous, big bags nowadays	shops gorgeous big bags nowadays
	1033361	1	1955775749	Thu May 28 20:36:19 PDT 2009	NO_QUERY	admance	@artfanatic411 Yes, I didTwitterholic is my	yes twitterholic middle name
	1413067	1	2056772328	Sat Jun 06 11:45:54 PDT 2009	NO_QUERY	Irv25	@David_DB your welcome!!!	welcome
	416893	0	2061407079	Sat Jun 06 20:48:16 PDT 2009	NO_QUERY	kulex	Aww shit. Why did I decide to watch Grave of t	aww shit decide watch grave fireflies
	812540	1	1548528643	Fri Apr 17 21:12:19 PDT 2009	NO_QUERY	xCAROxx	i love movies with sexy people in them. theyr	love movies sexy people theyre quite delicious
	496806	0	2185683159	Mon Jun 15 17:45:00 PDT 2009	NO_QUERY	oliviaistaken	Going to tweet all I can before "Downtime	going tweet quot downtime quot

In [673... sample_data.sample(10)

Out[673		Target	ids	date	flag	user	text	final_tweet
	37191	0	1572884527	Mon Apr 20 22:54:19 PDT 2009	NO_QUERY	meowwzer	just woke up feeling confused and anxious abou	woke feeling confused anxious upcoming events
	1018917	1	1882095993	Fri May 22 06:13:21 PDT 2009	NO_QUERY	JasmineYun	@Eva2Ta HAVE FUN!!! Wing big at the casino fi	fun wing big casino first spend shopping
	1109350	1	1971876299	Sat May 30 08:05:19 PDT 2009	NO_QUERY	avandenb	Washing the paint out of my hair	washing paint hair
	467856	0	2175870892	Mon Jun 15 02:14:22 PDT 2009	NO_QUERY	wiga_sunshine	4:13 and i'm in my boring maximum time!! huu	boring maximum time huuugghh
	1326555	1	2015269953	Wed Jun 03 04:11:56 PDT 2009	NO_QUERY	xtnabcn	@AlexDeQuerzen I always have the work/play the	always work play theory mind suggested maybe c
	110794	0	1824950118	Sun May 17 03:40:01 PDT 2009	NO_QUERY	_mayumi	lazy sunday today i think. 4 days of school le	lazy sunday today think days school left study
	674745	0	2247880438	Fri Jun 19 19:46:00 PDT 2009	NO_QUERY	mattielong	god fucking dammit	god fucking dammit
	124164	0	1834055000	Mon May 18 01:27:24 PDT 2009	NO_QUERY	supernovi	I hate gravel it has messed my car up	hate gravel messed car
	190540	0	1969381829	Fri May 29 23:51:36 PDT 2009	NO_QUERY	ObamaKitty	Re-pinging @neoncat: @kay_rose I didn't eat at	pinging eat day gt lt eat diner yesterday hung
	292041	0	1995885275	Mon Jun 01 14:11:56 PDT 2009	NO_QUERY	lexayyyy	im hungry now	im hungry

In [674... sample_data.info()

```
<class 'pandas.core.frame.DataFrame'>
        Index: 50000 entries, 1442958 to 496806
        Data columns (total 7 columns):
                           Non-Null Count Dtype
             Column
             Target
                           50000 non-null int64
         1
             ids
                           50000 non-null int64
         2
                          50000 non-null object
              date
                          50000 non-null object
         3
             flag
             user
                           50000 non-null object
         5
             text
                           50000 non-null object
             final tweet 50000 non-null object
        dtypes: int64(2), object(5)
        memory usage: 3.1+ MB
In [675... sample data.describe() # You can't analyze Ids and Target so use just text length
Out[675...
                                      ids
                      Target
          count 50000.000000 5.000000e+04
                    0.496260 1.998724e+09
          mean
            std
                    0.499991 1.943709e+08
           min
                    0.000000 1.467812e+09
           25%
                    0.000000 1.956743e+09
                    0.000000 2.002167e+09
           50%
           75%
                    1.000000 2.177721e+09
                    1.000000 2.329205e+09
           max
 In [ ]:
In [676... sample data.ids.duplicated().sum()
Out[676... 3
In [677... sample data["final_text_length"] = sample_data["final_tweet"].apply(len) # After Pre-processing
```

```
In [678... sample data.final text length
Out[678... 1442958
                     28
          34148
                     37
          789182
                     47
          724967
                     60
          956955
                     39
          1033361
                     28
          1413067
                      7
          416893
                     37
          812540
                     65
          496806
                     30
          Name: final text length, Length: 50000, dtype: int64
In [679... sample data["text length"] = sample data["text"].apply(len) # Before Preporessing
In [680... sample data.text length
Out[680... 1442958
                     91
          34148
                     85
          789182
                     78
          724967
                     87
          956955
                     57
                      . .
          1033361
                     59
          1413067
                     26
          416893
                     60
          812540
                     98
          496806
                     53
          Name: text length, Length: 50000, dtype: int64
In [681... max(sample data.final text length)
Out[681... 137
In [682... min(sample_data.final_text_length)
Out[682... 0
```

```
In [683... sample data["final tweet"].str.len().min() # minimum length tweet
Out[683... 0
In [684... sample_data["final_tweet"].str.len().max() # minimum length tweet
Out[684... 137
In [685... sample data.final tweet.str.len().idxmin() # index of minimum lenth text
Out[685... 448773
In [686... sample data.final tweet.str.len().idxmax() # index of maximum lenth text
Out[686... 1766
In [687... sample data.final tweet[sample data.final tweet.str.len().idxmax()] # index of maximum lenth text
Out[687... 'tks pa quot tapauing quot croissant tuna knowing dat ive back back meetings since morning amp zuraidah tks buying
          fav starbuck mocha frap'
In [688... sample data.final tweet[sample data.final tweet.str.len().idxmin()] # index of minimum lenth text
Out[688... ''
         sample_data
In [689...
```

7/1/24, 8:08 PM

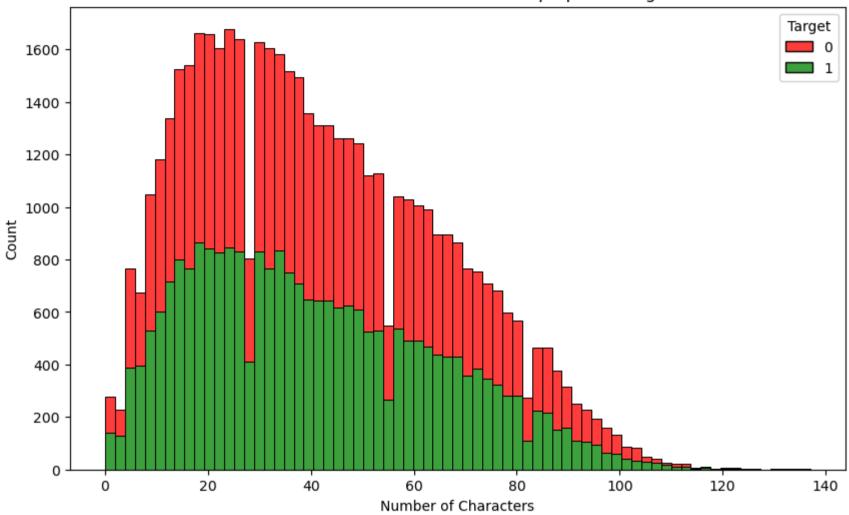
Out[689		Target	ids	date	flag	user	text	final_tweet	final_text_length	text_length
	1442958	1	2062053422	Sat Jun 06 22:13:19 PDT 2009	NO_QUERY	KellyG5	http://twitpic.com/1uoev - ha ha what are you	ha ha saying lol cute eh lol	28	91
	34148	0	1564713481	Mon Apr 20 04:39:20 PDT 2009	NO_QUERY	cl4irethomas	@AnnieC1 could do with some sympathy she's	could sympathy work sick today offers	37	85
	789182	0	2325319004	Thu Jun 25 05:13:08 PDT 2009	NO_QUERY	acullenatheart	my iPhone is attempting to kill itself. Can't	iphone attempting kill ca wait get home try fix	47	78
	724967	0	2262207910	Sat Jun 20 22:04:45 PDT 2009	NO_QUERY	locn	Zion and Vegas was so much awesomeness. Too ba	zion vegas much awesomeness bad stupid camera	60	87
	956955	1	1825348470	Sun May 17 05:21:54 PDT 2009	NO_QUERY	CarlaJaynee	Aww, 100 followers! I feel loved I love you g	aww followers feel loved love guys lt x	39	57
	•••									•••
	1033361	1	1955775749	Thu May 28 20:36:19 PDT 2009	NO_QUERY	admance	@artfanatic411 Yes, I didTwitterholic is my	yes twitterholic middle name	28	59
	1413067	1	2056772328	Sat Jun 06 11:45:54 PDT 2009	NO_QUERY	Irv25	@David_DB your welcome!!!	welcome	7	26
	416893	0	2061407079	Sat Jun 06 20:48:16	NO_QUERY	kulex	Aww shit. Why did I decide to watch Grave of t	aww shit decide watch grave fireflies	37	60

	Target	ids	date	flag	user	text	final_tweet	final_text_length	text_length
			PDT 2009						
812540	1	1548528643	Fri Apr 17 21:12:19 PDT 2009	NO_QUERY	xCAROxx	i love movies with sexy people in them. theyr	love movies sexy people theyre quite delicious	65	98
496806	0	2185683159	Mon Jun 15 17:45:00 PDT 2009	NO_QUERY	oliviaistaken	Going to tweet all I can before "Downtime	going tweet quot downtime quot	30	53

50000 rows × 9 columns

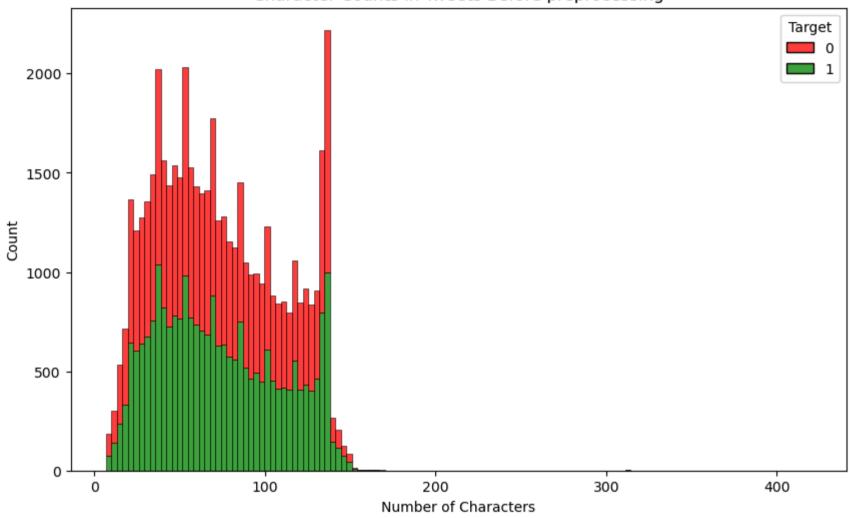
```
In [690... # Plotting
   plt.figure(figsize=(10, 6))
   sns.histplot(data=sample_data, x='final_text_length', hue='Target', multiple='stack', palette={0: 'red', 1: 'green'
        plt.xlabel('Number of Characters')
        plt.ylabel('Count')
        plt.title('Character Counts in Tweets After preprocessing')
        plt.show()
```

Character Counts in Tweets After preprocessing



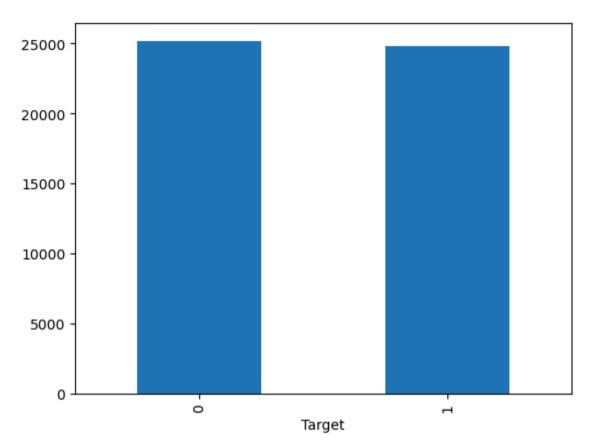
```
In [691... # Plotting
plt.figure(figsize=(10, 6))
sns.histplot(data=sample_data, x='text_length', hue='Target', multiple='stack', palette={0: 'red', 1: 'green'})
plt.xlabel('Number of Characters')
plt.ylabel('Count')
plt.title('Character Counts in Tweets Before preprocessing')
plt.show()
```

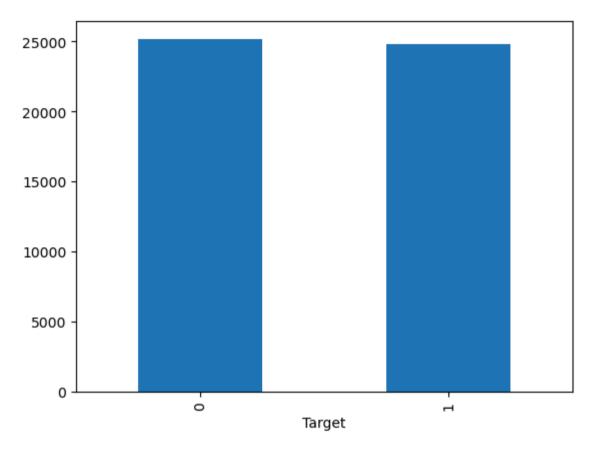




In [692... sample_data.isnull().sum()

```
Out[692... Target
          ids
          date
          flag
          user
          text
          final_tweet
          final_text_length
                               0
          text length
          dtype: int64
In [693... sample_data["date"].min() # First tweet
Out[693... 'Fri Apr 17 20:31:46 PDT 2009'
In [694... sample_data.date.max() # Last Tweet
Out[694... 'Wed May 27 07:27:36 PDT 2009'
In [695... sample_data.groupby('Target').size().plot(kind='bar')
Out[695... <Axes: xlabel='Target'>
```





```
In [698... # As in Data set there is No "2" Neutral. So lets change 4 to 1
In [699... # using_data.replace({"Target":{4:1}}, inplace = True)
In [700... sample_data
```

Out[700		Target	ids	date	flag	user	text	final_tweet	final_text_length	text_length
	1442958	1	2062053422	Sat Jun 06 22:13:19 PDT 2009	NO_QUERY	KellyG5	http://twitpic.com/1uoev - ha ha what are you	ha ha saying lol cute eh lol	28	91
	34148	0	1564713481	Mon Apr 20 04:39:20 PDT 2009	NO_QUERY	cl4irethomas	@AnnieC1 could do with some sympathy she's	could sympathy work sick today offers	37	85
	789182	0	2325319004	Thu Jun 25 05:13:08 PDT 2009	NO_QUERY	acullenatheart	my iPhone is attempting to kill itself. Can't	iphone attempting kill ca wait get home try fix	47	78
	724967	0	2262207910	Sat Jun 20 22:04:45 PDT 2009	NO_QUERY	locn	Zion and Vegas was so much awesomeness. Too ba	zion vegas much awesomeness bad stupid camera	60	87
	956955	1	1825348470	Sun May 17 05:21:54 PDT 2009	NO_QUERY	CarlaJaynee	Aww, 100 followers! I feel loved I love you g	aww followers feel loved love guys lt x	39	57
	•••								•••	
	1033361	1	1955775749	Thu May 28 20:36:19 PDT 2009	NO_QUERY	admance	@artfanatic411 Yes, I didTwitterholic is my	yes twitterholic middle name	28	59
	1413067	1	2056772328	Sat Jun 06 11:45:54 PDT 2009	NO_QUERY	Irv25	@David_DB your welcome!!!	welcome	7	26
	416893	0	2061407079	Sat Jun 06 20:48:16	NO_QUERY	kulex	Aww shit. Why did I decide to watch Grave of t	aww shit decide watch grave fireflies	37	60

		Target	ids	date	flag	user	text	final_tweet	final_text_length	text_length
				PDT 2009						
	812540	1	1548528643	Fri Apr 17 21:12:19 PDT 2009	NO_QUERY	xCAROxx	i love movies with sexy people in them. theyr	love movies sexy people theyre quite delicious	65	98
	496806	0	2185683159	Mon Jun 15 17:45:00 PDT 2009	NO_QUERY	oliviaistaken	Going to tweet all I can before "Downtime	going tweet quot downtime quot	30	53

50000 rows × 9 columns

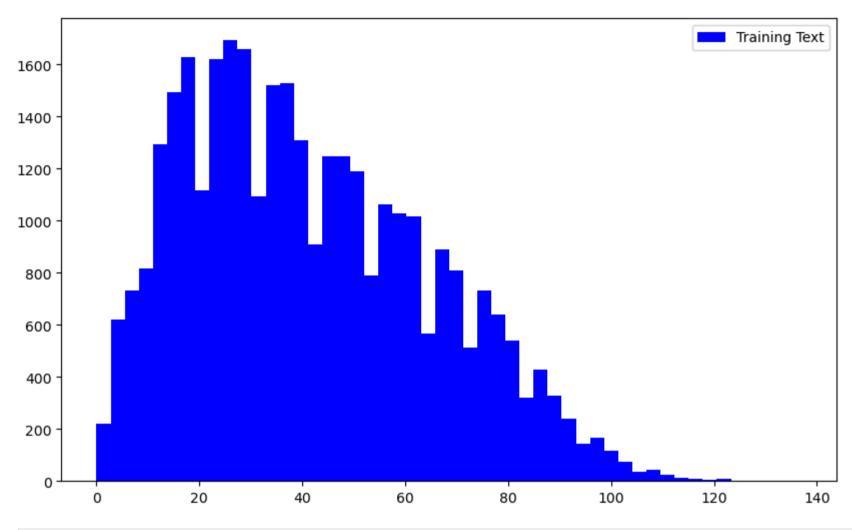
Working On model Implementation

```
In [702... X = sample_data["final_tweet"]
Y = sample_data["Target"]

In [703... X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.33, stratify = Y, random_state = 42)

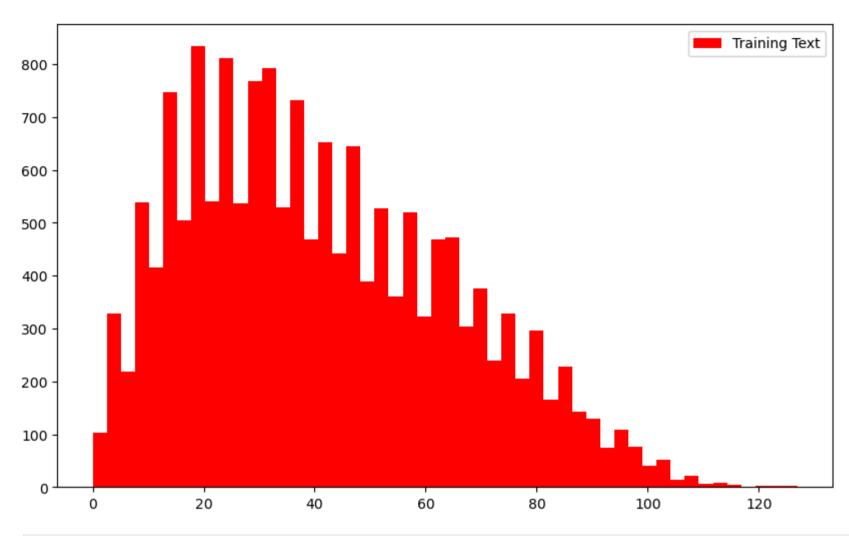
In [704... length_train = X_train.str.len()
plt.figure(figsize=(10,6))
plt.hist(length_train, bins=50, label="Training Text", color = "blue")
plt.legend()

Out[704... <matplotlib.legend.Legend at 0x7195d8fd9e90>
```



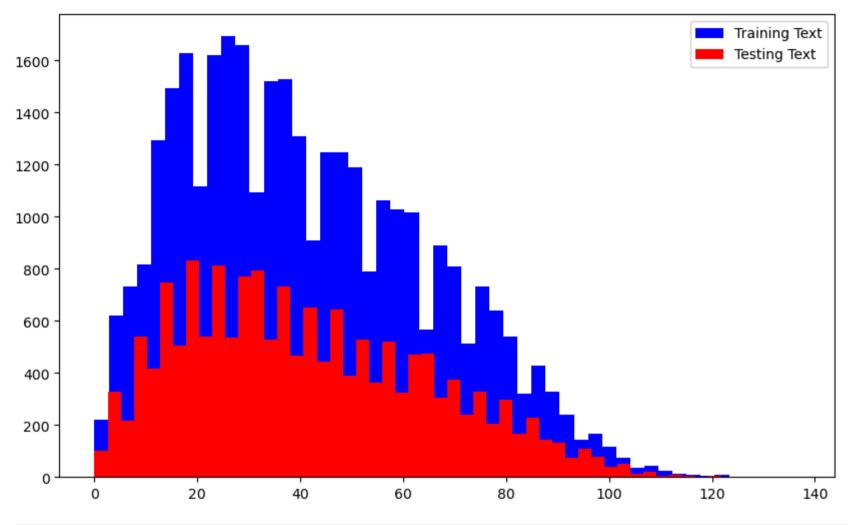
```
In [705... length_test = X_test.str.len()
   plt.figure(figsize=(10,6))
   plt.hist(length_test, bins=50, label="Training Text", color = "red")
   plt.legend()
```

Out[705... <matplotlib.legend.Legend at 0x7195d8fbf250>



```
In [706... length_train = X_train.str.len()
    length_test = X_test.str.len()
    plt.figure(figsize=(10,6))
    plt.hist(length_train, bins=50, label='Training Text', color = "blue")
    plt.hist(length_test, bins=50, label='Testing Text', color = "red")
    plt.legend()
```

Out[706... <matplotlib.legend.Legend at 0x7195d8be9e90>



In [707... sample_data

Out[707		Target	ids	date	flag	user	text	final_tweet	final_text_length	text_length
	1442958	1	2062053422	Sat Jun 06 22:13:19 PDT 2009	NO_QUERY	KellyG5	http://twitpic.com/1uoev - ha ha what are you	ha ha saying lol cute eh lol	28	91
	34148	0	1564713481	Mon Apr 20 04:39:20 PDT 2009	NO_QUERY	cl4irethomas	@AnnieC1 could do with some sympathy she's	could sympathy work sick today offers	37	85
	789182	0	2325319004	Thu Jun 25 05:13:08 PDT 2009	NO_QUERY	acullenatheart	my iPhone is attempting to kill itself. Can't	iphone attempting kill ca wait get home try fix	47	78
	724967	0	2262207910	Sat Jun 20 22:04:45 PDT 2009	NO_QUERY	locn	Zion and Vegas was so much awesomeness. Too ba	zion vegas much awesomeness bad stupid camera	60	87
	956955	1	1825348470	Sun May 17 05:21:54 PDT 2009	NO_QUERY	CarlaJaynee	Aww, 100 followers! I feel loved I love you g	aww followers feel loved love guys lt x	39	57
	•••								•••	
	1033361	1	1955775749	Thu May 28 20:36:19 PDT 2009	NO_QUERY	admance	@artfanatic411 Yes, I didTwitterholic is my	yes twitterholic middle name	28	59
	1413067	1	2056772328	Sat Jun 06 11:45:54 PDT 2009	NO_QUERY	Irv25	@David_DB your welcome!!!	welcome	7	26
	416893	0	2061407079	Sat Jun 06 20:48:16	NO_QUERY	kulex	Aww shit. Why did I decide to watch Grave of t	aww shit decide watch grave fireflies	37	60

	Target	ids	date	flag	user	text	final_tweet	final_text_length	text_length
			PDT 2009						
812540	1	1548528643	Fri Apr 17 21:12:19 PDT 2009	NO_QUERY	xCAROxx	i love movies with sexy people in them. theyr	love movies sexy people theyre quite delicious	65	98
496806	0	2185683159	Mon Jun 15 17:45:00 PDT 2009	NO_QUERY	oliviaistaken	Going to tweet all I can before "Downtime	going tweet quot downtime quot	30	53

50000 rows × 9 columns

```
In [714... X = sample_data["final_tweet"]
    Y = sample_data["Target"]
    X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.33, stratify = Y, random_state = 42)

In [716... vectorizer = TfidfVectorizer()
    X_train = vectorizer.fit_transform(X_train)
    X_test = vectorizer.transform(X_test)
In [718... print(X_train)
```

```
(0, 4750)
              0.285336446729736
(0, 13684)
              0.3571419903170213
(0, 15232)
              0.41077986959540397
(0, 679)
              0.2536745622290616
(0, 24914)
              0.30743056300545
(0, 24298)
              0.30743056300545
(0, 23746)
              0.1755829535811562
(0, 23667)
              0.23216643872413337
(0, 14500)
              0.4201237923198626
(0, 26286)
              0.2565214670876748
(0, 15160)
              0.20463829354561308
(1, 1538)
              0.7057499015330406
(1, 26638)
              0.708461062081822
(2, 2025)
              0.1809951294187414
(2, 13615)
              0.16661832522136755
(2, 25467)
              0.2884551139570287
(2, 17752)
              0.3083834631458847
(2, 25839)
              0.3083834631458847
(2, 23941)
              0.22154464518925604
(2, 13748)
              0.2034251018597497
(2, 18675)
              0.5605181479153221
(2, 6031)
              0.26852676476817267
(2, 10298)
              0.3083834631458847
              0.24521209081465745
(2, 14010)
(2, 8249)
              0.18463930820371377
(33495, 17475)
                      0.6806719696195248
(33496, 11714)
                      0.4274372425922687
(33496, 14754)
                      0.4496949679040325
(33496, 4542) 0.2735308908606615
(33496, 25838)
                      0.3523727510616534
(33496, 7347) 0.2889628867829605
(33496, 19552)
                      0.3143566458210111
(33496, 24954)
                      0.2448853500077418
(33496, 13542)
                      0.3716243147586079
(33496, 9583) 0.18892596132645836
(33497, 20196)
                      0.32397996716544447
(33497, 8689) 0.45940091706545544
(33497, 24983)
                      0.3551275288754126
                      0.5881224732091495
(33497, 23415)
(33497, 7561) 0.24946065362374423
```

```
(33497, 18012)
                     0.3270541968062426
(33497, 18909)
                     0.20685528192707628
(33499, 8356) 0.40504391492803204
(33499, 15734)
                     0.3611737155657752
(33499, 3065) 0.39536682590058786
(33499, 13947)
                     0.307411915107463
(33499, 2880) 0.38746006845083775
(33499, 22705)
                     0.37778297942339356
(33499, 22937)
                     0.3556299527749894
(33499, 9583) 0.1880372678347481
```

In [720... print(X_test)

```
(0, 22263)
              0.424269200240435
(0, 22250)
              0.42714900686286034
(0, 19900)
              0.37114276033781674
(0, 16368)
              0.25543338875565047
(0, 13463)
              0.4636078656721006
(0, 9984)
              0.27011334705212664
              0.3829526519718875
(0, 4631)
(1, 13782)
              0.3219855368734756
(1, 9207)
              0.5556371646069063
(1, 6490)
              0.6053330674484935
(1, 3483)
              0.4702813336772944
(2, 22266)
              0.429223995777511
(2, 21657)
              0.24893320814803294
(2, 17959)
              0.48636034389004223
(2, 16416)
              0.31506174819544536
(2, 13716)
              0.20609627630349428
(2, 9641)
              0.36796540147042495
(2, 9361)
              0.30431548572166334
(2, 3542)
              0.3840658842043067
(3, 25799)
              0.5371959133350348
(3, 21617)
              0.6343218840433256
(3, 10800)
              0.555928321026968
(4, 26491)
              0.24329626112187816
(4, 25815)
              0.3936682943589734
(4, 25219)
              0.2605252598641996
(16494, 4411) 0.38609042754700823
(16495, 23613)
                      0.2356066948788948
(16495, 14881)
                      0.32331787856349326
(16495, 12174)
                      0.4664626217278268
                      0.5200245255865704
(16495, 10800)
(16495, 9725) 0.2657362403582129
(16495, 7338) 0.4173257378599872
(16495, 2377) 0.32736161828872
(16496, 23344)
                      0.3799237250395624
(16496, 22923)
                      0.3433036406225573
(16496, 18885)
                      0.40095994399442025
(16496, 18530)
                      0.3812104680138591
(16496, 17453)
                      0.4005140111634836
                      0.3390542739843072
(16496, 13782)
(16496, 8928) 0.3953998002411207
```

```
(16497, 26138)
                                0.64034562602012
          (16497, 19484)
                                0.47514622586541305
          (16497, 9442) 0.25555319653740594
           (16497, 4320) 0.41519942247531755
          (16497, 698) 0.3556621242112117
           (16498, 1982) 1.0
          (16499, 26084)
                                0.5666576677269365
          (16499, 21535)
                                0.5822216854913147
          (16499, 20437)
                                0.29319852459662654
          (16499, 1054) 0.5039361286158026
In [725... lg = LogisticRegression(max_iter = 1000)
         lg.fit(X train, Y train)
         X train predict = lg.predict(X train)
         print("SVM Accuracy Training Data:",accuracy score(Y train, X train predict))
        SVM Accuracy Training Data: 0.8317910447761194
In [746... # Accuracy
         X test predict = lq.predict(X test)
         print("SVM Accuracy Tresting Data:",accuracy score(Y_test, X_test_predict))
        SVM Accuracy Tresting Data: 0.7531515151515151
In [724... # Training Data
         svm = SVC()
         svm.fit(X train, Y train)
         X train predict = svm.predict(X train)
         print("SVM Accuracy Training Data:", accuracy score(Y_train, X_train_predict))
         print("SVM Classification Report:\n", classification report(Y train, X train predict))
        SVM Accuracy Training Data: 0.9574626865671642
        SVM Classification Report:
                       precision
                                     recall f1-score
                                                        support
                   0
                            0.95
                                     0.96
                                                0.96
                                                        16875
                   1
                            0.96
                                     0.95
                                               0.96
                                                        16625
                                                0.96
                                                        33500
            accuracy
                                               0.96
                                                        33500
           macro avg
                            0.96
                                     0.96
        weighted avg
                                     0.96
                                               0.96
                           0.96
                                                         33500
```

```
In [748... # Testing Data
          svm = SVC()
         svm.fit(X train, Y train)
         X test predict = svm.predict(X test)
         print("SVM Accuracy Testing Data:", accuracy score(Y_test, X_test_predict))
         print("SVM Classification Report:\n", classification report(Y test, X test predict))
        SVM Accuracy Testing Data: 0.7519393939393939
        SVM Classification Report:
                        precision
                                     recall f1-score
                                                        support
                    0
                            0.76
                                      0.75
                                                0.75
                                                           8312
                    1
                            0.75
                                      0.76
                                                0.75
                                                           8188
                                                0.75
                                                         16500
             accuracy
                            0.75
                                      0.75
                                                0.75
                                                         16500
           macro avq
        weighted avg
                            0.75
                                      0.75
                                                0.75
                                                         16500
In [749... # Training Data
         dt = DecisionTreeClassifier()
         dt.fit(X train, Y train)
         X train predict = dt.predict(X train)
         print("Decision Tree Accuracy Training Data", accuracy_score(Y_train, X_train_predict))
         print("Decision Tree Classification Report:\n", classification report(Y train, X train predict))
        Decision Tree Accuracy Training Data 0.9932835820895523
        Decision Tree Classification Report:
                                     recall f1-score
                        precision
                                                        support
                    0
                            0.99
                                      1.00
                                                0.99
                                                         16875
                                                0.99
                    1
                                      0.99
                            1.00
                                                         16625
                                                0.99
                                                          33500
             accuracy
                                                0.99
           macro avg
                            0.99
                                      0.99
                                                          33500
        weighted avg
                            0.99
                                      0.99
                                                0.99
                                                          33500
In [750... # Testing Data
         dt = DecisionTreeClassifier()
         dt.fit(X_train, Y_train)
```

```
X test predict = dt.predict(X test)
         print("Decision Tree Accuracy Testing Data", accuracy score(Y test, X test predict))
         print("Decision Tree Classification Report:\n", classification report(Y test, X test predict))
        Decision Tree Accuracy Testing Data 0.6803030303030303
        Decision Tree Classification Report:
                        precision
                                     recall f1-score
                                                        support
                                      0.71
                   0
                            0.67
                                                0.69
                                                          8312
                                      0.65
                   1
                            0.69
                                                0.67
                                                          8188
                                                0.68
                                                         16500
            accuracy
                                                0.68
                                                         16500
           macro avq
                            0.68
                                      0.68
                                      0.68
        weighted avg
                            0.68
                                                0.68
                                                         16500
In [751... # Training Data
         from sklearn.neighbors import KNeighborsClassifier
         knn = KNeighborsClassifier()
         knn.fit(X train, Y train)
         X train predict = knn.predict(X train)
         print("Decision Tree Accuracy Training Data", accuracy_score(Y_train, X_train_predict))
         print("Decision Tree Classification Report:\n", classification report(Y train, X train predict))
        Decision Tree Accuracy Training Data 0.726089552238806
        Decision Tree Classification Report:
                                     recall f1-score
                        precision
                                                        support
                   0
                            0.73
                                      0.72
                                                0.73
                                                         16875
                   1
                            0.72
                                      0.74
                                                0.73
                                                         16625
                                                0.73
                                                         33500
            accuracy
                                      0.73
                                                         33500
                                                0.73
           macro avq
                            0.73
                                      0.73
        weighted avg
                            0.73
                                                0.73
                                                         33500
         # Testing Data
In [752...
         knn = KNeighborsClassifier()
         knn.fit(X train, Y train)
         X test predict = knn.predict(X test)
```

```
print("KNN Accuracy Testing Data:", accuracy score(Y test, X test predict))
         print("KNN Classification Report:\n", classification report(Y test, X test predict))
        KNN Accuracy Testing Data: 0.56969696969697
        KNN Classification Report:
                       precision
                                     recall f1-score
                                                        support
                   0
                            0.58
                                      0.56
                                                0.57
                                                          8312
                   1
                                     0.58
                                                0.57
                            0.56
                                                          8188
                                                0.57
            accuracy
                                                         16500
                           0.57
                                      0.57
                                                0.57
                                                         16500
           macro avq
                                     0.57
        weighted avg
                            0.57
                                                0.57
                                                         16500
         # Training Data
In [753...
         xgb = XGBClassifier(use label encoder=False, eval metric='logloss')
         xgb.fit(X train, Y train)
         X train predict = xqb.predict(X train)
         print("XGBoost Accuracy Training Data:", accuracy_score(Y_train, X_train_predict))
         print("XGBoost Classification Report:\n", classification report(Y train, X train predict))
        /home/ubuntu/anaconda3/lib/python3.11/site-packages/xgboost/core.py:158: UserWarning: [19:50:44] WARNING: /workspac
        e/src/learner.cc:740:
        Parameters: { "use label encoder" } are not used.
          warnings.warn(smsg, UserWarning)
        XGBoost Accuracy Training Data: 0.7711044776119403
        XGBoost Classification Report:
                       precision
                                     recall f1-score
                                                        support
                   0
                            0.83
                                      0.68
                                                0.75
                                                         16875
                                     0.86
                   1
                            0.73
                                                0.79
                                                         16625
                                                0.77
                                                         33500
            accuracy
                                      0.77
                                                0.77
                                                         33500
           macro avq
                            0.78
        weighted avg
                            0.78
                                      0.77
                                                0.77
                                                         33500
In [754... # Testing Data
         xgb = XGBClassifier(use label encoder=False, eval metric='logloss')
```

```
xgb.fit(X train, Y train)
         X test predict = xgb.predict(X test)
         print("XGBoost Accuracy Testing:", accuracy score(Y test, X test predict))
         print("XGBoost Classification Report:\n", classification report(Y test, X test predict))
        /home/ubuntu/anaconda3/lib/python3.11/site-packages/xgboost/core.py:158: UserWarning: [19:50:56] WARNING: /workspac
        e/src/learner.cc:740:
        Parameters: { "use label encoder" } are not used.
          warnings.warn(smsg, UserWarning)
        XGBoost Accuracy Testing: 0.7246060606060606
        XGBoost Classification Report:
                                     recall f1-score
                       precision
                                                        support
                   0
                           0.77
                                     0.64
                                                0.70
                                                          8312
                   1
                           0.69
                                     0.81
                                               0.74
                                                          8188
                                               0.72
                                                         16500
            accuracy
                                     0.73
                           0.73
                                                0.72
                                                         16500
           macro avq
        weighted avg
                           0.73
                                     0.72
                                                0.72
                                                         16500
In [755... # Training Data
         rf = RandomForestClassifier()
         rf.fit(X train, Y train)
         X train predict rf = rf.predict(X train)
         # Evaluate the model on the training data
         print("Random Forest Accuracy on Training Data:", accuracy score(Y train, X train predict rf))
```

print("Random Forest Classification Report on Training Data:\n", classification report(Y train, X train predict rf)

```
Random Forest Accuracy on Training Data: 0.9932835820895523
        Random Forest Classification Report on Training Data:
                                     recall f1-score
                        precision
                                                        support
                    0
                            0.99
                                      1.00
                                                0.99
                                                         16875
                                      0.99
                    1
                            1.00
                                                0.99
                                                         16625
                                                0.99
                                                         33500
             accuracy
                            0.99
                                      0.99
                                                0.99
                                                         33500
           macro avq
                                      0.99
        weighted avg
                            0.99
                                                0.99
                                                         33500
In [756... # Testing Data
          rf = RandomForestClassifier()
         rf.fit(X train, Y train)
         X test predict = rf.predict(X test)
         # Evaluate the model on the training data
         print("Random Forest Accuracy on Training Data:", accuracy_score(Y_test, X_test_predict))
         print("Random Forest Classification Report on Training Data:\n", classification report(Y test, X test predict))
        Random Forest Accuracy on Training Data: 0.7333939393939394
        Random Forest Classification Report on Training Data:
                        precision
                                     recall f1-score support
                    0
                            0.73
                                      0.75
                                                0.74
                                                          8312
                                      0.71
                    1
                            0.74
                                                0.73
                                                          8188
                                                0.73
                                                         16500
             accuracy
                                      0.73
                                                0.73
                                                         16500
           macro avg
                            0.73
        weighted avg
                            0.73
                                      0.73
                                                0.73
                                                         16500
In [757...
         #####################################logistic regressio
         # Saving Model for Future use Im just saving Logistic Regression Model
In [758...
         import pickle
In [759...
```

```
In [760... save model lg = 'trained Model logistic regression.sav'
         pickle.dump(lg, open(save model lg, "wb"))
In [761... save model xgb = 'trained Model XGBoost.sav'
         pickle.dump(xqb, open(save model xqb, "wb"))
In [762... save model rf = 'trained Model Random Forest.sav'
         pickle.dump(rf, open(save model rf, "wb"))
In [763... save model knn = 'trained Model KNN.sav'
         pickle.dump(knn, open(save model knn, "wb"))
In [764... save model svm = 'trained Model SVM.sav'
         pickle.dump(svm, open(save model svm, "wb"))
In [765... save model dt = 'trained Model DT.sav'
         pickle.dump(dt, open(save model dt, "wb"))
In [766... load saved model = pickle.load(open(save model dt, "rb")) # loading models
In [767... ls
         nltk data/
                                                      trained Model Random Forest.sav
                                                      trained Model.sav
         'Screenshot from 2024-06-30 01-59-16.png'
         'Sentimental Analysis.ipynb'
                                                      trained Model SVM.sav
                                                      trained Model XGBoost.sav
         trained Model DT.sav
         trained Model KNN.sav
                                                      tweets.csv
         trained Model logistic regression.sav
                                                      tweets.zip
In [772... new_input = X_test[29]
In [774... predict = load saved model.predict(new input)
         if predict[0] == 0:
              print("-ve Tweet")
          else:
              print("+ve Tweet")
```