```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import re
from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.model selection import train test split
from sklearn.linear model import LogisticRegression
from sklearn.metrics import accuracy score
from collections import Counter
import numpy as np
from sklearn.model selection import train test split
import nltk
nltk.download('stopwords')
print(stopwords.words('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', "it's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves', 'what', 'which', 'who', 'whom',
'their', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having', 'do',
'were', 'be', 'been', 'being', 'have', 'has', nau , naving , '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' 'above' 'helow'. 'to', 'from', 'up', 'down', 'in', 'out',
 'on', 'off', 'over', 'under', 'again', 'further', 'then', 'once',
'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some', 'such', 'no', 'nor',
each, Tew', 'more', 'most', 'other', 'some', 'such', 'no', 'nor', 'not', 'only', 'own', 'same', 'so', 'than', 'too', 'very', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'wouldn', "wouldn', "wo
"weren't", 'won', "won't", 'wouldn', "wouldn't"]
 [nltk data] Downloading package stopwords to /root/nltk data...
 [nltk data] Package stopwords is already up-to-date!
#Loading data from the csv file using pandas
X data=pd.read csv('/content/X data.csv',encoding='ISO-8859-1')
#Checking number of rows and columns
X data.shape
 (162980, 2)
```

```
#Printing firt 5 rows of data
X data.head()
{"type": "dataframe", "variable name": "X data"}
#Naming columns and reading data sets again
COL NAMES=['TEXT', 'FLAG']
X data=pd.read csv('/content/X data.csv',encoding='ISO-8859-
1', names=COL NAMES)
X_data.head()
{"type":"dataframe", "variable name": "X data"}
#counting number of missing values in X data
X data.isnull().sum()
TEXT
        4
FLAG
        7
dtype: int64
#Filling null values with -1
X data=X data.fillna(-1)
X data['TEXT'].replace(to replace="-1",value="Modi will win")
                                                  clean text
          when modi promised â(minimum government maxim...
1
2
          talk all the nonsense and continue all the dra...
3
          what did just say vote for modi welcome bjp t...
4
          asking his supporters prefix chowkidar their n...
          why these 456 crores paid neerav modi not reco...
162976
162977
          dear rss terrorist payal gawar what about modi...
162978
          did you cover her interaction forum where she ...
          there big project came into india modi dream p...
162979
          have you ever listen about like gurukul where ...
162980
Name: TEXT, Length: 162981, dtype: object
#checking the distributions of target column
X data['TEXT'].value counts()
TEXT
- 1
4
2019
clean text
should vote modi for cpas after years
lok sabha election 2019 live modi has ignored his own constituency
varanasi says priyanka gandhi
```

```
1
modi destroying india for personal benefit
sreeniwho announced buddha laughing nuclear testjust for sake dont
frowl before knowing the achievementthis not achievement modiits
nations achievementmake your heart bit enlarged
modi thunders indiaâ( entry into indiaâ( space club raga calls â(happy
theatre dayâ(
back basics jobs\nfarmers\nsmall businesses\ngst reform\neducation\
nhealth\nenvironment\nwater\ninfrastructure\ninvestment revival
national security indiaâ( armed forces are handling that donâ( worry
space narendra modi thanks
have you ever listen about like gurukul where discipline are
maintained even narendra modi rss only maintaining the culture indian
more attack politics but someone attack hinduism rss will take action
that proud for
Name: count, Length: 162977, dtype: int64
X_data['FLAG'].value_counts()
FLAG
1
            72250
            55213
0
- 1
            35510
- 1
category
                1
Name: count, dtype: int64
```

0 -->neutral tweet 1 -->positive tweet -1 --> negative tweet

Stemming is the process of reducing a word to its root form i.e Swimming to swim. We do this using the porter stemmer function

```
def stemming(content):
    # Convert content to string to handle non-string values
    content = str(content)
    stemmed_content=re.sub('[^a-zA-Z]',' ',content)
    stemmed_content=stemmed_content.lower()
    stemmed_content=stemmed_content.split()
    # Correct variable name from stemmed_conted to stemmed_content
    stemmed_content=[port_stem.stem(word) for word in stemmed_content
if not word in stopwords.words('english')]
    stemmed_content=' '.join(stemmed_content)
```

```
return stemmed content
X data['STEM TEXT']=X data['TEXT'].apply(stemming) #about 7mins to
complete this execution!!! S00000 L0000NG!!!!!
#Viewing new data
X data.head()
{"type": "dataframe", "variable_name": "X_data"}
print(X data['FLAG'])
          category
1
                - 1
2
                 0
3
                 1
4
                 1
162976
                - 1
                - 1
162977
162978
                 0
                 0
162979
162980
                 1
Name: FLAG, Length: 162981, dtype: object
#seperating data and label
X=X_data['STEM_TEXT'].values
Y=X data['FLAG'].values.astype(str)
print(X)
['clean text'
 'modi promis minimum govern maximum govern expect begin difficult job
reform state take year get justic state busi exit psu templ'
 'talk nonsens continu drama vote modi' ... 'cover interact forum
left'
 'big project came india modi dream project happen realiti'
 'ever listen like gurukul disciplin maintain even narendra modi rss
maintain cultur indian attack polit someon attack hinduism rss take
action proud']
print(Y)
['category' '-1' '0' ... '0' '0' '1']
# Calculate class distribution
class distribution = Counter(Y)
# Find classes with only one sample
classes to remove = [cls for cls, count in class distribution.items()
if count < 2]
```

```
# Remove samples belonging to the under-represented classes
mask = \sim np.isin(Y, classes to remove)
X = X[mask]
Y = Y[mask]
#SPLITTING DATA TO TRAIN AND TEST
X train,X test,Y train,Y test=train test split(X,Y,test size=0.2,strat
ify=Y,random state=2)
print(X.shape,Y train.shape,X test.shape)
(162980,) (130384,) (32596,)
print(X train)
['kaha tha nachoron sarkarscamist parti hai modi ruin countri dont
forget vote'
 'chines citizen power right speak'
 'missil defenc india strengthen modi govern' ...
 'pm narasimha rao communist lack vision like communist includ shastri
moraji desai today due vajpaye narendra modi other lack vision decis'
 'smita prakash modi fangirl would modi without ani would ani without
modi'
 'lok sabha elect campaign live make sure rahul defeat say prakash
karat financialxpress']
print(X test)
['modi anti nation lie creat commun divid mislead manipul data'
 'jaya pradha join bjp convass goten shorten amount also popular bjp
big rich leader bythat famou cine field enjoy modi also'
 'power popular leadership modi one tweet whole countri goe
desper' ... ''
 'modi address massiv ralli kurnool andhra pradesh via namo app'
 'watch video voic whatsapp section volunt modul narendra modi app']
#Converting textual Data to numerical data
vectorizer=TfidfVectorizer(lowercase=False)
# Fit and transform the original text data 'X text' (assuming this is
your original text data variable)
X = vectorizer.fit transform(X text)
# Now transform the training and testing sets using the fitted
vectorizer
X train=vectorizer.transform(X train)
X test=vectorizer.transform(X test)
print(X train)
  (0, 5877)
                0.23211504791694398
  (0, 7538)
                0.23548945446948935
  (0, 9418)
                0.354204268832559
```

```
(0, 10840)
                 0.2753084460429206
  (0, 13783)
                 0.47083934854041315
  (0, 16942)
                 0.06500689091180703
  (0, 19727)
                 0.25053561795816454
  (0, 22855)
                 0.4318397312580794
  (0, 26449)
                 0.4087484798215461
  (0, 28611)
                 0.2171804773161407
  (1, 4756)
                 0.5918469474568181
  (1, 5044)
                 0.4551164112905416
  (1, 20743)
                 0.3366327029443095
  (1, 22610)
                 0.3759564931415472
  (1, 25003)
                 0.4334989983070124
  (2, 6582)
                 0.5300020214201105
  (2, 10358)
                 0.35074894432798903
  (2, 12299)
                 0.23366050593559784
  (2, 16809)
                 0.43637908746264226
  (2, 16942)
                 0.0916516983467352
  (2, 25434)
                 0.5853623335377406
  (3, 405) 0.2523482267005551
  (3, 3567)
                 0.35357562788310454
  (3, 8468)
                 0.2191756447841152
  (3, 8638)
                 0.5081332140365633
     :
  (130381, 21875)
                       0.23022080815445803
  (130381, 24083)
                       0.25638451452365046
  (130381, 26956)
                      0.13746263513427678
  (130381, 28173)
                      0.20083845408518594
  (130381, 28539)
                      0.3585143380350224
  (130382, 1140) 0.5902580758686817
  (130382, 8875) 0.37522459976450834
  (130382, 16942)
                      0.12281166280105922
  (130382, 20820)
                      0.31199308145928956
  (130382, 24710)
                       0.3647377075919171
  (130382, 29193)
                       0.3930048745543684
  (130382, 29344)
                      0.33300614024675346
  (130383, 4076) 0.22549246814283908
  (130383, 6574) 0.2608135213898228
  (130383, 8051) 0.16215926088637278
  (130383, 9165) 0.43666985999467206
  (130383, 13931)
                       0.4285314472717363
  (130383, 15275)
                       0.21640635333884672
  (130383, 15360)
                       0.25961397199213104
  (130383, 15839)
                       0.18627600953805795
  (130383, 20820)
                       0.38204335439728315
  (130383, 21659)
                       0.18595018555306403
  (130383, 22961)
                       0.2584458163033459
  (130383, 23449)
                      0.16610043871510108
  (130383, 25794)
                      0.23804619772332206
print(X test)
```

```
(0, 1230)
              0.26958717265727083
(0, 5360)
              0.3121892578044157
(0, 6004)
              0.2877293106675747
(0, 6386)
              0.3445650741673561
(0, 7407)
              0.3619912275820138
(0, 15163)
              0.2895043909183832
(0, 16023)
              0.44650548079773944
(0, 16789)
              0.41099085054222745
(0, 16942)
              0.05972087114921265
(0, 18019)
              0.20408265440005308
(1, 909) 0.24233964199864133
(1, 1018)
              0.1940119208092101
(1, 3142)
              0.14963154522697322
(1, 3273)
              0.19966985781677263
(1, 4013)
              0.3035493266924947
(1, 5018)
              0.3035493266924947
(1, 5740)
              0.3035493266924947
(1, 8257)
              0.18786060786814415
(1, 8859)
              0.2158496436901553
(1, 9107)
              0.2003429948755047
(1, 10342)
              0.3035493266924947
(1, 13257)
              0.24283093573712491
(1, 13578)
              0.16215342950663475
(1, 14945)
              0.1276272236408723
(1, 16942)
              0.03186232085438606
(32592, 23569) 0.39236097060969066
(32592, 23918) 0.39236097060969066
(32592, 24964) 0.16872467473369665
(32592, 25723) 0.22384292133048414
(32592, 26234) 0.21329268036979332
(32594, 405)
              0.2572984313499474
(32594, 1106) 0.3839118348072669
(32594, 1396) 0.31007248344097077
(32594, 14649) 0.45649845172633596
(32594, 16203) 0.3686756673788957
(32594, 16942) 0.05947675840780109
(32594, 17905) 0.28432703479309335
(32594, 20803) 0.3451404640437686
(32594, 21798) 0.28523238717232124
(32594, 28371) 0.24716975759466478
(32595, 1396) 0.27895174324173294
(32595, 16942) 0.053507313051794896
(32595, 17232) 0.4767332578224748
(32595, 17969) 0.1762845855602194
(32595, 23700) 0.3469172423358947
(32595, 28397) 0.27296612439328555
(32595, 28589) 0.3263140040976818
(32595, 28601) 0.39772489655740606
```

```
(32595, 28827) 0.24696891876614663
  (32595, 29002) 0.37476711518967865
#Training the Machine learning model
model=LogisticRegression()
model.fit(X train,Y train)
/usr/local/lib/python3.10/dist-packages/sklearn/linear model/
logistic.py:469: ConvergenceWarning: lbfgs failed to converge
(status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear model.html#logistic-
regression
  n iter i = check optimize result(
#Model Evaluation accurace on traininf data
X train prediction = model.predict(X train)
training data accuracy=accuracy score(Y train,X train prediction)
print('Accuracy score on the training Data:',training data accuracy)
Accuracy score on the training Data: 0.8684807951895939
```

Testing Data Accuracy = 86.84%

```
#Model Evaluation accurace on training data
X_test_prediction = model.predict(X_test)
test_data_accuracy=accuracy_score(Y_test,X_test_prediction)
print('Accuracy score on the training Data:',test_data_accuracy)
Accuracy score on the training Data: 0.8392747576389741
```

Model Accuracy = 83.92%

We are saving this model to be used for later to use model directily without having to train multiple times! using the pickle library!!!

```
import pickle
#Saving the pretrained model to a .sav file using Pickle moduld
filename='trained_model.sav'
pickle.dump(model,open(filename,'wb'))
```