```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import nltk
from nltk.stem import WordNetLemmatizer
import re #regular expression(to rmove special characers)
df=pd.read_csv("/content/twitter_validation.csv",encoding="ISO-8859-1",header=None) #to read the emojis
df.columns=['id','location','target','text']
df
```

	id	location	target	text
0	3364	Facebook	Irrelevant	I mentioned on Facebook that I was struggling
1	352	Amazon	Neutral	BBC News - Amazon boss Jeff Bezos rejects clai
2	8312	Microsoft	Negative	@Microsoft Why do I pay for WORD when it funct
3	4371	CS-GO	Negative	CSGO matchmaking is so full of closet hacking,
4	4433	Google	Neutral	Now the President is slapping Americans in the
995	4891	GrandTheftAuto(GTA)	Irrelevant	âï¸ Toronto is the arts and culture capital
996	4359	CS-GO	Irrelevant	this is actually a good move tot bring more vi
997	2652	Borderlands	Positive	Today sucked so itâs time to drink wine n pl
998	8069	Microsoft	Positive	Bought a fraction of Microsoft today. Small wins.
999	6960	johnson&johnson	Neutral	Johnson & Johnson to stop selling talc baby po

1000 rows \times 4 columns

df.head()

	id	location	target	text
0	3364	Facebook	Irrelevant	I mentioned on Facebook that I was struggling
1	352	Amazon	Neutral	BBC News - Amazon boss Jeff Bezos rejects clai
2	8312	Microsoft	Negative	@Microsoft Why do I pay for WORD when it funct
3	4371	CS-GO	Negative	CSGO matchmaking is so full of closet hacking,
4	4433	Google	Neutral	Now the President is slapping Americans in the

df.tail()

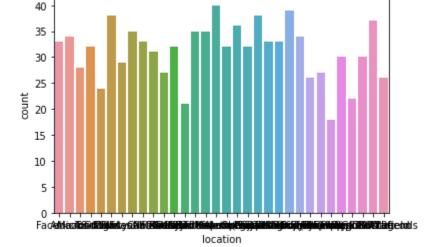
	id	location	target	text
995	4891	GrandTheftAuto(GTA)	Irrelevant	âï¸ Toronto is the arts and culture capital
996	4359	CS-GO	Irrelevant	this is actually a good move tot bring more vi
997	2652	Borderlands	Positive	Today sucked so itâs time to drink wine n pl
998	8069	Microsoft	Positive	Bought a fraction of Microsoft today. Small wins.
000	6060	iohnson Ciohnson	Noutral	Johnson & Johnson to stan calling tale baby no

```
×
                                                                           ✓ 0s completed at 9:44 PM
df.columns
    Index(['id', 'location', 'target', 'text'], dtype='object')
df.shape
    (1000, 4)
df.isna().sum()
    id
                0
                0
    location
                0
    target
                0
    text
    dtype: int64
df['location'].value_counts()
                                         40
    RedDeadRedemption(RDR)
    johnson&johnson
                                         39
                                         38
                                         38
    PlayerUnknownsBattlegrounds(PUBG)
                                         37
    LeagueOfLegends
                                         36
    ApexLegends
    TomClancysRainbowSix
                                         35
    Nvidia
                                         35
                                         35
    GrandTheftAuto(GTA)
                                         34
    Amazon
                                         34
    Fortnite
                                         33
    Facebook
                                         33
    PlayStation5(PS5)
```

33 AssassinsCreed 33 Borderlands 32 Overwatch 32 Hearthstone Verizon 32 CS-G0 32 CallOfDuty 31 30 Cyberpunk2077 WorldOfCraft 30 MaddenNFL 29 28 Microsoft 27 Dota2 27 CallOfDutyBlackopsColdWar Xbox(Xseries) 26 Battlefield 26 24 Google 22 TomClancysGhostRecon NBA2K 21 18 HomeDepot Name: location, dtype: int64 import seaborn as sns

sns.countplot(x='location',data=df)

<matplotlib.axes._subplots.AxesSubplot at 0x7f637a367e80>



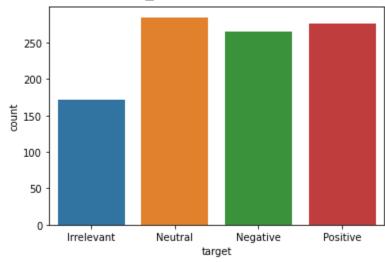
df['target'].value_counts()

Neutral 285 Positive 277 Negative 266 Irrelevant 172

Name: target, dtype: int64

sns.countplot(x='target',data=df)

<matplotlib.axes._subplots.AxesSubplot at 0x7f637a34c1f0>



```
# drop irrelevent in target
df.drop(df.index[(df['target']=='Irrelevant')],axis=0,inplace=True)
df
# reset the index value
df.reset_index(drop=True,inplace=True)
df
# drop id and location
df.drop(['id','location'],axis=1,inplace=True)
df
```

target Neutral BBC News - Amazon boss Jeff Bezos rejects clai... Negative @Microsoft Why do I pay for WORD when it funct... Negative CSGO matchmaking is so full of closet hacking,... Neutral Now the President is slapping Americans in the...

```
    Negative Hi @EAHelp lâve had Madeleine McCann in my c...
    ... ...
    Negative Please explain how this is possible! How can t...
    Positive Good on Sony. As much as I want to see the new...
    Positive Today sucked so itâs time to drink wine n pl...
    Positive Bought a fraction of Microsoft today. Small wins.
    Neutral Johnson & Johnson to stop selling talc baby po...
```

828 rows × 2 columns

```
# Target====>positive (+1) negative (-1) neutral (0)
# here we use map function
df['target']=df['target'].map({'Positive':1,'Negative':-1,'Neutral':0})
df
```

	target	text
0	0	BBC News - Amazon boss Jeff Bezos rejects clai
1	-1	@Microsoft Why do I pay for WORD when it funct
2	-1	CSGO matchmaking is so full of closet hacking,
3	0	Now the President is slapping Americans in the
4	-1	Hi @EAHelp lâve had Madeleine McCann in my c
823	-1	Please explain how this is possible! How can t
824	1	Good on Sony. As much as I want to see the new
825	1	Today sucked so itâs time to drink wine n pl
826	1	Bought a fraction of Microsoft today. Small wins.
827	0	Johnson & Johnson to stop selling talc baby po

828 rows × 2 columns

```
nltk.download('stopwords')
nltk.download('punkt')
nltk.download('averaged perceptron tagger')
nltk.download('wordnet')
nltk.download('omw-1.4')
    [nltk data] Downloading package stopwords to /root/nltk data...
    [nltk data] Unzipping corpora/stopwords.zip.
    [nltk_data] Downloading package punkt to /root/nltk_data...
    [nltk data] Unzipping tokenizers/punkt.zip.
    [nltk data] Downloading package averaged perceptron tagger to
                    /root/nltk data...
    [nltk data]
    [nltk data] Unzipping taggers/averaged perceptron tagger.zip.
    [nltk_data] Downloading package wordnet to /root/nltk_data...
    [nltk data] Downloading package omw-1.4 to /root/nltk data...
    True
```

```
# assign text into variable
tweets=df.text
tweets
    0
           BBC News - Amazon boss Jeff Bezos rejects clai...
    1
           @Microsoft Why do I pay for WORD when it funct...
           CSGO matchmaking is so full of closet hacking,...
           Now the President is slapping Americans in the...
           Hi @EAHelp Iâve had Madeleine McCann in my c...
    823
           Please explain how this is possible! How can t...
    824
           Good on Sony. As much as I want to see the new...
    825
           Today sucked so itâs time to drink wine n pl...
    826
           Bought a fraction of Microsoft today. Small wins.
    827
           Johnson & Johnson to stop selling talc baby po...
    Name: text, Length: 828, dtype: object
# tokenisation
from nltk import TweetTokenizer
tk=TweetTokenizer()
tweets=tweets.apply(lambda x:tk.tokenize(x)).apply(lambda x:" ".join(x))
tweets
           BBC News - Amazon boss Jeff Bezos rejects clai...
           @Microsoft Why do I pay for WORD when it funct...
           CSGO matchmaking is so full of closet hacking ...
           Now the President is slapping Americans in the...
           Hi @EAHelp Iâ ve had Madeleine McCann in m...
    823
           Please explain how this is possible! How can ...
           Good on Sony . As much as I want to see the ne...
    824
    825
           Today sucked so itâ s time to drink wine n...
           Bought a fraction of Microsoft today . Small w...
    827
           Johnson & Johnson to stop selling talc baby po...
    Name: text, Length: 828, dtype: object
# remove the special characters
tweets=tweets.str.replace('[^a-zA-Z0-9]+',' ') # ^ denote except
tweets
    <ipython-input-16-bc435e4ca2fe>:2: FutureWarning: The default value of regex will change from True to False in a future version.
      tweets=tweets.str.replace('[^a-zA-Z0-9]+',' ') # ^ denote except
           BBC News Amazon boss Jeff Bezos rejects claims...
            Microsoft Why do I pay for WORD when it funct...
    1
           CSGO matchmaking is so full of closet hacking ...
           Now the President is slapping Americans in the...
           Hi EAHelp I ve had Madeleine McCann in my cell...
    823
           Please explain how this is possible How can th...
           Good on Sony As much as I want to see the new ...
    824
    825
           Today sucked so it s time to drink wine n play...
    826
            Bought a fraction of Microsoft today Small wins
    827
           Johnson Johnson to stop selling talc baby powd...
    Name: text, Length: 828, dtype: object
# collect character length above 3
```

from nltk.tokenize import word tokenize

tweets=tweets.apply(lambda x:' '.join([w for w in word tokenize(x) if len(w)>=3]))

```
tweets
    0
           BBC News Amazon boss Jeff Bezos rejects claims...
           Microsoft Why pay for WORD when functions poor...
           CSGO matchmaking full closet hacking truly awf...
           Now the President slapping Americans the face ...
           EAHelp had Madeleine McCann cellar for the pas...
    823
           Please explain how this possible How can they ...
    824
           Good Sony much want see the new PS5 what going...
           Today sucked time drink wine play borderlands ...
    825
    826
                  Bought fraction Microsoft today Small wins
    827
           Johnson Johnson stop selling talc baby powder ...
    Name: text, Length: 828, dtype: object
# stemming
# snowball Stemmer
from nltk import SnowballStemmer
from nltk.tokenize import word tokenize
stemmer=SnowballStemmer('english')
tweets=tweets.apply(lambda x:[stemmer.stem(i.lower()) for i in tk.tokenize(x)]).apply(lambda x:' '.join(x))
tweets
    0
           bbc news amazon boss jeff bezo reject claim co...
    1
           microsoft whi pay for word when function poor ...
                csgo matchmak full closet hack truli aw game
           now the presid slap american the face that rea...
           eahelp had madelein mccann cellar for the past...
    823
           pleas explain how this possibl how can they le...
           good soni much want see the new ps5 what go ri...
    824
    825
           today suck time drink wine play borderland unt...
                   bought fraction microsoft today small win
    826
    827
           johnson johnson stop sell talc babi powder and...
    Name: text, Length: 828, dtype: object
#remove the stopwords
from nltk.corpus import stopwords
stop=stopwords.words('english')
tweets=tweets.apply(lambda x:[i for i in tk.tokenize(x) if i not in stop]).apply(lambda x:' '.join(x))
tweets
    0
           bbc news amazon boss jeff bezo reject claim co...
           microsoft whi pay word function poor samsungus...
                csgo matchmak full closet hack truli aw game
           presid slap american face realli commit unlaw ...
           eahelp madelein mccann cellar past year littl ...
    823
           pleas explain possibl let compani overcharg sc...
    824
           good soni much want see new ps5 go right much ...
    825
           today suck time drink wine play borderland sun...
                   bought fraction microsoft today small win
    826
           johnson johnson stop sell talc babi powder can...
    827
    Name: text, Length: 828, dtype: object
# vectorisation
from sklearn.feature extraction.text import TfidfVectorizer
vec=TfidfVectorizer()
train data=vec.fit transform(tweets)
print(train data)
```

```
(0, 691)
                     0.2608257828483461
       (0, 1004)
                     0.2608257828483461
       (0, 1130)
                     0.23509805002803952
       (0, 1996)
                     0.13277165480466424
       (0, 309)
                     0.22681557001542715
       (0, 860)
                     0.17354914655342313
       (0, 807)
                     0.21432663830218204
       (0, 2761)
                     0.2608257828483461
       (0, 568)
                     0.2608257828483461
       (0, 1833)
                     0.24577602391989378
       (0, 633)
                     0.22681557001542715
       (0, 376)
                     0.1515362387424402
       (0, 2287)
                     0.38864111655856126
       (0, 538)
                     0.49155204783978756
       (1, 797)
                     0.4055823664694651
       (1, 2891)
                     0.4055823664694651
       (1, 2558)
                     0.3821800909185634
       (1, 1405)
                     0.4055823664694651
       (1, 3679)
                     0.36557591217188057
       (1, 2462)
                     0.3126902562590763
       (1, 3639)
                     0.26216072802580975
       (1, 2155)
                     0.24555654927912696
       (2, 1427)
                     0.1689251539717079
       (2, 486)
                     0.36574263611909275
       (2, 3432)
                     0.36574263611909275
       (825, 3393)
                     0.3395996844494919
       (825, 3383)
                     0.2560582225152134
       (825, 631)
                     0.22981061112100945\\
       (825, 997)
                     0.2315686698425631
       (825, 3373)
                     0.21750175079084832
       (825, 3226)
                     0.2904718522758868
       (825, 2527)
                     0.17148706662740873
       (826, 1381)
                     0.5079831062080814
       (826, 3070)
                     0.47867226429410115
       (826, 636)
                     0.4174215841659411
       (826, 3650)
                     0.353278941165688
       (826, 3383)
                     0.34523850330234374
       (826, 2155)
                     0.3075542453642147
       (827, 195)
                     0.3283693467320579
       (827, 1132)
                     0.3283693467320579
       (827, 2808)
                     0.3283693467320579
       (827, 143)
                     0.3283693467320579
       (827, 712)
                     0.2770320970909926
       (827, 2946)
                     0.2635889502019104
       (827, 3274)
                     0.2635889502019104
       (827, 3193)
                     0.2311987519368367
       (827, 1686)
                     0.15963411936668057
       (827, 2576)
                     0.24870786898500463
       (827, 506)
                     0.23743856420618148
       (827, 1854)
                     0.3947412386878786
train data.shape
# 828 sentence and 3783 uniquie value
     (828, 3783)
```

array([0, -1, -1, 0, -1, 1, 1, -1, 1, 1, -1, 0, -1, 1, 1, -1,

1, -1, -1, 0, -1, 0, 0, -1, -1, 1, 1, -1, 1, -1, 0, 0, 1,

y=df['target'].values

```
0,
                      -1, -1,
      -1, -1,
             1,
                1,
                   -1,
                      0, -1,
                             0, -1, 1, -1, -1,
                                              1,
      0, 1, 1, 0, 1, 0, -1, -1, 0, 0, -1, 1, -1, -1,
                   0, 1, 0,
                             1, -1, 0, 0, 0, -1, 0, -1,
      0, -1, 1, 1,
             0, -1, -1,
                      1, -1,
                             0, -1, 1,
                                       0, -1,
       0, 0, 0, 0, 1, 0, 1, 1, -1, 0, 0, 0,
      0, -1, 0, -1, -1, -1, 1, 1, 1, 0, 0, 1,
   0, -1, -1,
            0, 1, 1, 0, 1, 1, 0, 0, -1, -1, -1, 1,
0, 0, 1, 1, 1, 1, -1, 1, 1, 0, -1, -1, -1, 1, 1, -1, -1,
1, 1, -1, 1, 1, -1, 1, 0, -1, 0, 0, 1, -1, 1,
                   0, 0, 1, -1,
                                 0, 1,
1, 1, -1, 0, 1, 0, -1, 0, -1, 1, 1, -1, -1, -1, 1, -1, 0,
1, 0, 0, -1, 1, -1, 1, -1, 0, 0, 1, -1, 0, -1, 1, -1, 1,
1, 1, 1, 1, -1, -1, 1, -1, 0, 0, 0, 1, 0,
0, 0, 0, -1, 1, -1, -1, 1, 1, 0, 0, -1, -1, -1,
-1, 1, 0, -1, -1, -1, 1, 0, 0, -1, 1, 1, 0, 1,
                0, -1, -1, 1, 1, 1, 1,
      0, 1, -1,
                                       0, -1,
1, -1, -1, -1, 1, 0, 1, -1, 0, -1, 1, 1, 1, 1, 0, 0, 0,
-1, 1, 0, -1, 1, 0, -1, -1, -1, -1, -1, 0, 0,
-1, -1, 0, -1, 0, 0, -1, 1, -1, 1, 1, 1,
                                       0, 1,
1, 0, 0, 0, 0, 0, 0, 0, -1, -1, 1, 1, 0, -1, -1, 1,
1, -1, 1, 1, 1, 1, 0, -1, 1, 0, 0, 1, 1,
-1, -1, -1, -1, 0, 1, -1, -1, 1, 0, 0, -1, -1,
-1, -1, 0, 0, 1, -1, -1, -1, 0, 0, 0, -1, -1, 1, -1, 0, -1,
0, 1, -1, 0, 1, 1, -1, 0, 0, 1, -1, -1, 0, 0, -1, 1, -1,
0, -1, -1, -1, 1, -1, 1, -1, 1, -1, -1, 0, -1, 0, -1, 1, -1,
0, -1, -1, 0, 0, 1, -1, 1, 0, 0, 0, 0, -1, 0,
-1, 0, 1, 0, 0, -1, 0, 1, 0, 0, 0, 0, 0, 1,
1, 0, -1, 1,
             0, 0, -1, 1, 0, 0, -1, 0, -1, 0,
      0, 0, 0, 0, 1, 1, 1, -1, -1, 0, 1, 0,
1, 0, 1, -1, -1, 0, 1, -1, 1, -1, 0, 1, 1, 0,
0, -1, 0, 0, -1, 1, -1, 0, 1, 1, 1, 1, 0, -1,
1, 1, 1, -1, 0, 1, 0, 0, -1, -1, -1, 0, 1, 0, -1, 1, 1,
1, 0, 1, -1, 0, -1, 0, -1, 0, 0, 1, -1, 1, 1,
-1, -1, -1, -1, 1, 1, 1, 0, -1, -1, 1, -1, -1,
0, -1, 0, 1, -1,
                0, 1, -1, 0,
                             0, 1, -1,
                                       0, -1,
1, 0, 1, -1, 0, 0, 1, 0, 0, -1, 1, 0, -1, -1, 0, 0,
1, -1, -1, -1, -1, 1, 0, 0, 1, 0, -1, 1, 1, -1, 1,
      1, 1, -1, -1, -1, 1, -1, 0, -1, 0, 0, 1,
1, -1, -1, -1, -1, -1, -1, -1, 0, -1, 0, 0,
0, -1, 0, 1, 0, -1, -1, 1, 0, 1, 0, 1, 0, -1, 1, 1,
                   0, 0,
1, -1, -1, 1, 0, 0,
                             0, -1, -1, -1, -1, 1, -1,
                          0,
1, 0, -1, 1, 1, -1, 1, 0, 0, 1, -1, 0, -1, 0, 1, 1, 0,
-1, 1, -1, -1, 0, -1, 0, -1, 1, 0, -1, -1, 1, -1, 0, -1,
0, 0, 0, 0, 0, 1, 0, 1, 1, 1, -1, 0, 1, 0,
1, 0, 1, 0, -1, -1, 1, 1, 1, 0, -1, 1, 1, -1, -1,
0, 1, 0, 1, 1, 0, 1, -1, 1, 1, 0])
```

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(train_data,y,test_size=0.30,random_state=42)
```

```
x_{test}
```

<249x3783 sparse matrix of type '<class 'numpy.float64'>'
with 3255 stored elements in Compressed Sparse Row format>

y_train

```
-1, 1, 0, 1, -1, -1, -1, -1, -1, -1, 0, -1, 1,
                      0, -1, 1, -1, -1,
0, 1, 1, 0, 1, -1, 1,
                    1,
                       0, 1,
                                       1,
                    0, 1, 0, -1, 0,
-1, 1, 1, 1, -1, 1,
                    0, 1, 1, 1, 1,
                                    0, 0, -1,
0, -1,
      0, 0, 0, 1,
                   1,
                      0, -1, -1, 0, 0,
                                       0, -1,
-1, -1, 0, 0, -1, -1, 0, 1, -1, -1, 1, -1, 0, 0, -1, -1, -1,
   0, -1, 0, 0, 1, 0, -1, -1, -1, 0, 1, 1, 1,
      -1, 1, -1, -1, -1,
                      0, -1, 1, 1, -1,
                                       1, -1,
1, 0, -1, 1, 1, 0, 1, -1, -1, 1, 0, 0, -1,
      1, -1, 1, 1, 0, 1, 0, -1, -1, 1,
                                       1, 1,
      0, 0, 1, -1, 0, 1, -1, 1, -1, 0,
                                       0, 1,
1, -1, 1, 1, 0, 1, 0, -1, 0, 1, 0, 0, 1, 0,
   0, -1, 1, -1, 0, 1, 1, -1, 1, -1, 0, 0, -1,
             0,
                0, 0, -1, 0, 0, -1, 0, -1, 0,
1, 0, 1, 1, 0, -1, -1, 0, -1, -1, 0, -1, 1, -1, -1, 1, 0,
      0, 0, 1, -1, 0, 1, 0, 1, 0, -1, 1, -1, -1, 0,
       0, 1, -1, 1, 0, 1, 0, 1, 0, -1, 1, -1,
                                                 1, 1,
   0, 0, -1, 0, -1, 1, 0, 1, -1, 1, 1, 1, -1, 0, -1, -1,
                                       0, 0, -1, 0, 0,
1, 1, -1, -1, 0, 1, -1, -1, 1, 0, 1,
                 0, -1, 1, 1, 0,
   0, -1, 0, 1,
                                 0, 0, 1, 1,
      0, 0, 0, 0, 1, -1, -1, 0, 0, -1, -1, -1, -1, -1,
1, -1, -1, -1, -1, -1, 1, 1, 0, 0, 1, 0, -1,
1, 0, 1, 0, 0, 0, 1, 1, -1, -1, 0, 1, -1,
-1, -1, 1, -1, 0, -1, 0, 1, 0, -1,
                                       0, 1, -1, -1, 0,
-1, 0, 1, 1, 1, 1, 1, 1, 0, -1, 0, 1, -1, -1, -1, 0,
-1, -1, 0, 1, 0, 0, 1, 0, 0, 0, 1, -1, -1, 1, -1, 0,
-1, 0, 1, 0, -1, 0, -1, 1, -1, -1, -1, 0, -1, -1, 0, 1, 0,
1, 1, 0, 1, -1, -1, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0,
0, 1, -1, 0, 0, -1, -1, 0, -1, 0, -1, 1, 0, -1, 0,
0])
```

from sklearn.svm import SVC

nb model=MultinomialNB()

- 1

0

1

accuracy

macro avg

svm model=SVC()

from sklearn.naive bayes import MultinomialNB

from sklearn.ensemble import RandomForestClassifier
from sklearn.tree import DecisionTreeClassifier

0.67

0.46

0.76

0.63

0.52

0.76

0.48

0.59

0.59

0.57

0.59

0.58

0.58

79

79

91

249

249

weighted avg	0.64	0.58	0.58	249					

<pre>MultinomialNB() ******</pre>									
	precision	recall	f1-score	support					
-1	0.60	0.54	0.57	79 70					
0 1	0.47 0.64	0.54 0.59	0.50 0.61	79 91					
accuracy			0.56	249					
macro avg	0.57	0.56	0.56	249					
weighted avg	0.57	0.56	0.56	249					

RandomForest(*****	<pre>RandomForestClassifier() ******</pre>								
	precision	recall	f1-score	support					
-1	0.51	0.73	0.60	79					
0	0.53	0.53	0.53	79					
1	0.71	0.43	0.53	91					
accuracy			0.56	249					
macro avg	0.58	0.56	0.55	249					
weighted avg	0.59	0.56	0.55	249					

<pre>DecisionTreeClassifier() *****</pre>									
	precision	recall	f1-score	support					
-1	0.47	0.56	0.51	79					
0	0.47	0.48	0.48	79					
1	0.58	0.48	0.53	91					
accuracy			0.51	249					
macro avg	0.51	0.51	0.51	249					
weighted avg	0.51	0.51	0.51	249					

Colab paid products - Cancel contracts here