Twitter Sentiment Analysis

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

from google.colab import drive
drive.mount('/content/drive')

→ Mounted at /content/drive

dff = pd.read_csv("/content/drive/MyDrive/COLLAB/Modi_Tweets.csv" , encoding =
dff.head()

→		clean_text	category
	0	when modi promised â□□minimum government maxim	-1.0
	1	talk all the nonsense and continue all the dra	0.0
	2	what did just say vote for modi welcome bjp t	1.0
	3	asking his supporters prefix chowkidar their n	1.0
	4	answer who among these the most powerful world	1.0

Sampling of data

```
# prompt: create a sample of random 10000 rows only

df = dff.sample(n=10000, random_state=1)

df.shape

→ (10000, 2)

dff.shape

→ (162980, 2)
```

```
df.info()
    <class 'pandas.core.frame.DataFrame'>
    Index: 10000 entries, 102593 to 62391
    Data columns (total 2 columns):
         Column
                      Non-Null Count
                                      Dtype
     0
         clean_text 10000 non-null object
     1
         category
                      10000 non-null
                                      float64
    dtypes: float64(1), object(1)
    memory usage: 234.4+ KB
df.shape
\rightarrow (10000, 2)
df.isnull().sum()
→ clean_text
    category
    dtype: int64
```

check distribution of the target count

Data Exploration

```
import pandas as pd
import numpy as np
# Assuming your DataFrame is named df and the column is named 'category'
# Replace non-finite values with a placeholder (e.g., 0)
df['category'].fillna(0, inplace=True)
# Convert the column to integers
df['category'] = df['category'].astype(int)
# If you want to specifically map float values to integer values
value_mapping = \{1.0: 1, 0.0: 0, -1.0: -1\}
df['category'] = df['category'].map(value_mapping)
df["category"].value_counts()
→ category
      1
           4465
     0
           3376
     -1
           2159
    Name: count, dtype: int64
```

prompt: generate a column length that contains the character count of column
df["length"] = df["clean_text"].str.len()
df

→		clean_text	category	length
	102593	opportunistic alliance which has purpose other	1	223
	148753	modi winsbuy mutual fundsgood shares govt bond	1	228
	2704	shameful return gaali politics now andhra prad	0	109
	138001	kuntia dubs modi kcr thieves cong alone saviou	0	55
	73506	modi has liberated indiabuilt india and looted	1	122
	27748	kutta bhaunke laath maro aur yeh pagal nikla m	1	204
	107598	had identified raghuram rajan sonias mole stoo	-1	215
	151666	modi and all minister should also give intervi	0	156
	77114	media also have never ask modi many critical i	1	97
	62391	congratulations modi proud and new india	1	40

10000 rows × 3 columns

df.isnull().sum()

clean_text category length dtype: int64

Roughly how many positive and negative feedbacks are

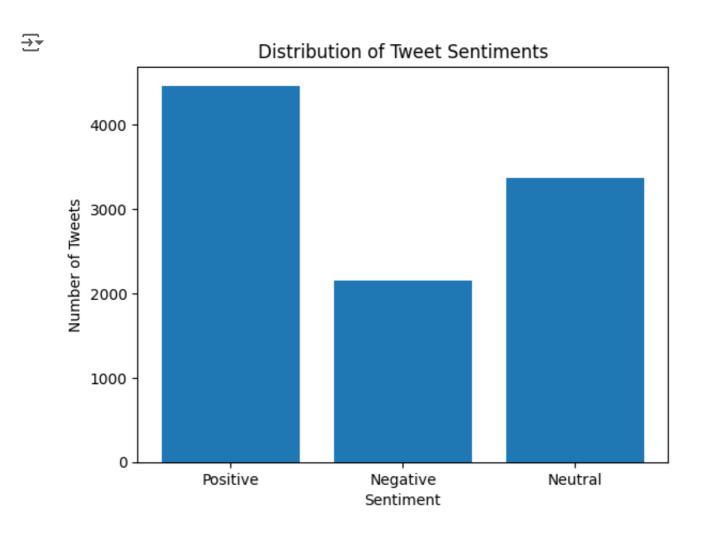
```
df['category'].value_counts()
    category
     1
          4465
     0
          3376
    -1
          2159
    Name: count, dtype: int64
num_positive_feedback = df[df['category'] == 1].shape[0]
num_negative_feedback = df[df['category'] == -1].shape[0]
num_neutral_feedback = df[df['category'] == 0].shape[0]
print(f"Number of positive feedbacks: {num_positive_feedback}")
print(f"Number of negative feedbacks: {num_negative_feedback}")
print(f"Number of neutral feedbacks: {num_neutral_feedback}")
    Number of positive feedbacks: 4465
    Number of negative feedbacks: 2159
    Number of neutral feedbacks: 3376
```

```
# prompt: bar plot for negative positive and neutral tweets with labelling as r
import matplotlib.pyplot as plt
# Count the number of positive, negative, and neutral tweets
num_positive = df[df['category'] == 1].shape[0]
num_negative = df[df['category'] == -1].shape[0]
num_neutral = df[df['category'] == 0].shape[0]

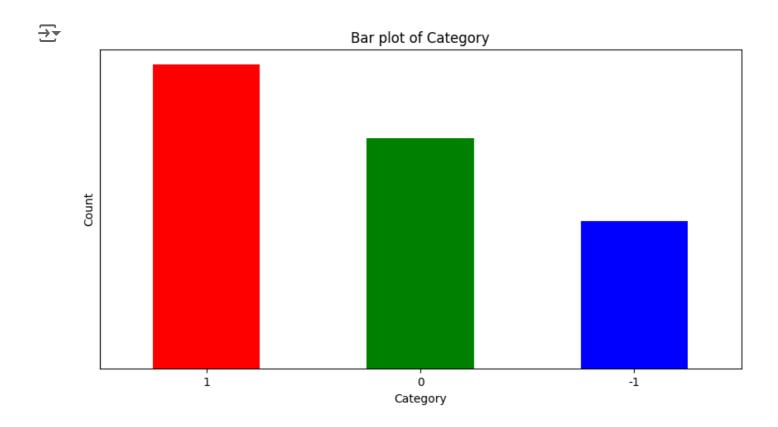
# Create a bar plot
plt.bar(['Positive', 'Negative', 'Neutral'], [num_positive, num_negative, num_r

# Add labels
plt.xlabel('Sentiment')
plt.ylabel('Number of Tweets')
plt.title('Distribution of Tweet Sentiments')

# Show the plot
plt.show()
```



```
import matplotlib.pyplot as plt
import numpy as np
df['category'].value_counts().plot(kind='bar', color=['red','green','blue'], fi
plt.xlabel("Category")
plt.ylabel("Count")
plt.title("Bar plot of Category")
plt.xticks(rotation=0)
plt.yticks(np.arange(0, 0000, 10000))
plt.show()
```



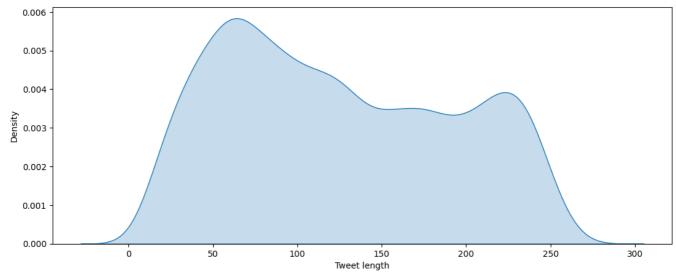
plt.figure(figsize=(13,5)) sns.kdeplot(df['length'], shade= True) plt.xlabel("Tweet length")



<ipython-input-19-2a0aa83ea8cc>:2: FutureWarning:

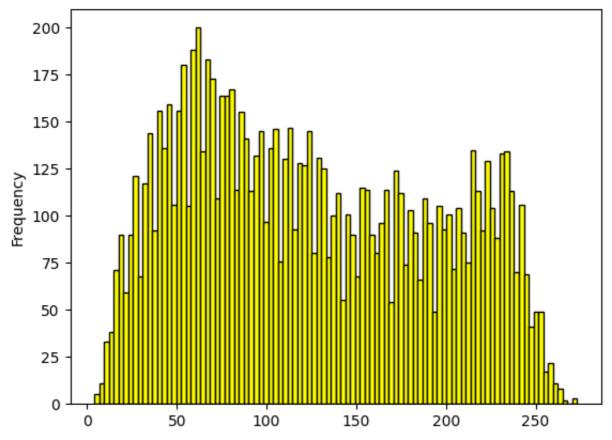
`shade` is now deprecated in favor of `fill`; setting `fill=True`. This will become an error in seaborn v0.14.0; please update your code.

sns.kdeplot(df['length'], shade= True) Text(0.5, 0, 'Tweet length')



df["length"].plot(bins = 100 , kind = "hist", color="yellow",edgecolor='black')





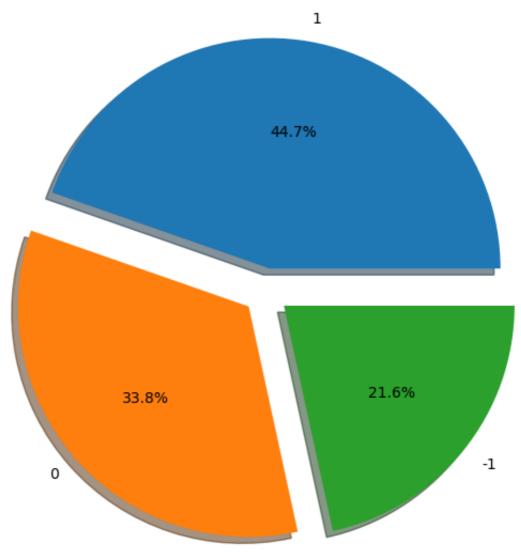
df.describe()

→		category	length
	count	10000.000000	10000.000000
	mean	0.230600	124.747300
	std	0.780567	67.931072
	min	-1.000000	4.000000
	25%	0.000000	66.000000
	50%	0.000000	116.000000
	75%	1.000000	183.000000
	max	1.000000	273.000000

```
import pandas as pd
explode = [0.1] * len(df["category"].value_counts())
plt.figure(figsize=(10, 7))
plt.pie(df["category"].value_counts(), labels=df["category"].value_counts().inc
plt.title("Category of Sentiments with percentage in dataset")
plt.show()
```



Category of Sentiments with percentage in dataset



df.describe()

→ ▼		category	length
	count	10000.000000	10000.000000
	mean	0.230600	124.747300
	std	0.780567	67.931072
	min	-1.000000	4.000000
	25%	0.000000	66.000000
	50%	0.000000	116.000000
	75%	1.000000	183.000000
	max	1.000000	273.000000

get the longest message review

df[df["length"] == 273.000000]

→		clean_text	category	length
	93146	startup policy biggest successful policy under	1	273

get the shortest message review

get the average message review

df[df["length"] == 125]

→		clean_text	category	length
	57697	pride moment for nation great achievement our	1	125

23616	thread explicitly proving modi has left every	1	125
29132	masterpiece writing the best thing like about	1	125
133935	unemployment year high modi and yogi busy chan	1	125
50587	chidambaram enough your drama dont need your m	1	125
35937	pranam sadgurui personally feel that you shoul	-1	125
30873	these people still asking votes stupid emotion	-1	125
95203	pahle ground surgical strike fir hawa airstrik	0	125
133325	assam rest northeast suffering from infiltrati	-1	125
161009	says and bsp have lost ground during lok sabha	0	125
104539	pahle ground surgical strike fir hawa airstrik	0	125
146104	dont blame modi after all hes just self confes	0	125
157023	did you ask this modi who will contest from tw	0	125
137846	you see modi unke pass toh qualification bhi n	0	125
40913	such the impact his wonderful leadership that	1	125
18518	lakhs modi have made many above poverty line r	1	125
85562	indias identity still unclear due article 370e	1	125
159306	sir thats bad long marans were the owners all	-1	125
119596	modi government did tremendous work last year	1	125
115181	hahahahahaha why dont you ever find the cour	1	125
65771	they had last time one more raju shetti was nd	1	125
77931	modi bhakt says that modi have family then why	0	125
83490	this chowkidar has shown the courage carry out	0	125
131206	india everyone calls statesman and they doubt	0	125
113647	the same rogues keep calling names modi they m	1	125
130365	lakhs youngsters assam have been able earn liv	1	125
153831	air marshal dhillon appointed new strategic fo	1	125
130046	perception that some dangerous conspiracy has	-1	125
56505	after said this swamy will the biggest embarra	1	125
59880	tell where taking tge credit some leaders forg	1	125
99206	all loot hannened during upa and pappu and mot	1	125

00200	an look happonou dannig apa and pappa and mot		120
101430	sab saath sab bikas creation new india and new	1	125
114548	how hard you try people are not fools everyone	1	125
41033	prime minister narendra modi will address the	1	125
67340	the same guy who brought aadharand modi didnt	0	125
64299	modi the only man who has this type will power	1	125
152171	really true sir\nâ□□modi mistakeâ□□ chandrabab	1	125
62388	sir there are many such reason modi the only I	1	125
89850	neither modi nor congress president rahul gand	0	125
139480	great toronto canada and catch with colleagues	1	125
11546	nothing hidden under modi govt but you piddis	-1	125
137787	please dont like modis promises are tired that	1	125
26591	rajan thats why have reginated out your post h	1	125
138790	feel horrible for chandra babu naidu has becom	-1	125
66554	manmohan singh was internally unhappy with son	-1	125
60695	and the best part always gives proof\nsee phot	1	125
110678	the accused was identified who facebook post s	0	125
141806	quoting masood azhar totally unacceptable but	0	125
14124	some more noteworthy achievements india today	1	125
76142	election commission examine whether modiâ□□ mi	-1	125

Word Cloud

import important modules

```
#!pip install wordcloud
#!pip install numpy matplotlib pillow
from wordcloud import WordCloud, STOPWORDS
import numpy as np
import matplotlib.pyplot as plt
import PIL.Image
```

v positive dataframe

positive_Df = df[df["category"] == 1]
positive_Df

→		clean_text	category	length
	102593	opportunistic alliance which has purpose other	1	223
	148753	modi winsbuy mutual fundsgood shares govt bond	1	228
	73506	modi has liberated indiabuilt india and looted	1	122
	138003	yes danger for sangh parivar they wont win thi	1	100
	77303	why narendra modis address the nation successf	1	108
	77822	this the only fitting cartoon which doesnt hav	1	71
	109442	just saw indias no1propaganda channelrepulsive	1	241
	27748	kutta bhaunke laath maro aur yeh pagal nikla m	1	204
	77114	media also have never ask modi many critical i	1	97
	62391	congratulations modi proud and new india	1	40

4465 rows × 3 columns

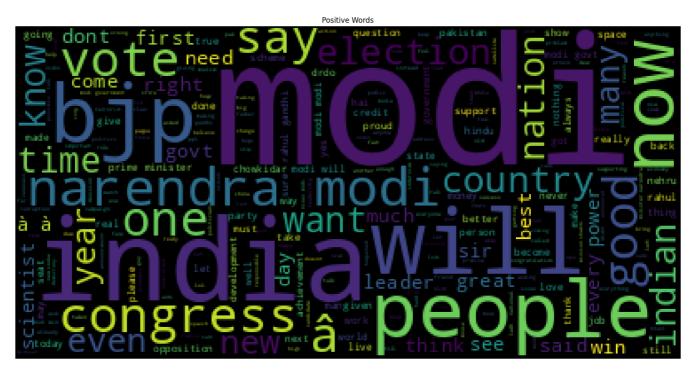
one_positive_state = " ".join(positive_Df["clean_text"].astype(str))
len(one_positive_state)

→ 631723

```
plt.figure(figsize = (20,20))
text = one_positive_state

wc = WordCloud(stopwords=STOPWORDS, contour_color="black", contour_width=3, mir
plt.imshow(wc)
plt.title("Positive Words")
plt.axis("off")
plt.show()
```





neutral_Df = df[df["category"] == 0]
neutral_Df

→		clean_text	category	length
	2704	shameful return gaali politics now andhra prad	0	109
	138001	kuntia dubs modi kcr thieves cong alone saviou	0	55
	84977	congress its desperation questioned the and ga	0	157
	148165	global	0	7
	127794	you fear that modi may treat every indian citi	0	93
	56022	who said that these build modi matter highligh	0	74
	145921	election commission examining modis address an	0	65
	17764	got money promised modi and will vote for him	0	52
	8516	india markets election tourism forget the taj	0	90
	151666	modi and all minister should also give intervi	0	156

3376 rows × 3 columns

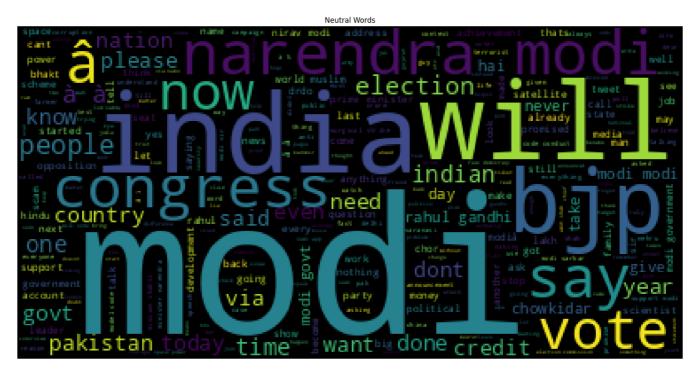
one_neutral_state = " ".join(neutral_Df["clean_text"].astype(str))
len(one_neutral_state)

→ 304901

```
plt.figure(figsize = (20,20))
text = one_neutral_state

wc = WordCloud(stopwords=STOPWORDS, contour_color="black", contour_width=3, mir
plt.imshow(wc)
plt.title("Neutral Words")
plt.axis("off")
plt.show()
```





negative dataframe

prompt: separate negative dataframe
negative_Df = df[df["category"] == -1]
negative_Df

_				
→		clean_text	category	length
	91341	now you realized that you are losing against m	-1	78
	62588	lots fake news keep you viting for modi	-1	40
	31740	you fake gandhis can never come with original	-1	79
	142164	before placing campaign for campaigning upbiha	-1	202
	16545	seriouslyhope you are not supporting modi	-1	41
	28641	â□□modi journey common manâ□□ trailer out web	-1	86
	64013	started working\nâ□¢ before modi had his first	-1	258
	114828	you are accepting that jumla modi told false p	-1	59
	123630	bjp supporters say job opportunities india hav	-1	72
	107598	had identified raghuram rajan sonias mole stoo	-1	215

2159 rows × 3 columns

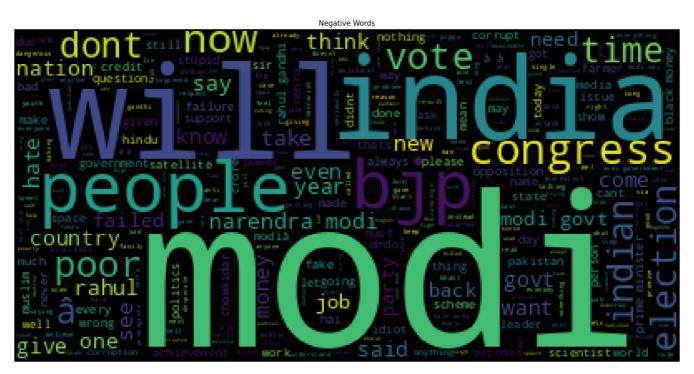
one_negative_state = " ".join(negative_Df["clean_text"].astype(str))
len(one_negative_state)

→ 320846

```
plt.figure(figsize = (20,20))
text = one_negative_state
```

wc = WordCloud(stopwords=STOPWORDS, contour_color="black", contour_width=3, mir
plt.imshow(wc)
plt.title("Negative Words")
plt.axis("off")
plt.show()





df.head()

_				
→		clean_text	category	length
	102593	opportunistic alliance which has purpose other	1	223
	148753	modi winsbuy mutual fundsgood shares govt bond	1	228
	2704	shameful return gaali politics now andhra prad	0	109
	138001	kuntia dubs modi kcr thieves cong alone saviou	0	55
	73506	modi has liberated indiabuilt india and looted	1	122

Y TEXT CLEANING

Creating A pipeline to remove Punctuations, Stopwords and Perform Count Vectorization

import string
string.punctuation

```
import string
import nltk
from nltk.corpus import stopwords
nltk.download('stopwords')
# Store stopwords in a variable
stop words = set(stopwords.words('english'))
def message_cleans(data_sample):
    # Use set for faster membership checking
    test punc removed = [char for char in data sample if char not in string.pur
    test_punc_join = "".join(test_punc_removed)
    # Use list comprehension and set membership check for stopwords
    test punc removed join clean = [word for word in test punc join.split() if
    return test_punc_removed_join_clean
[nltk_data] Downloading package stopwords to /root/nltk_data...
     [nltk_data]
                  Unzipping corpora/stopwords.zip.
df['clean_text'] = df['clean_text'].astype(str)
modi tweet = df['clean text'].apply(message cleans)
print(df['clean_text'])
    102593
              opportunistic alliance which has purpose other...
    148753
              modi winsbuy mutual fundsgood shares govt bond...
               shameful return gaali politics now andhra prad...
    2704
              kuntia dubs modi kcr thieves cong alone saviou...
    138001
    73506
              modi has liberated indiabuilt india and looted...
              kutta bhaunke laath maro aur yeh pagal nikla m...
    27748
              had identified raghuram rajan sonias mole stoo...
    107598
    151666
              modi and all minister should also give intervi...
    77114
              media also have never ask modi many critical i...
    62391
                        congratulations modi proud and new india
    Name: clean_text, Length: 10000, dtype: object
```

```
print(modi_tweet)
```

```
102593
               [opportunistic, alliance, purpose, defeat, bjp...
               [modi, winsbuy, mutual, fundsgood, shares, gov...
    148753
               [shameful, return, gaali, politics, andhra, pr...
    2704
    138001
               [kuntia, dubs, modi, kcr, thieves, cong, alone...
               [modi, liberated, indiabuilt, india, looted, i...
    73506
    27748
               [kutta, bhaunke, laath, maro, aur, yeh, pagal,...
    107598
               [identified, raghuram, rajan, sonias, mole, st...
               [modi, minister, also, give, interview, region...
    151666
               [media, also, never, ask, modi, many, critical...
    77114
    62391
                      [congratulations, modi, proud, new, india]
    Name: clean_text, Length: 10000, dtype: object
modi_tweet = pd.Series(modi_tweet)
modi tweet
    102593
               [opportunistic, alliance, purpose, defeat, bjp...
    148753
               [modi, winsbuy, mutual, fundsgood, shares, gov...
               [shameful, return, gaali, politics, andhra, pr...
    2704
               [kuntia, dubs, modi, kcr, thieves, cong, alone...
    138001
               [modi, liberated, indiabuilt, india, looted, i...
    73506
    27748
               [kutta, bhaunke, laath, maro, aur, yeh, pagal,...
    107598
               [identified, raghuram, rajan, sonias, mole, st...
    151666
               [modi, minister, also, give, interview, region...
               [media, also, never, ask, modi, many, critical...
    77114
                      [congratulations, modi, proud, new, india]
    62391
    Name: clean_text, Length: 10000, dtype: object
```

Concatenating the TF_IDF and Bag of words

```
!pip install scikit—learn
```

```
import nltk
from nltk.corpus import stopwords
from nltk.tokenize import RegexpTokenizer
from sklearn.feature extraction.text import CountVectorizer
df["clean_text"]
    102593
              opportunistic alliance which has purpose other...
              modi winsbuy mutual fundsgood shares govt bond...
    148753
    2704
              shameful return gaali politics now andhra prad...
    138001
              kuntia dubs modi kcr thieves cong alone saviou...
    73506
              modi has liberated indiabuilt india and looted...
              kutta bhaunke laath maro aur yeh pagal nikla m...
    27748
              had identified raghuram rajan sonias mole stoo...
    107598
              modi and all minister should also give intervi...
    151666
              media also have never ask modi many critical i...
    77114
                        congratulations modi proud and new india
    62391
    Name: clean_text, Length: 10000, dtype: object
```

COUNT VECTORIZATION (BAG OF WORDS)

prompt: create a dataframe containing X

import pandas as pd
df_X = pd.DataFrame(X.toarray(), columns=vectorizer.get_feature_names_out())
df_X

→		00000	000kwkajhatka	0554	0800	100	1000	10000	100000	1000000	100
	0	0	0	0	0	0	0	0	0	0	
	1	0	0	0	0	0	0	0	0	0	
	2	0	0	0	0	0	0	0	0	0	
	3	0	0	0	0	0	0	0	0	0	
	4	0	0	0	0	0	0	0	0	0	
	9995	0	0	0	0	0	0	0	0	0	
	9996	0	0	0	0	0	0	0	0	0	
	9997	0	0	0	0	0	0	0	0	0	
	9998	0	0	0	0	0	0	0	0	0	
	9999	0	0	0	0	0	0	0	0	0	

10000 rows × 20659 columns

TF-IDF PERFORMING ON THE DOCUMENTS FOR VECTORIZATION

```
# prompt: perform tf-idf on dataset

from sklearn.feature_extraction.text import TfidfVectorizer

# Create a TfidfVectorizer object
vectorizer = TfidfVectorizer()

# Fit and transform the data
X = vectorizer.fit_transform(df['clean_text'])

# Convert the sparse matrix to a dense array
X = X.toarray()

# Print the shape of the transformed data
print(X.shape)

$\frac{10000}{2000}, 20659)
```

df_T

prompt: create a data frame of X_combined which is a numpy.ndarray
import pandas as pd
Convert the NumPy array to a Pandas DataFrame
df_T = pd.DataFrame(X)
Print the DataFrame

→		0	1	2	3	4	5	6	7	8	9	 20649	20650	20651	2065
	0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	3	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9995	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9996	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9997	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9998	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9999	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0

10000 rows × 20659 columns

CONCATENATION OF BOTH TF-IDF AND BOW METHODS

- TF-IDF (Term Frequency and Inverse Document Frequency)
- BoW (Bag Of Words)

[0., 0., 0., ..., 0., 0., 0.], [0., 0., 0., ..., 0., 0., 0.]]) $\hbox{\it\# prompt: create a data frame of X_combined which is a numpy.ndarray}$

import pandas as pd

Convert the NumPy array to a Pandas DataFrame
df_combined = pd.DataFrame(X_combined)

Print the DataFrame
df_combined

→		0	1	2	3	4	5	6	7	8	9	 41308	41309	41310	4131
	0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	3	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9995	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9996	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9997	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9998	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0
	9999	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0

10000 rows × 41318 columns

DataSet Splitting

INDEPENDANT VARIABLES

x = df_T
print("X SHAPE : ",x.shape)
x

→ X SHAPE : (10000, 20659)

	0	1	2	3	4	5	6	7	8	9	• • •	20649	20650	20651	20652
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
3	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
9995	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
9996	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
9997	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
9998	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
9999	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0

10000 rows × 20659 columns

x.shape

(10000, 20659)

V DEPENDANT VARIABLE

```
y = df['category']
print("Y SHAPE : ",y.shape[0])
print()
У
→ Y SHAPE : 10000
    102593
              1
    148753
              1
    2704
    138001
    73506
              1
    27748
              1
    107598
             -1
    151666
    77114
              1
    62391
              1
    Name: category, Length: 10000, dtype: int64
```

SPLITTING TRAIN TEST DATASETS

```
from sklearn.model_selection import train_test_split
x_train , x_test , y_train , y_test = train_test_split(x,y,test_size = 0.2)
```

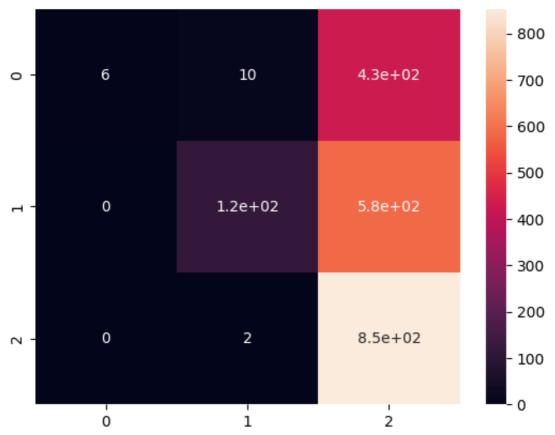
TRAIN & TEST NAIVE BASED CLASSIFIER MODEL

```
from sklearn.naive_bayes import MultinomialNB
NB_classifier = MultinomialNB()
NB_classifier.fit(x_train , y_train)
TymultinomialNB
```

MultinomialNB()

from sklearn.metrics import classification_report, confusion_matrix
predicting the test result
y_pred = NB_classifier.predict(x_test)
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm , annot = True)

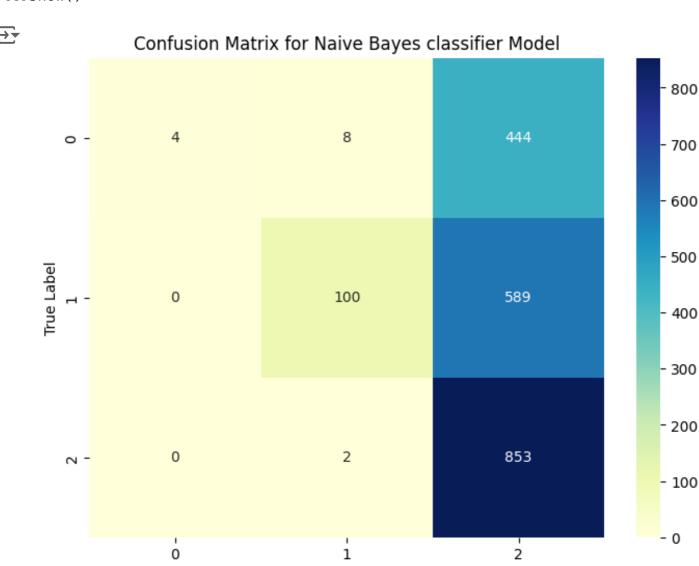




from sklearn.metrics import classification_report
print("******Naive Bayes using TF-IDF only******")
print(classification_report(y_test,y_pred))

\rightarrow	*****Naive Ba	ayes using	TF-IDF onl	y*****	
		precision	recall	f1-score	support
	-1	1.00	0.01	0.02	456
	0	0.91	0.15	0.25	689
	1	0.45	1.00	0.62	855
	accuracy			0.48	2000
	macro avg	0.79	0.38	0.30	2000
	weighted avg	0.73	0.48	0.36	2000

```
# Plot the confusion matrix
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt="d", cmap="YlGnBu")
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.title("Confusion Matrix for Naive Bayes classifier Model")
plt.show()
```



Predicted Label

Now Applying Losgistic model on this

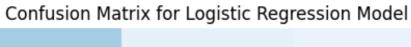
```
# prompt: apply logistic model
import matplotlib.pyplot as plt
from sklearn.linear model import LogisticRegression
# Create a LogisticRegression object
logistic_model = LogisticRegression()
# Fit the model on the training data
logistic_model.fit(x_train, y_train)
# Predict the labels for the test data
y_pred_logistic = logistic_model.predict(x_test)
# Evaluate the model's performance
accuracy_logistic = logistic_model.score(x_test, y_test)
print("Accuracy of Logistic Regression model:", accuracy_logistic)
# Print the classification report
print(classification_report(y_test, y_pred_logistic))
# Create a confusion matrix
cm_logistic = confusion_matrix(y_test, y_pred_logistic)
```

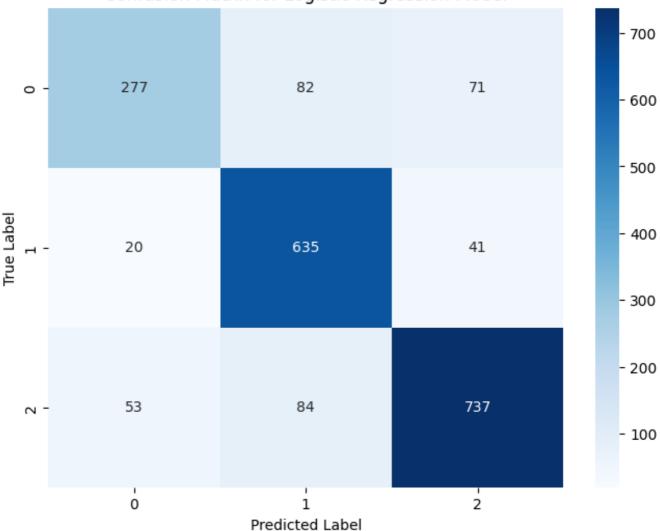
→ ▼	Accuracy of l	ogistic Reg		del: 0.765 f1-score	support
		p			
	-1	0.86	0.45	0.59	456
	0	0.73	0.85	0.78	689
	1	0.77	0.86	0.81	855
	accuracy			0.77	2000
	macro avg	0.79	0.72	0.73	2000
	weighted avg	0.78	0.77	0.75	2000

Training accuracy

```
# Plot the confusion matrix
plt.figure(figsize=(8, 6))
sns.heatmap(cm_logistic, annot=True, fmt="d", cmap="Blues")
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.title("Confusion Matrix for Logistic Regression Model")
plt.show()
```







```
y_predd = NB_classifier.predict(x_train)
print(classification_report(y_train,y_predd))
```

→		precision	recall	f1-score	support
	-1	0.98	0.78	0.87	1729
	0	0.98	0.72	0.83	2680
	1	0.77	0.99	0.87	3591
	accuracy			0.86	8000
	macro avg	0.91	0.83	0.85	8000
	weighted avg	0.88	0.86	0.85	8000

```
print(classification_report(y_test, y_pred1))
```

```
NameError

in the state of the
```

Logistic Test Accuracy check

```
print(classification_report(y_test, y_pred1))
```

CATBOOST CLASSIFIER

!pip install catboost

. .

→ Collecting catboost

```
Downloading catboost-1.2.5-cp310-cp310-manylinux2014_x86_64.whl (98.2 MB)
                                            - 98.2/98.2 MB 3.7 MB/s eta 0:0
Requirement already satisfied: graphviz in /usr/local/lib/python3.10/dist-p
Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist
Requirement already satisfied: numpy>=1.16.0 in /usr/local/lib/python3.10/d
Requirement already satisfied: pandas>=0.24 in /usr/local/lib/python3.10/di
Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-pack
Requirement already satisfied: plotly in /usr/local/lib/python3.10/dist-pac
Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packag
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/pyt
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/di
Requirement already satisfied: tzdata>=2022.1 in /usr/local/lib/python3.10/
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.1
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/di
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/d
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.1
Requirement already satisfied: tenacity>=6.2.0 in /usr/local/lib/python3.10
Installing collected packages: catboost
Successfully installed cathoost-1.2.5
```

from catboost import CatBoostClassifier
cat_model = CatBoostClassifier(verbose = 500)
training
cat_model.fit(x_train, y_train)

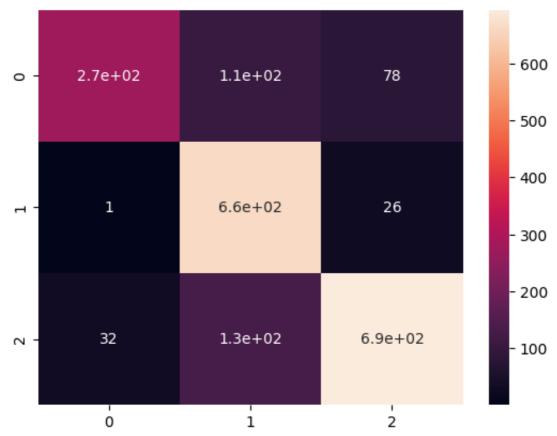
Learning rate set to 0.087979

0: learn: 1.0792775 total: 950ms remaining: 15m 49s 500: learn: 0.5381602 total: 4m 3s remaining: 4m 2s 999: learn: 0.4588707 total: 7m 22s remaining: 0us

<catboost.core.CatBoostClassifier at 0x7ab63a9c76a0>

y_pred2 = cat_model.predict(x_test)
cm = confusion_matrix(y_test , y_pred2)
sns.heatmap(cm, annot = True)





print("********Catboost only tfidf********")
print(classification_report(y_test, y_pred2))

******Catboost only tfidf***** recall f1-score precision support -10.89 0.60 0.71 456 0.74 0.96 0.83 689 0 0.87 0.84 1 0.81 855 0.81 2000 accuracy 0.83 0.79 0.80 2000 macro avg weighted avg 0.83 0.81 0.81 2000

df.info()
df["category"].value_counts()

<<class 'pandas.core.frame.DataFrame'>
 Index: 10000 entries, 102593 to 62391
 Data columns (total 3 columns):
 # Column Non-Null Count Dtype

0 clean_text 10000 non-null object 1 category 10000 non-null int64 2 length 10000 non-null int64

dtypes: int64(2), object(1)

memory usage: 312.5+ KB

category

1 4465

0 3376

-1 2159

Name: count, dtype: int64

Start coding or generate with AI.

import tensorflow as tf
from transformers import BertTokenizer, TFBertForSequenceClassification
import pandas as pd

Load the dataset (Assuming df is already loaded)

df = pd.read_csv("your_dataset.csv")

Preprocess the data
df["category"] = df["category"].astype(int)

Create a tokenizer

tokenizer = BertTokenizer.from pretrained("bert-base-uncased")

```
# Encode the text
encoded_data = tokenizer.batch_encode_plus(
    df["clean_text"].values.tolist(),
    add_special_tokens=True,
    return attention mask=True,
    padding="max_length",
    max_length=64, # Adjust max_length based on your dataset and available men
    return_tensors="tf",
    truncation=True,
)
# Create a TensorFlow dataset
input ids = encoded data["input ids"]
attention_masks = encoded_data["attention_mask"]
dataset = tf.data.Dataset.from_tensor_slices((input_ids, attention_masks, df["c
dataset = dataset.shuffle(100).batch(16) # Reduce batch size
# Load the BERT model
model = TFBertForSequenceClassification.from_pretrained("bert-base-uncased")
# Compile the model
model.compile(
    optimizer="adam",
    loss="sparse_categorical_crossentropy",
    metrics=["accuracy"],
)
# Train the model
model.fit(dataset, epochs=5)
# Evaluate the model
evaluation = model.evaluate(dataset)
print("Evaluation Loss:", evaluation[0])
print("Evaluation Accuracy:", evaluation[1])
```

```
# prompt: apply RNN MODEL to df dataset
import tensorflow as tf
from tensorflow.keras.layers import Embedding, LSTM, Dense, Input, GlobalMaxPoc
from tensorflow.keras.models import Model
from tensorflow.keras.callbacks import EarlyStopping
import tensorflow as tf
import tensorflow_hub as hub
max_length = df['clean_text'].str.len().max()
# Define the model
inputs = Input(shape=(max_length,), name="input_layer")
inputs = tf.expand_dims(inputs, axis=0)
# Use pre-trained word embeddings or remove this line if you want to train embe
# embedding_layer = Embedding(vocab_size, embedding_dim, input_length=max_lengt
lstm_layer = LSTM(128, return_sequences=True)(inputs)
# Add Batch Normalization for stability
lstm_layer = BatchNormalization()(lstm_layer)
# You can try Bidirectional LSTM for better representation learning
# lstm_layer = Bidirectional(LSTM(128, return_sequences=True))(embedding_layer)
global max pooling layer = GlobalMaxPooling1D()(lstm layer)
dropout_layer = Dropout(0.2)(global_max_pooling_layer)
outputs = Dense(3, activation="softmax")(dropout_layer)
model = Model(inputs=inputs, outputs=outputs)
# Compile the model
model.compile(loss="sparse_categorical_crossentropy", optimizer="adam", metrics
# Define early stopping to prevent overfitting
early_stopping = EarlyStopping(monitor='val_loss', patience=3, restore_best_wei
# Train the model with early stopping
model.fit(dataset, epochs=20, validation_split=0.2, callbacks=[early_stopping])
# Evaluate the model
evaluation = model.evaluate(dataset)
print("Evaluation Loss:", evaluation[0])
print("Evaluation Accuracy:", evaluation[1])
```

```
print(inputs.shape)
```

Start coding or generate with AI.

RNN

```
# prompt: split dataset df into train and test for RNN
X_train, X_test, y_train, y_test = train_test_split(df, y, test_size=0.2)
def build_vocabulary(text_corpus):
  Builds a vocabulary from a text corpus.
  Args:
    text_corpus: A list of strings representing the text corpus.
  Returns:
    A dictionary mapping words to their frequency in the corpus.
  vocabulary = {}
  for text in text_corpus:
    for word in text.split():
      word = word.lower().strip()
      if word not in vocabulary:
        vocabulary[word] = 0
      vocabulary[word] += 1
  return vocabulary
!pip install genism
```

```
# prompt: Now apply RNN
vocab_size = 10000 # Replace with the actual vocabulary size
embedding_dim = 128 # Replace with the desired embedding dimension
max length = 100 # Replace with the maximum sequence length
# Define the model
model = tf.keras.Sequential([
    tf.keras.layers.Embedding(vocab_size, embedding_dim, input_length=max_lengt
    tf.keras.layers.LSTM(128, return_sequences=True),
    tf.keras.layers.GlobalMaxPooling1D(),
    tf.keras.layers.Dense(3, activation="softmax")
1)
# Compile the model
model.compile(loss="sparse_categorical_crossentropy", optimizer="adam", metrics
# Train the model
model.fit(x_train, y_train, epochs=10)
# Evaluate the model
evaluation = model.evaluate(X_test, y_test)
print("Evaluation Loss:", evaluation[0])
print("Evaluation Accuracy:", evaluation[1])
```

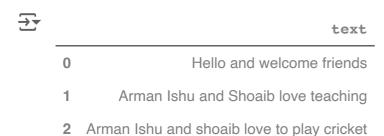
Dont Touch me!!!!!!!!!

Download the required NLTK data:

text = ["Hello and welcome friends", "Arman Ishu and Shoaib love teaching" , "Ar

prompt: convert above list as data frame and name of the column as text

import pandas as pd
data = {'text': ["Hello and welcome friends","Arman Ishu and Shoaib love teachi
df = pd.DataFrame(data)
df



```
# prompt: perform tf-idf and bow concatenation on above text col
import numpy as np
# Import necessary libraries
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
# Create a TfidfVectorizer object
vectorizer_tfidf = TfidfVectorizer()
# Fit and transform the data
X_tfidf = vectorizer_tfidf.fit_transform(df['text'])
# Create a CountVectorizer object
vectorizer_bow = CountVectorizer()
# Fit and transform the data
X_bow = vectorizer_bow.fit_transform(df['text'])
# Concatenate the TF-IDF and Bag of Words vectors
X_combined = np.concatenate((X_tfidf.toarray(), X_bow.toarray()), axis=1)
# Print the concatenated vector
print(X_combined)
     [[0.32274454 0.
                              0.
                                          0.54645401 0.54645401 0.
                                                                 0.54645401
       0.
                  0.
                              0.
                                          0.
                                                     0.
       1.
                  0.
                              0.
                                          1.
                                                     1.
                                                                 0.
                                          0.
                                                     0.
                  0.
      [0.30861775 0.39740155 0.
                                                                 0.39740155
                                          0.
                                                     0.
                              0.39740155 0.52253528 0.
       0.39740155 0.
                                                                 0.
       1.
                  1.
                              0.
                                          0.
                                                     0.
                                                                 1.
       1.
                  0.
                                          1.
                                                     0.
                              1.
                  0.31960436 0.42024133 0.
                                                                 0.31960436
      [0.2482013
       0.31960436 0.42024133 0.31960436 0.
                                                     0.42024133 0.
       1.
                  1.
                              1.
                                          0.
                                                     0.
                                                                 1.
                                                                            ]]
       1.
                  1.
                              1.
                                          0.
                                                                 0.
                                                     1.
```

prompt: convert above X_combined to dataframe

import pandas as pd
import numpy as np
df_combined = pd.DataFrame(X_combined, columns=np.concatenate((vectorizer_tfid1))
df_combined

→		and	arman	cricket	friends	hello	ishu	love	play	sh		
	0	0.322745	0.000000	0.000000	0.546454	0.546454	0.000000	0.000000	0.000000	0.0		
	1	0.308618	0.397402	0.000000	0.000000	0.000000	0.397402	0.397402	0.000000	0.39		
	2	0.248201	0.319604	0.420241	0.000000	0.000000	0.319604	0.319604	0.420241	0.3		
	3 rows × 24 columns											

prompt: print the list of positive words classified

positive_words = [word for word in one_positive_state.split() if word not in st positive_words)

['opportunistic', 'alliance', 'purpose', 'defeat', 'bjp', 'going', 'achieve Start coding or generate with AI.

Double-click (or enter) to edit