```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import nltk
import re  # regular expression (for removing special charachters)
```

# we use encoding here to encode the text, data like emoji
df=pd.read\_csv('/content/drive/MyDrive/datasets/twitter\_validation.csv',encoding='ISO-8859-1',header=None)
df

1

1

	0	1	2	3
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	â □ ï, □ 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 itâ□□s 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 × 4 columns

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	â□ï¸□ 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 itâ□□s 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 × 4 columns

df.shape

(1000, 4)

df.tail()

ID location target

text

10.

## df.head()

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

## df.isna().sum()

ID 0 location 0 target 0 text 0 dtype: int64

## df.dtypes

ID int64
location object
target object
text object
dtype: object

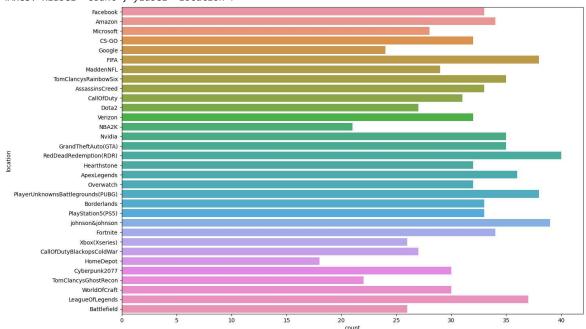
## loc=df['location'].value\_counts()

loc

₽	<pre>RedDeadRedemption(RDR)</pre>	40
	johnson&johnson	39
	FIFA	38
	PlayerUnknownsBattlegrounds(PUBG)	38
	LeagueOfLegends	37
	ApexLegends	36
	TomClancysRainbowSix	35
	Nvidia	35
	GrandTheftAuto(GTA)	35
	Amazon	34
	Fortnite	34
	Facebook	33
	PlayStation5(PS5)	33
	AssassinsCreed	33
	Borderlands	33
	Overwatch	32
	Hearthstone	32
	Verizon	32
	CS-G0	32
	CallOfDuty	31
	Cyberpunk2077	30
	WorldOfCraft	30
	MaddenNFL	29
	Microsoft	28
	Dota2	27
	CallOfDutyBlackopsColdWar	27
	Xbox(Xseries)	26
	Battlefield	26
	Google	24
	TomClancysGhostRecon	22
	NBA2K	21
	HomeDepot	18
	Name: location, dtype: int64	

sns.countplot(y=df['location'])

<Axes: xlabel='count', ylabel='location'>



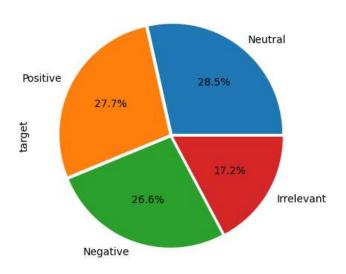
tar=df['target'].value\_counts()
tar

Neutral 285 Positive 277 Negative 266 Irrelevant 172

Name: target, dtype: int64

 ${\tt tar.plot(kind='pie',explode=(0.02,0.02,0.02,0.02),autopct='\%1.1f\%')}$ 

<Axes: ylabel='target'>



plt.figure(figsize=(15,10))
sns.countplot(x='location',data=df,hue='target')
plt.xticks(rotation=90)

```
(array([ 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31]),
 [Text(0, 0, 'Facebook'),
  Text(1, 0, 'Amazon'),
Text(2, 0, 'Microsoft'),
  Text(3, 0, 'CS-GO'),
  Text(4, 0, 'Google'),
  Text(5, 0, 'FIFA'),
Text(6, 0, 'MaddenNFL'),
   Text(7, 0, 'TomClancysRainbowSix'),
  Text(8, 0, 'AssassinsCreed'),
Text(9, 0, 'CallOfDuty'),
  Text(10, 0, 'Dota2'),
  Text(11, 0, 'Verizon'),
  Text(12, 0,
                     'NBA2K'),
                     'Nvidia'),
  Text(13, 0,
  Text(14, 0, 'GrandTheftAuto(GTA)'),
  Text(15, 0,
                     'RedDeadRedemption(RDR)'),
   Text(16, 0,
                     'Hearthstone'),
  Text(17, 0, 'ApexLegends'),
  Text(18, 0, 'Overwatch'),
  Text(19, 0, 'PlayerUnknownsBattlegrounds(PUBG)'),
  Text(20, 0,
                     'Borderlands'),
  Text(21, 0, 'PlayStation5(PS5)'),
  Text(22, 0, 'johnson&johnson'),
   Text(23, 0, 'Fortnite'),
  Text(24, 0, 'Xbox(Xseries)'),
   Text(25, 0, 'CallOfDutyBlackopsColdWar'),
  Text(26, 0, 'HomeDepot'),
   Text(27, 0, 'Cyberpunk2077'),
  Text(28, 0, 'TomClancysGhostRecon'),
  Text(29, 0, 'WorldOfCraft'),
  Text(30, 0, 'LeagueOfLegends'),
Text(31, 0, 'Battlefield')])
   25
                                                                                                                                                 target
                                                                                                                                                Irrelevant
                                                                                                                                                 Neutral
                                                                                                                                                 Negative
                                                                                                                                                Positive
   20
   15
   10
                      CS-60
                           Google
                                                       Dota2
                                                           Verizon
                                                                NBA2K
                                                                    Nvidia
                                                                                                                   Fortnite
                               FIFA
                                        TomClancysRainbowSix
                                             AssassinsCreed
                                                  CallofDuty
                                                                         GrandTheftAuto(GTA)
                                                                                                          PlayStation5(PS5)
                                                                                                                                      Cyberpunk2077
                                                                                                                                          TomClancysGhostRecon
                                                                                                                                               WorldOfCraft
                                    MaddenNFL
                                                                              RedDeadRedemption(RDR)
                                                                                       ApexLegends
                                                                                            Overwatch
                                                                                                 PlayerUnknownsBattlegrounds(PUBG)
                                                                                                     Borderlands
                                                                                                              johnson&johnson
                                                                                                                        Xbox(Xseries)
                                                                                                                            CallOfDutyBlackopsColdWar
                                                                                                                                 HomeDepot
                                                                                                                                                    LeagueOfLegends
                                                                                                                                                        Battlefield
                                                                              location
```

```
text=df['text'].value_counts()
text
```

Wow 2

I mentioned on Facebook that I was struggling for motivation to go for a run the other day, which has been translated by Tomâres great auntie as ârehayley canâret get out of bedâre and told to his grandma, who now thinks Iârem a lazy, terrible person ðex£

Update: I actually finished Odyssey. I wish the three-ponged questline structure was communicated a bit more clearly but that game is a lot of fun once it hits its stride. twitter.com/iv\_javy/statusâ@; 1

Know Your History...unless it involves China...because like....we have investors and shit...

1

.. eSports: Curiosities you probably didnâT@t know about Red Dead Redemption 2 goalsn.com/2020/02/esportâT https://t.co/JMLBSxAUou

Nvidia GeForce Now: Google Stadia rival suffers another blow news89.net/nvidia-geforce⮦ https://t.co/9ijNjV7L4C

I SAW A TWEET ABOUT HOW IT'S A STORY TOLD OVER AND OVER CUZ THEY'RE MAD MEN DIED IN THE WAR AND IT DIDN'T HAVE BATTLEFIELD 1
TREATMENT, I CAN'T EVEN THINK OF ANOTHER WW1 MOVIE PPL WOULD EVEN KNOW

Bro I had an abusive internet boyfriend that used to call me a stupid bitch whenever niggas used to like my pics on facebook and I used to beg him to not be mad at me. I was such a stupid bitch FOR REAL

Johnson & Johnson to stop selling talc baby powder in U.S. and Canada j.mp/3e1YtDV (Reuters) <a href="https://t.co/dsaUTgb5p9">https://t.co/dsaUTgb5p9</a>

Name: text, Length: 999, dtype: int64

df=df.drop(df.index[(df['target']=='Irrelevant')],axis=0)
df

	ID	location	target	text	7
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	
5	6273	FIFA	Negative	Hi @EAHelp Iâ□□ve had Madeleine McCann in my c	
993	314	Amazon	Negative	Please explain how this is possible! How can t	
994	9701	PlayStation5(PS5)	Positive	Good on Sony. As much as I want to see the new	
997	2652	Borderlands	Positive	Today sucked so itâ□□s 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	
000 =	1	aalumna			

828 rows × 4 columns

# reset index values
df.reset\_index(drop=True,inplace=True)
df

	ID	location	target	text
0	352	Amazon	Neutral	BBC News - Amazon boss Jeff Bezos rejects clai
1 8	8312	Microsoft	Negative	@Microsoft Why do I pay for WORD when it funct
2 4	4371	CS-GO	Negative	CSGO matchmaking is so full of closet hacking,
3 4	4433	Google	Neutral	Now the President is slapping Americans in the
4 6	6273	FIFA	Negative	Hi @EAHelp lâ□□ve had Madeleine McCann in my c
23	314	Amazon	Negative	Please explain how this is possible! How can t
<b>24</b> 9	9701	PlayStation5(PS5)	Positive	Good on Sony. As much as I want to see the new

text

df=df. df

target

```
BBC News - Amazon boss Jeff Bezos rejects clai...
       0
             Neutral
       1
           Negative
                         @Microsoft Why do I pay for WORD when it funct...
       2
           Negative
                          CSGO matchmaking is so full of closet hacking,...
       3
             Neutral
                           Now the President is slapping Americans in the...
           Negative Hi @EAHelp Iâ□□ve had Madeleine McCann in my c...
       4
       ...
      823
           Negative
                            Please explain how this is possible! How can t...
            Positive
                         Good on Sony. As much as I want to see the new...
      824
                           Today sucked so itâ□□s time to drink wine n pl...
      825
            Positive
      826
            Positive
                           Bought a fraction of Microsoft today. Small wins.
      827
             Neutral
                           Johnson & Johnson to stop selling talc baby po...
     828 rows × 2 columns
# df['target']=df['target'].str.replace('Neutral','0')
# df['target']=df['target'].str.replace('Negative','-1')
# df['target']=df['target'].str.replace('Positive','1')
# df['target']=df['target'].astype(int)
# using map function
df['target']=df['target'].map({'Positive':1, 'Negative':-1, 'Neutral':0})
df.dtypes
     target
                 int64
                object
     dtype: object
nltk.download('wordnet')
nltk.download('stopwords')
nltk.download('punkt')
nltk.download('omw-1.4')
     [nltk_data] Downloading package wordnet to /root/nltk_data...
     [nltk_data]
                    Package wordnet is already up-to-date!
     [nltk_data] Downloading package stopwords to /root/nltk_data...
                    Package stopwords is already up-to-date!
     [nltk_data]
     [nltk_data] Downloading package punkt to /root/nltk_data...
                    Package punkt is already up-to-date!
     [nltk_data]
     [nltk_data] Downloading package omw-1.4 to /root/nltk_data...
     [nltk_data]
                   Package omw-1.4 is already up-to-date!
     True
```

```
0
            BBC News - Amazon boss Jeff Bezos rejects clai...
            @Microsoft Why do I pay for WORD when it funct...
     1
            CSGO matchmaking is so full of closet hacking,...
     3
            Now the President is slapping Americans in the...
     4
           Hi @EAHelp Iâllve had Madeleine McCann in my c...
           Please explain how this is possible! How can \mathsf{t}\dots
    823
     824
            Good on Sony. As much as I want to see the new...
    825
            Today sucked so itâlls time to drink wine n pl...
            Bought a fraction of Microsoft today. Small wins.
     826
            Johnson & Johnson to stop selling talc baby po...
     827
     Name: text, Length: 828, dtype: object
# preprocessing
from nltk.tokenize import TweetTokenizer
tk=TweetTokenizer()
tweets=tweets.apply(lambda x:tk.tokenize(x)).apply(lambda x:" ".join(x))
tweets
    0
            BBC News - Amazon boss Jeff Bezos rejects clai...
            @Microsoft Why do I pay for WORD when it funct...
     1
     2
            CSGO matchmaking is so full of closet hacking ...
     3
           Now the President is slapping Americans in the...
     4
           Hi @EAHelp Iâ ☑ ☑ ve had Madeleine McCann in m...
           Please explain how this is possible ! How can \dots
     823
            Good on Sony . As much as I want to see the ne...
     824
            Today sucked so itâ 🛚 🗗 s time to drink wine n...
     825
            Bought a fraction of Microsoft today . Small w...
     826
     827
            Johnson & Johnson to stop selling talc baby po...
    Name: text, Length: 828, dtype: object
# special charachter remove
# re : regular expression we remove special charachters
tweets=tweets.str.replace('[^a-zA-Z-0-9]+',' ')
tweets
     <ipython-input-124-d387f0562dfc>:4: FutureWarning: The default value of regex will change from True to False in a future version.
      tweets=tweets.str.replace('[^a-zA-Z-0-9]+',' ')
     a
            BBC News - Amazon boss Jeff Bezos rejects clai...
            Microsoft Why do I pay for WORD when it funct...
    1
    2
            CSGO matchmaking is so full of closet hacking ...
     3
           Now the President is slapping Americans in the...
     4
           Hi EAHelp I ve had Madeleine McCann in my cell...
            Please explain how this is possible How can th...
     823
            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
            Johnson Johnson to stop selling talc baby powd...
     827
    Name: text, Length: 828, dtype: object
# example for removal of special charachters
str1='wonderful @@@@peacock!12345#@!'
str2=re.sub('[^a-zA-Z-0-9]+',' ',str1)
                                                   # sub = replace
                                                   # '[^a-zA-Z-0-9]+' here '^' denotes 'not in the string'
str2
                                                                         here '+' denotes 'replace the combination of symbols
     'wonderful peacock 12345 '
# stemming or lematization
from nltk.stem import SnowballStemmer
ss=SnowballStemmer('english')
tweets=tweets.apply(lambda x:[ss.stem(i.lower()) for i in tk.tokenize(x)]).apply(lambda x:" ".join(x))
tweets
    0
            bbc news - amazon boss jeff bezo reject claim ...
            microsoft whi do i pay for word when it functi...
     1
    2
            csgo matchmak is so full of closet hack it s a...
            now the presid is slap american in the face th...
     3
     4
            hi eahelp i ve had madelein mccann in my cella...
     823
            pleas explain how this is possibl how can they...
            good on soni as much as i want to see the new ...
     824
```

```
825
            today suck so it s time to drink wine n play b...
     826
               bought a fraction of microsoft today small win
     827
            johnson johnson to stop sell talc babi powder ...
    Name: text, Length: 828, dtype: object
# stopwords
nltk.download('stopwords')
from nltk.corpus import stopwords
sw=stopwords.words('english')
tweets=tweets.apply(lambda \ x:[i \ for \ i \ in \ tk.tokenize(x) \ if \ i \ not \ in \ sw]).apply(lambda \ x:' \ '.join(x))
tweets
     [nltk_data] Downloading package stopwords to /root/nltk_data...
     [nltk_data] Package stopwords is already up-to-date!
            bbc news - amazon boss jeff bezo reject claim ...
            microsoft whi pay word function poor samsungus...
                 csgo matchmak full closet hack truli aw game
     2
     3
            presid slap american face realli commit unlaw ...
     4
            hi eahelp madelein mccann cellar past 13 year ...
     823
            pleas explain possibl let compani overcharg sc...
            good soni much want see new ps5 go right much \dots
     824
     825
            today suck time drink wine n play borderland s...
     826
                    bought fraction microsoft today small win
     827
            johnson johnson stop sell talc babi powder u c...
    Name: text, Length: 828, dtype: object
# vectorization
from sklearn.feature_extraction.text import TfidfVectorizer
vec=TfidfVectorizer()
train_data=vec.fit_transform(tweets)
print(train_data)
                     0.2456282916977699
       (0, 739)
       (0, 524)
                     0.2456282916977699
       (0, 3629)
                     0.19717097284763976
       (0, 883)
                     0.11802111433875656
                     0.2456282916977699
       (0, 1063)
                     0.22139963227270484
       (0, 1193)
       (0, 2103)
                     0.12503547156807654
       (0, 342)
                     0.2135997461023189
       (0, 913)
                     0.16343698819954885
       (0, 857)
                     0.2018385048310212
       (0, 2902)
                     0.2456282916977699
                     0.2456282916977699
       (0, 612)
       (0, 1930)
                     0.23145543449136263
       (0, 679)
                     0.2135997461023189
       (0, 412)
                     0.14270670271218175
       (0, 2404)
                     0.365996231282465
       (0, 581)
                     0.46291086898272527
       (1, 847)
                     0.4055823664694651
       (1, 3039)
                     0.4055823664694651
       (1, 2693)
                     0.3821800909185634
       (1, 1479)
                     0.4055823664694651
       (1, 3849)
                     0.36557591217188057
       (1, 2595)
                     0.3126902562590763
       (1, 3808)
                     0.26216072802580975
       (1, 2266)
                     0.24555654927912696
       (825, 677)
                     0.22981061112100945
       (825, 1056)
                     0.2315686698425631
       (825, 3532)
                     0.21750175079084832
                     0.2904718522758868
       (825, 3380)
       (825, 2661)
                     0.17148706662740873
       (826, 1453)
                     0.5079831062080814
       (826, 3222)
                     0.47867226429410115
       (826, 682)
                     0.4174215841659411
       (826, 3819)
                     0.353278941165688
       (826, 3542)
                     0.34523850330234374
       (826, 2266)
                     0.3075542453642147
       (827, 222)
(827, 1195)
                     0.3085327836746446
                     0.3085327836746446
```

(827, 2950)

(827, 166)

(827, 2341)

(827, 760)

(827, 3096)

(827, 3428)

0.3085327836746446

0.3085327836746446

0.3085327836746446

0.2602967814546125

0.24766572568673034

0.24766572568673034

```
(827, 1776)
                  0.14999073362910845
      (827, 2711)
                  0.23368360020019152
      (827, 547)
                   0.22309506625787626
      (827, 1952)
                  0.37089519596031884
      (827, 883)
                   0.14824588278342318
train_data.shape
    (828, 3961)
y=df['target'].values
    0,
            0, 1, 0, 0, 0, 1, 0, -1, -1, -1, 0, 1, -1, -1, 1,
                         1,
                             1, -1, 0, -1, 0, -1, 1, -1, -1, 1,
               1, -1, -1,
                                                                     1.
                  0, 1,
                          1,
                             0, 1,
                                     0, -1, -1, 0,
                                                   0, -1, 1, -1, -1,
               1, 0, -1, 1, 1, 0, 1, 0, 1, -1,
                                                   0, 0, 0, -1, 0, -1,
               0,
            0,
                  1, 1, 0, -1, -1, 1, -1, 0, -1, 1, 0, -1, 0, 1,
               1,
                          0, 0, 1, 0, 1, 1, -1,
                   0, 0,
                                                              0. -1.
                          0, -1, -1, -1, 1, 1, 0, 0, 1, 0, 0,
           1, -1, 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,
                                     0, -1, 0, 0, 1, -1, 1,
               1, -1,
                      1,
                          1, -1, 1,
           -1, -1, 1, 1, 1, 0, 0, 1, -1, 0, 1, 0, -1,
                                                              0, 0, -1,
           1, 1, -1, 0, 1, 0, -1, 0, -1, 1, 1, -1, -1, -1,
                                                              1, -1,
                  0, -1, 1, -1, 1, -1, 0, 0, 1, -1, 0, -1,
               0,
                                                              1, -1,
                                                                      1,
                          1, -1, -1, 1, -1, 0, 0, 0, 1, 0, 1, -1,
                   1, 1,
                                                                      0,
                   0, -1, 1, -1, -1, 1, 0, 0, -1, -1, -1,
                                                              0,
           -1, 1, 0, -1, -1, -1, 1, 0, 0, -1, 1, 1, 0, 1,
                                                                      1.
            1, -1,
                   0, 1, -1,
                              0, -1, -1, 1, 1,
                                                1,
                                                   1,
                                                       0, -1,
                                                              0,
           1, -1, -1, -1, 1,
                             0, 1, -1, 0, -1, 1, 1, 1, 1,
                                                                      0,
           -1, 1, 1, 0, -1,
                             1, 0, -1, -1, -1, -1, -1,
                                                       0, 0, 0, 1, 1,
           -1, -1, 0, -1, 0, 0, -1, 1, -1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, -1, -1, 1, 1,
           -1, -1, 0, -1, 0,
                                                           1, 0, 0, -1,
                                                          0, -1, -1,
                                                                     1,
           1, -1, 1, 1, 1, 1, 0, -1, 1, 0, 0, 1, 1, 1, 1,
           -1, -1, -1, -1, 0, 1, -1, -1, 1, 1, 0, 0, -1, -1, 1, 0, -1,
           -1, -1, 0, 0,
                          1, -1, -1, -1,
                                        0, 0, 0, -1, -1,
                                                           1, -1, 0, -1,
           0, \quad 1, \ -1, \quad 0, \quad 1, \quad 1, \ -1, \quad 0, \quad 0, \quad 1, \ -1, \quad -1, \quad 0, \quad 0, \ -1, \quad 1, \ -1,
           0, -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, 1, 0, 0, 0, 0, 0, 1,
                                                         0,
                                                              0, 0, -1,
                                                              0, 1,
           -1,
               0, -1, 1,
                          0, 0, -1, 1, 0, 0, -1, 0, -1,
           -1, -1, 0, 0, 0,
                             0, 1, 1, 1, -1, -1, 0, 1, 0, 0, -1,
                                                                     1,
           1, 0,
                  1, -1, -1,
                             0, 1, -1, 1, -1, 0, 1,
                                                                      1,
                             1, -1, 0, 1, 1, 1, 0, -1, 0, 1,
                                                                      1,
           0, -1, 0, 0, -1,
           1, 1, 1, -1, 0, 1, 0, 0, -1, -1, -1, 0, 1, 0, -1, 1,
           1, 0, 1, -1, 0, -1, 0, -1, 0, 0, 1, -1, 1, 1, 0, -1,
           -1, -1, -1, -1, 1, 1,
                                 1, 1,
                                        0, -1, -1, 1, -1, -1,
                                                             0,
            0, -1, 0, 1, -1, 0, 1, -1, 0, 0, 1, -1, 0, -1, 1,
           1, 0, 1, -1, 0, 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,
                                                                      0,
           -1, 0, 1, 1, -1, -1, -1, 1, -1,
                                            0, -1, 0, 0, 1,
                                                              1, -1,
                                                                      0,
           1, -1, -1, -1, -1, -1, -1, -1,
                                            0, -1, 0, 0, 0, 1, 0,
           0, -1, 0, 1, 0, -1, -1, 1, 0,
                                            1, 0, 1, 0, -1, 1, 1,
                                                                      1.
              -1, -1,
                          0, 0, 0, 0, 0,
                                            0, -1, -1, -1, -1,
                      1,
               0, -1, 1,
                          1, -1, 1, 0, 0, 1, -1, 0, -1, 0, 1, 1,
               1, -1, -1,
                          0, -1, 0, -1, 1, 0, -1, -1, 1, 1, -1, 0, -1,
               0,
                  0,
                      0, 0, 0, 1, 0, 1, 1, 1, -1, 0, 1, 0, 1, 0,
                      0, -1, -1, 1, 1, 1, 1,
                                                          1, -1, -1, -1,
            1,
               0, 1,
                                                0, -1, 1,
            0, 1, 0, 1, 1, 0, 1, -1, 1, 1, 1, 0])
# train test split
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)
print(x_train)
      (0, 2833)
                   0.22445333746210416
      (0, 721)
                   0.23819743082652528
      (0, 0)
                   0.23819743082652528
      (0, 1717)
                   0.22445333746210416
      (0, 3411)
                   0.23819743082652528
      (0, 3470)
                   0.23819743082652528
      (0, 1447)
                   0.22445333746210416
      (0, 2063)
                   0.23819743082652528
      (0, 335)
                   0.23819743082652528
      (0, 2582)
                   0.19573239287831176
      (0, 1715)
                   0.22445333746210416
```

(827, 3347)

(0, 409)

0.1774619718703711

0.21723219669277322

```
(0, 3554)
                      0.1804114056568988
       (0, 1294)
                      0.14874102728657873
       (0, 2140)
                      0.284319400937232
       (0, 3777)
                      0.16565538326448254
       (0, 59)
                      0.16771038243892572
       (0, 3049)
                      0.15258254163782356
       (0, 1313)
                      0.14997378514152887
       (0, 3605)
                      0.15849261671802412
       (0, 2834)
                      0.16771038243892572
       (0, 3609)
                      0.15258254163782356
       (0, 3171)
                      0.19573239287831176
                      0.19120606523479225
       (0, 3629)
       (1, 307)
                      0.23042284485480724
       (577, 742)
                      0.14968891974696927
       (577, 1947)
(577, 3767)
                      0.16302270346202263
                      0.18158863086173657
       (577, 2317)
                      0.18717313645751385
       (577, 1165)
                      0.18717313645751385
       (577, 3396)
(577, 953)
                      0.16302270346202263
                      0.20281971497818488
       (577, 1205)
                      0.15790584022871224
       (577, 1595)
(577, 2459)
                      0.24851073610897026
                      0.2625744571143809
       (577, 752)
                      0.14043006275740014
       (577, 2401)
                      0.12126470370721307
       (577, 771)
(577, 1502)
                      0.16302270346202263
                      0.08667889229898602
       (577, 3808)
                      0.2782529252576799
       (578, 2968)
                      0.2618429705732027
       (578, 3138)
                      0.37189303347534525
       (578, 1757)
                      0.37189303347534525
       (578, 1639)
                      0.2618429705732027
       (578, 3576)
                      0.37189303347534525
       (578, 2264)
                      0.37189303347534525
       (578, 1994)
                      0.28167283203049437
       (578, 2613)
                      0.23614849222634068
       (578, 2738)
                      0.2985263248739169
       (578, 1256)
                     0.2922929058414592
# model creation
from sklearn.metrics import confusion_matrix,classification_report
from sklearn.svm import SVC
from sklearn.naive_bayes import MultinomialNB
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.tree import DecisionTreeClassifier
s_model=SVC()
n_model=MultinomialNB()
k_model=KNeighborsClassifier()
r_model=RandomForestClassifier()
d_model=DecisionTreeClassifier()
lst_model=[s_model,n_model,k_model,r_model,d_model]
for i in lst_model:
  i.fit(x_train,y_train)
  y_pred=i.predict(x_test)
  print(i)
  print('*'*100)
  print(confusion_matrix(y_test,y_pred))
```

print(classification\_report(y\_test,y\_pred))

\_

[45 1/ 29]]				
	precision	recall	f1-score	support
-1	0.41	0.81	0.55	79
0	0.47	0.30	0.37	79
1	0.67	0.32	0.43	91
accuracy			0.47	249
macro avg	0.52	0.48	0.45	249
weighted avg	0.53	0.47	0.45	249
RandomForest		to also also also also also also also als		
*********	*********	******	*****	*****
[[56 16 7]				
[28 41 10] [25 26 40]]				
[25 26 40]]	precision	nocall	f1-score	support
	precision	recall	11-3001-6	Support
-1	0.51	0.71	0.60	79
0	0.49	0.52	0.51	79
1	0.70	0.44	0.54	91
accuracy			0.55	249
macro avg	0.57	0.56	0.55	249
weighted avg	0.58	0.55	0.55	249
DecisionTree(				
********	********	******	******	******
[[49 22 8]				
[22 37 20]				
[31 28 32]]	precision	nocall	f1-score	cuppont
	bijectztoli	recall	11-2001.6	support
-1	0.48	0.62	0.54	79
0	0.43	0.47	0.45	79
1	0.53	0.35	0.42	91
-	0.33	2.33		
accuracy			0.47	249
macro avg	0.48	0.48	0.47	249
weighted avg	0.48	0.47	0.47	249
0 0				