```
+ Code + Text
```

```
import re  #regular expression
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import nltk
from nltk.stem import WordNetLemmatizer

df=pd.read_csv('/content/twitter_validation.csv')
df.columns=['id','location','target','text']
df
```

	id	location	target	text
0	352	Amazon	Neutral	BBC News - Amazon boss Jeff Bezos rejects clai
1	8312	Microsoft	Negative	@Microsoft Why do I pay for WORD when it funct
2	4371	CS-GO	Negative	CSGO matchmaking is so full of closet hacking,
3	4433	Google	Neutral	Now the President is slapping Americans in the
4	6273	FIFA	Negative	Hi @EAHelp I've had Madeleine McCann in my cel
994	4891	GrandTheftAuto(GTA)	Irrelevant	☆ Toronto is the arts and culture capital of
995	4359	CS-GO	Irrelevant	this is actually a good move tot bring more vi
996	2652	Borderlands	Positive	Today sucked so it's time to drink wine n play
997	8069	Microsoft	Positive	Bought a fraction of Microsoft today. Small wins.
998	6960	johnson&johnson	Neutral	Johnson & Johnson to stop selling talc baby po
000				

999 rows × 4 columns

df.head()

text	target	location	id	
BBC News - Amazon boss Jeff Bezos rejects clai	Neutral	Amazon	352	0
@Microsoft Why do I pay for WORD when it funct	Negative	Microsoft	8312	1
CSGO matchmaking is so full of closet hacking,	Negative	CS-GO	4371	2
Now the President is slapping Americans in the	Neutral	Google	4433	3
Hi @EAHelp I've had Madeleine McCann in my cel	Negative	FIFA	6273	4

```
print('lenth\ of\ data\ is',len(df))
```

lenth of data is 999

df.isna().sum()

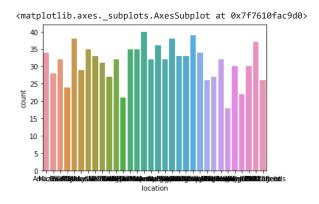
id 0 location 0 target 0 text 0 dtype: int64

df['location'].value_counts()

```
RedDeadRedemption(RDR)
                                       40
johnson&johnson
                                       39
FIFA
                                       38
PlayerUnknownsBattlegrounds(PUBG)
                                       38
LeagueOfLegends
ApexLegends
Nvidia
                                       36
                                       35
{\tt TomClancysRainbowSix}
                                       35
GrandTheftAuto(GTA)
                                       35
Fortnite
                                       34
Amazon
                                       34
AssassinsCreed
                                       33
Borderlands
```

PlayStation5(PS5)	33
Hearthstone	32
Overwatch	32
Verizon	32
CS-G0	32
Facebook	32
CallOfDuty	31
Cyberpunk2077	30
WorldOfCraft	30
MaddenNFL	29
Microsoft	28
Dota2	27
CallOfDutyBlackopsColdWar	27
Battlefield	26
Xbox(Xseries)	26
Google	24
TomClancysGhostRecon	22
NBA2K	21
HomeDepot	18
Name: location, dtype: int64	

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

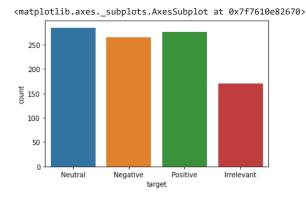


df['target'].value_counts()

Neutral 285 Positive 277 Negative 266 Irrelevant 171

Name: target, dtype: int64

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



#remove irrelevent from targrt
df.drop(df.index[(df['target']=='Irrelevant')],axis=0,inplace=True)
df

	id	location	target	text
0	352	Amazon	Neutral	BBC News - Amazon boss Jeff Bezos rejects clai
1	8312 Microsoft Negative @Microsoft Why do I pay for WORD when it fur		@Microsoft Why do I pay for WORD when it funct	
2	2 4371 CS-GO Negative CSGO matchmaking is so full of closet hacking		CSGO matchmaking is so full of closet hacking,	
3	4433	Google	Neutral	Now the President is slapping Americans in the
4	6273	FIFA	Negative	Hi @EAHelp I've had Madeleine McCann in my cel
992	314	Amazon	Negative	Please explain how this is possible! How can t
993	9701	PlayStation5(PS5)	Positive	Good on Sony. As much as I want to see the new
et indexvalue				

#reset indexvalue
df.reset_index(drop=True,inplace=True)
#f

	id	location	target	text
0	352	Amazon	Neutral	BBC News - Amazon boss Jeff Bezos rejects clai
1	8312	Microsoft	Negative	@Microsoft Why do I pay for WORD when it funct
2	4371	CS-GO	Negative	CSGO matchmaking is so full of closet hacking,
3	4433	Google	Neutral	Now the President is slapping Americans in the
4	6273	FIFA	Negative	Hi @EAHelp I've had Madeleine McCann in my cel
823	314	Amazon	Negative	Please explain how this is possible! How can t
824	9701	PlayStation5(PS5)	Positive	Good on Sony. As much as I want to see the new
825	2652	Borderlands	Positive	Today sucked so it's time to drink wine n play
826	8069	Microsoft	Positive	Bought a fraction of Microsoft today. Small wins.
827	6960	johnson&johnson	Neutral	Johnson & Johnson to stop selling talc baby po

828 rows × 4 columns

 $\label{def:drop} $$ $$ df.drop(['id','location'],axis=1,inplace=True) $$ df $$$

	target	text
0	Neutral	BBC News - Amazon boss Jeff Bezos rejects clai
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
4	Negative	Hi @EAHelp I've had Madeleine McCann in my cel
823	Negative	Please explain how this is possible! How can t
824	Positive	Good on Sony. As much as I want to see the new
825	Positive	Today sucked so it's time to drink wine n play
826	Positive	Bought a fraction of Microsoft today. Small wins.
827	Neutral	Johnson & Johnson to stop selling talc baby po

828 rows × 2 columns

```
#changing values of positive negative and neutral
df['target']=df['target'].map({'Positive':1,'Negative':-1,'Neutral':0})
```

```
target
                                                           text
       0
                0
                    BBC News - Amazon boss Jeff Bezos rejects clai...
                    @Microsoft Why do I pay for WORD when it funct...
       1
               -1
               -1
                     CSGO matchmaking is so full of closet hacking,...
       3
                0
                      Now the President is slapping Americans in the...
               -1 Hi @EAHelp I've had Madeleine McCann in my cel...
       ...
               -1
                       Please explain how this is possible! How can t...
      823
      824
                1
                    Good on Sony. As much as I want to see the new...
      825
                        Today sucked so it's time to drink wine n play...
                           nltk.download('wordnet')
nltk.download('stopwords')
nltk.download('punkt')
     [nltk_data] Downloading package wordnet to /root/nltk_data...
     [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.
     True
tweets=df.text
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...
            Hi @EAHelp I've had Madeleine McCann in my cel...
     4
     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 it's time to drink wine n play...
     826
            Bought a fraction of Microsoft today. Small wins.
            Johnson & Johnson to stop selling talc baby po...
     Name: text, Length: 828, dtype: object
#tokenization
from nltk 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 \dots
            Now the President is slapping Americans in the...
     3
     4
            Hi @EAHelp I ' ve had Madeleine McCann in my c...
     823
            Please explain how this is possible ! How can ...
     824
            Good on Sony . As much as I want to see the ne...
            Today sucked so it 's time to drink wine n pl...
     825
            Bought a fraction of Microsoft today . Small w...
     826
            Johnson & Johnson to stop selling talc baby po...
     827
     Name: text, Length: 828, dtype: object
#removing special charecters example
str1='Worldof anime/?[123>:'
str2=re.sub('[^a-zA-Z0-9]'," ",str1)
str2
     'Worldof anime 123 '
#removing special charecters
tweets.str.replace('[^a-zA-Z0-9]+','')
     <ipython-input-20-4cef94aace1b>:2: FutureWarning: The default value of regex will change from True to False in a future version.
       tweets.str.replace('[^a-zA-Z0-9]+','')
            BBCNewsAmazonbossJeffBezosrejectsclaimscompany...
```

```
MicrosoftWhydoIpayforWORDwhenitfunctionssopoor...
     2
            CSGOmatchmakingissofullofclosethackingitsatrul...
     3
            NowthePresidentisslappingAmericansinthefacetha...
     4
            HiEAHelpIvehadMadeleineMcCanninmycellarforthep...
            PleaseexplainhowthisispossibleHowcantheyletcom...
     823
     824
            GoodonSonyAsmuchasIwanttoseethenewPS5whatsgoin...
     825
            {\tt Today sucked so its time tod rink winenplay border land...}
     826
                     BoughtafractionofMicrosofttodaySmallwins
     827
            {\tt JohnsonJohnsontostopsellingtalcbabypowderinUSa...}
     Name: text, Length: 828, dtype: object
#lenth avobe 3 tokens only(for a meaning full word)
from nltk.tokenize import word_tokenize
tweets=tweets.apply(lambda x:' '.join([w for w in word_tokenize(x) if len(w)>=3]))
tweets
            BBC News Amazon boss Jeff Bezos rejects claims...
     a
     1
            Microsoft Why pay for WORD when functions poor...
     2
            CSGO matchmaking full closet hacking truly awf...
            Now the President slapping Americans the face ...
     3
     4
            EAHelp had Madeleine McCann cellar for the pas...
     823
            Please explain how this possible How can they ...
            Good Sony much want see the new PS5 what going...
     824
     825
            Today sucked time drink wine play borderlands ...
     826
                   Bought fraction Microsoft today Small wins
     827
            Johnson Johnson stop selling talc baby powder ...
     Name: text, Length: 828, dtype: object
from nltk.stem import SnowballStemmer
stemmer=SnowballStemmer('english')
tweets=tweets.apply(lambda \ x:[stemmer.stem(i.lower()) \ for \ i \ in \ tk.tokenize(x)]).apply(lambda \ x:' \ '.join(x))
tweets
            bbc news amazon boss jeff bezo reject claim co...
            microsoft whi pay for word when function poor ...
     1
                 csgo matchmak full closet hack truli aw game
     2
     3
            now the presid slap american the face that rea...
            eahelp had madelein mccann cellar for the past...
                                   . . .
     823
            plea explain how this possibl how can they let...
            good soni much want see the new ps5 what go ri...
     824
     825
            today suck time drink wine play borderland unt...
     826
                    bought fraction microsoft today small win
            johnson johnson stop sell talc babi powder and...
     Name: text, Length: 828, dtype: object
from nltk.corpus import stopwords
stop=stopwords.words('english')
tweets=tweets.apply(lambda x:[i \text{ for } i \text{ in } tk.tokenize(x) \text{ if } i \text{ not in } stop]).apply(lambda <math>x:"".join(x))
tweets
     0
            bbc news amazon boss jeff bezo reject claim co...
            microsoft whi pay word function poor samsungus...
     1
                 csgo matchmak full closet hack truli aw game
     2
     3
            presid slap american face realli commit unlaw ...
     4
            eahelp madelein mccann cellar past year littl ...
            plea explain possibl let compani overcharg sca...
     823
     824
            good soni much want see new ps5 go right much ...
            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
#vectorization
from sklearn.feature_extraction.text import TfidfVectorizer
vec=TfidfVectorizer()
train data=vec.fit transform(tweets)
print(train_data)
       (0, 629)
                     0.24330548895023374
       (0, 416)
                     0.24330548895023374
                     0.21930594969814493
       (0, 3496)
       (0, 774)
                     0.1169050386344535
```

У

```
0.24330548895023374
      (0, 951)
      (0, 1080)
                   0.22926665845317631
      (0, 1973)
                   0.123853064057564
      (0, 232)
                   0.21157982374854553
      (0, 803)
                   0.16189143380671017
      (0, 748)
                   0.19992980355585638
      (0, 2767)
                   0.24330548895023374
      (0, 507)
                   0.24330548895023374
      (0, 1809)
                   0.22926665845317631
      (0, 571)
                   0.21157982374854553
                   0.14135718585131646
      (0, 300)
      (0, 2271)
                   0.3685605265618393
      (0, 476)
                   0.45853331690635263
      (1, 738)
                   0.4055823664694651
      (1, 2898)
                   0.4055823664694651
      (1, 2556)
                   0.3821800909185634
      (1, 1364)
                   0.4055823664694651
      (1, 3715)
                   0.36557591217188057
      (1, 2457)
                   0.3126902562590763
      (1, 3673)
                   0.26216072802580975
      (1, 2134)
                   0.24555654927912696
      (825, 3410)
                   0.2560582225152134
      (825, 569)
                   0.22981061112100945
      (825, 944)
                   0.2315686698425631
      (825, 3399)
                   0.21750175079084832
                   0.2904718522758868
      (825, 3244)
      (825, 2523)
                   0.17148706662740873
      (826, 1338)
                   0.5079831062080814
      (826, 3083)
                   0.47867226429410115
      (826, 574)
                   0.4174215841659411
      (826, 3684)
                   0.353278941165688
      (826, 3410)
                   0.34523850330234374
      (826, 2134)
                   0.3075542453642147
      (827, 1083)
                   0.3243569563278974
      (827, 2815)
                   0.3243569563278974
      (827, 130)
                   0.3243569563278974
      (827, 2209)
                   0.3243569563278974
      (827, 653)
                   0.27364700363122035
      (827, 2956)
                   0.260368120410828
      (827, 3293)
                   0.260368120410828
      (827, 3209)
                   0.22837370245229333
      (827, 1658)
                   0.15768352801246954
      (827, 2575)
                   0.2456688731807802
      (827, 442)
                   0.23453726959363683
      (827, 1831)
                   0.38991784096820636
                   0.15584918644651963
      (827, 774)
train_data.shape
    (828, 3824)
y=df['target'].values
    \mathsf{array}([\ 0,\ -1,\ -1,\ \ 0,\ -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, 1, 0, 0, 0, 1, 0, -1, -1, -1, 0, 1, -1, -1, 1,
            1, 1, -1, -1, 1, 1, -1, 0, -1, 0, -1, 1, -1, -1, 1, 1,
            0, 0, 0, 1, 1, 0, 1, 0, -1, -1, 0, 0, -1, 1, -1, -1, -1,
            0, 1, 0, -1, 1, 1, 0, 1, 0, 1, -1,
                                                         0, 0, -1,
               0, 1, 1,
                          0, -1, -1, 1, -1,
                                             0, -1,
            1, 1, 0, 0, 0, 0, 1, 0, 1, 1, -1,
                                                     0, 0, 0, 0, -1,
            1, -1, 0, -1, 0, -1, -1, -1, 1, 1, 1,
                                                     0, 0, 1, 0, 0,
              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, 0, -1, 0, 0, 1, -1, 1,
            1,
                                                                       1,
                                                                    0,
           -1, -1, 1, 1, 1, 0, 0, 1, -1, 0, 1, 0, -1,
           1, 1, -1, 0, 1, 0, -1, 0, -1, 1, 1, -1, -1, -1, -1,
              0, 0, -1, 1, -1, 1, -1, 0, 0, 1, -1, 0, -1,
                                                               1, -1,
            1,
               1, 1, 1, -1, -1, 1, -1,
                                             0, 0, 0, 1, 0,
                   0, -1, 1, -1, -1, 1, 0, 0, -1, -1, -1,
                   0, -1, -1, -1, 1, 0, 0, -1, 1, 1,
           -1, 1,
            1, -1, 0, 1, -1, 0, -1, -1, 1, 1, 1, 1, 0, -1, 0, 1,
            1, -1, -1, -1, 1, 0, 1, -1, 0, -1, 1, 1, 1, 1,
           -1, 1, 1, 0, -1, 1, 0, -1, -1, -1, -1,
                                                                0,
           -1, -1, 0, -1, 0, 0, -1, 1, -1, 1, 1, 1, 0, 1, 0,
            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, -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, 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, \ 0, \ -1, \ 0, \ -1, \ 1, \ -1,
           0, -1, -1, 0, 0, 1, -1, 1,
                                       0, 0, 0,
                                                  0, -1,
           -1, 0, 1, 0, 0, -1, 0, 1, 0, 0, 0,
                                                  0, 0,
                                                             0, 1, 1,
                                                         1,
           1, 0, -1, 1, 0, 0, -1, 1, 0, 0, -1,
                                                  0, -1,
                                                         0.
                                                            1. -1.
           -1, -1, 0, 0, 0,
                            0, 1, 1,
                                       1, -1,
                                             -1,
                                                  0,
                                                      1,
                                                         0,
                                                             0, -1,
           1, 0, 1, -1, -1, 0, 1, -1, 1, -1, 0,
                                                  1,
                                                      1,
           0, -1, 0, 0, -1, 1, -1, 0, 1, 1, 1,
                                                      0, -1,
                                                            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,
                                       0, -1, -1,
                                                 1, -1, -1, 0,
                                                                0,
           0, -1, 0, 1, -1, 0, 1, -1,
                                       0, 0, 1, -1,
                                                      0, -1,
                                                            1,
                                                                0,
           1, 0, 1, -1, 0, 0,
                                0,
                                    1,
                                       0,
                                           0, -1, 1,
                                                      0, -1, -1,
           1, -1, -1, -1, -1, 0,
                                    0, 1,
                                           0, -1,
                                                  1,
                                                     1.
                                                        -1.
                                                            1.
           -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, -1, 0,
                     1,
                         0, -1, -1, 1,
                                       0,
                                           1, 0,
                                                      0,
                                                  1,
                                                        -1,
           1, -1, -1, 1, 0, 0, 0,
                                    0, 0,
                                           0, -1, -1, -1, -1, 1, -1,
           1, 0, -1, 1, 1, -1, 1,
                                   0, 0, 1, -1, 0, -1, 0, 1, 1,
                                                                   0.
              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, 1, 0, -1, -1, 1, 1, 1, 0, -1, 1, 1, -1, -1,
           0, 1, 0, 1, 1,
                            0,
                                1, -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_train
    <579x3824 sparse matrix of type '<class 'numpy.float64'>'
           with 7378 stored elements in Compressed Sparse Row format>
y_train
    array([ 1, 1, -1, -1, 0, -1, 0, 1, 1, 0, -1, 0, -1, -1, 1,
           1, -1, -1, 1, 0, 1, -1, -1, 0, 0, 1, -1, 1, -1,
                                                                0, -1,
           -1, -1, -1, 0, 0, 1, -1, 0, 0, -1, 1, 1,
                                                     1, -1,
          -1, 1, 0, 1, -1, -1, 1, 1, -1, 1, 0,
                                                  1, 1,
                                                         0, 1,
                  0, 1, -1, -1, -1, -1, -1, -1, 0, -1,
                                                         1, -1,
           0, 1, 1, 0, 1, -1, 1, 0, -1, 1, -1, -1, 0,
                                                         0, -1,
          -1, -1, 1, -1, 0, 1, 1, 0, 1, 0, -1, 1, 1,
                                                         0,
                                                            0,
                                                                0.
                                                                    0,
           1, -1, 1,
                                0,
                                       0, -1,
                                              0,
                                                  0,
                      1, 1, 1,
                                    1,
                                                         0,
                                                            -1,
           -1, 1, 1, -1, 1, 0, 1, 1, 1,
                                              1.
                                                  0.
                                                      0. -1. -1.
                                                                0.
                  0, 0, 0, 1, 1, 0, -1, -1,
                                              0,
           0, -1,
                                                 0,
                                                     0, -1, -1, -1,
           -1, -1, 0,
                     0, -1, -1, 0, 1, -1, -1,
                                              1, -1,
                                                     0,
                                                        0, -1, -1,
           0, 0, -1, 0, 0, 1, 0, -1, -1, -1, 0, 1, 1, 1, 1, 1, 1,
           0,
              1, -1,
                     1, -1, -1, -1, 0, -1,
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           0])
from sklearn.svm import SVC
from sklearn.naive bayes import MultinomialNB
from sklearn.ensemble import RandomForestClassifier
from sklearn.tree import DecisionTreeClassifier
svm_model=SVC()
nb_model=MultinomialNB()
rf model=RandomForestClassifier()
```

de_model=DecisionTreeClassifier() lstmodel=[svm_model,nb_model,rf_model,de_model]

```
from sklearn.metrics import confusion_matrix,classification_report
for i in lstmodel:
 print(i)
 i.fit(x_train,y_train)
 y_pred=i.predict(x_test)
 print(classification_report(y_test,y_pred))
 SVC()
   precision recall f1-score support
          -1
                 0.68
                        0.52
                                0.59
                                         79
                 0.48
                        0.76
                                0.59
                                         79
                 0.72
                        0.51
                                0.59
                                         91
                                        249
                                0.59
      accuracy
     macro avg
                 0.63
                        0.59
                                0.59
                                        249
   weighted avg
                 0.63
                        0.59
                                0.59
                                        249
   *****************
   MultinomialNB()
             precision recall f1-score support
           -1
                 0.62
                        0.56
                                0.59
                                         79
           0
                 0.48
                        0.57
                                0.52
                                         79
                 0.64
                      0.59
                                0.62
      accuracy
                                0.57
                                        249
                 0.58
     macro avg
                        0.57
                                0.57
                                        249
   weighted avg
                 0.58
                        0.57
                                0.58
                                        249
   ******************
   RandomForestClassifier()
   *********************
             precision recall f1-score support
           -1
                 0.47
                       0.72
                                0.57
                                         79
           0
                 0.54
                        0.52
                                0.53
                                         79
           1
                 0.69
                        0.38
                                0.49
                                         91
                                0.53
                                        249
      accuracy
                0.56
                        0.54
                                        249
                                0.53
     macro avg
   weighted avg
                0.57
                        0.53
                                0.53
                                        249
   DecisionTreeClassifier()
   **************************************
             precision recall f1-score support
          -1
                 0.47
                        0.67
                                0.55
                                         79
                 0.47
                         0.44
                                0.46
                                         79
                 0.57
                        0.40
                                0.47
                                         91
                                0.50
                                        249
      accuracy
                 0.51
                         0.50
                                        249
     macro avg
                                0.49
   weighted avg
                 0.51
                        0.50
                                0.49
                                        249
   **************************************
```

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