

```
#Financial Market Sentiment Analysis
```

Importing the libraries

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

Importing the dataset

```
df =
pd.read_csv('https://raw.githubusercontent.com/YBI-Foundation/Dataset/main/Financial%20Market%20News.csv',encoding="ISO-8859-1")
```

Get Information of Dataframe

```
df.info()    #gives column name, count, not null category, D-type(data type)
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 4101 entries, 0 to 4100
```

```
Data columns (total 27 columns):
```

#	Column	Non-Null Count	Dtype
0	Date	4101 non-null	object
1	Label	4101 non-null	int64
2	News 1	4101 non-null	object
3	News 2	4101 non-null	object
4	News 3	4101 non-null	object
5	News 4	4101 non-null	object
6	News 5	4101 non-null	object
7	News 6	4101 non-null	object
8	News 7	4101 non-null	object
9	News 8	4101 non-null	object
10	News 9	4101 non-null	object
11	News 10	4101 non-null	object
12	News 11	4101 non-null	object
13	News 12	4101 non-null	object
14	News 13	4101 non-null	object
15	News 14	4101 non-null	object
16	News 15	4101 non-null	object
17	News 16	4101 non-null	object
18	News 17	4101 non-null	object
19	News 18	4101 non-null	object
20	News 19	4101 non-null	object
21	News 20	4101 non-null	object
22	News 21	4101 non-null	object
23	News 22	4101 non-null	object

```

24 News 23 4100 non-null object
25 News 24 4098 non-null object
26 News 25 4098 non-null object
dtypes: int64(1), object(26)
memory usage: 865.2+ KB

```

`df.describe()` *#gives the linear relation of each column with another column*

```

count 4101.000000
mean   0.528164
std    0.499267
min    0.000000
25%    0.000000
50%    1.000000
75%    1.000000
max    1.000000

```

`df.head()`

	Date	Label	News
1 \			
0	01-01-2010	0	McIlroy's men catch cold from Gudjonsson
1	02-01-2010	0	Warning from history points to crash
2	03-01-2010	0	Comment: Why Israel's peaceniks feel betrayed
3	04-01-2010	1	£750,000-a-goal Weah aims parting shot
4	05-01-2010	1	Leeds arrive in Turkey to the silence of the fans

	News 2 \
0	Obituary: Brian Walsh
1	Investors flee to dollar haven
2	Court deals blow to seizure of drug assets
3	Newcastle pay for Fletcher years
4	One woman's vision offers loan lifeline

	News 3 \
0	Workplace blues leave employers in the red
1	Banks and tobacco in favour
2	An ideal target for spooks
3	Brown sent to the stands for Scotland qualifier
4	Working Lives: How world leaders worked

	News 4 \
0	Classical review: Rattle
1	Review: Llama Farmers

2 World steps between two sides intent on war
3 Tourists wary of breaking new ground
4 Working Lives: Tricks of the trade

News 5 \

0 Dance review: Merce Cunningham
1 War jitters lead to sell-off
2 What the region's papers say
3 Canary Wharf climbs into the FTSE 100
4 Working Lives: six-hour days, long lunches and...

News 6 \

0 Genetic tests to be used in setting premiums
1 Your not-so-secret history
2 Comment: Fear and rage in Palestine
3 Review: Bill Bailey
4 Pop review: We Love UK

News 7 \

0 Opera review: La Bohème
1 Review: The Northern Sinfonia
2 Poverty and resentment fuels Palestinian fury
3 Review: Classical
4 World music review: Marisa Monte

News 8 ... \

0 Pop review: Britney Spears ...
1 Review: Hysteria ...
2 Republican feud fear as dissident is killed ...
3 Review: New Contemporaries 2000 ...
4 Art review: Hollingsworth/Heyer ...

News 16

News 17 \

0 Finland 0 - 0 England Healy a
marked man
1 Why Wenger will stick to his Gunners Out of luck England hit rock
bottom
2 FTSE goes upwardly mobile At this price? BP
Amoco
3 More cash on way for counties Cairns carries Kiwis to
victory
4 Duisenberg in double trouble Pru to cut pension
charges

News 18 \

0 Happy birthday Harpers & Queen
1 Wilkinson out of his depth
2 Go fish
3 Year of Blanchflower's flourish when Spurs sto...

4 Art review: Paul Graham

News 19 \

0 Win unlimited access to the Raindance film fes...
1 Kinsella sparks Irish power play
2 Bosnian Serb blows himself up to evade law
3 New direct approach brings only pay-per-blues
4 Shearer shot sparks Boro humiliation

News 20 \

0 Labour pledges £800m to bridge north-south divide
1 Brown banished as Scots rebound
2 Orange float delayed to 2001
3 Third Division round-up
4 Ridsdale's lingering fears as Leeds revisit Tu...

News 21 \

0 Wales: Lib-Lab pact firm despite resignation
1 Battling Wales cling to lifeline
2 Angry factory workers root out fear, favours a...
3 Second Division round-up
4 Champions League: Rangers v Galatasaray

News 22 \

0 Donald Dewar
1 Ehiogu close to sealing Boro move
2 Smith defied advice on dome payout
3 First Division round-up
4 Champions League: Lazio v Arsenal

News 23 \

0 Regenerating homes regenerates well-being in ...
1 Man-to-man marking
2 Xerox takes the axe to jobs
3 McLean ends his career with a punch
4 Lazio 1 - 1 Arsenal

News 24 \

0 Win £100 worth of underwear
1 Match stats
2 Comment: Refugees in Britain
3 Heskey grabs triple crown
4 England in Pakistan

News 25

0 TV guide: Random views
1 French referee at centre of storm is no strang...
2 Maverick who sparked the new intifada
3 Weah on his way as City march on

4 England given olive-branch reception

[5 rows x 27 columns]

`df.isnull().sum()` *#(df.isna()).sum() gives same result)*
#gives the sum of all null values columns-wise

Date	0
Label	0
News 1	0
News 2	0
News 3	0
News 4	0
News 5	0
News 6	0
News 7	0
News 8	0
News 9	0
News 10	0
News 11	0
News 12	0
News 13	0
News 14	0
News 15	0
News 16	0
News 17	0
News 18	0
News 19	0
News 20	0
News 21	0
News 22	0
News 23	1
News 24	3
News 25	3

dtype: int64

`df.nunique()` *#gives total no. of unique entries*

Date	4101
Label	2
News 1	4074
News 2	4082
News 3	4081
News 4	4087
News 5	4070
News 6	4075
News 7	4078
News 8	4080
News 9	4085
News 10	4077
News 11	4074

```
News 12    4080
News 13    4082
News 14    4081
News 15    4083
News 16    4083
News 17    4077
News 18    4082
News 19    4078
News 20    4078
News 21    4077
News 22    4078
News 23    4072
News 24    4078
News 25    4066
dtype: int64
```

```
df.columns #give column names in the dataframe
```

```
Index(['Date', 'Label', 'News 1', 'News 2', 'News 3', 'News 4', 'News 5',
      'News 6', 'News 7', 'News 8', 'News 9', 'News 10', 'News 11',
      'News 12', 'News 13', 'News 14', 'News 15', 'News 16', 'News 17', 'News 18',
      'News 19', 'News 20', 'News 21', 'News 22', 'News 23', 'News 24',
      'News 25'],
      dtype='object')
```

```
df.shape
```

```
(4101, 27)
```

```
#Get Feature Selection
```

```
' '.join(str(x) for x in df.iloc[1,2:27])
```

```
{"type": "string"}
```

```
df.index
```

```
RangeIndex(start=0, stop=4101, step=1)
```

```
len(df.index)
```

```
4101
```

```
news=[]
```

```
for row in range(0, len(df.index)):
    news.append(' '.join(str(x) for x in df.iloc[row,2:27]))
```

```
type(news)
```

```
list
news[0]
{"type": "string"}
X=news
type(news)
list
```

Get Feature Text conversion to Bag of Words

```
from sklearn.feature_extraction.text import CountVectorizer
cv=CountVectorizer(lowercase=True, ngram_range=(1,1))
X=cv.fit_transform(X)
X.shape
(4101, 48527)
X
<4101x48527 sparse matrix of type '<class 'numpy.int64'>'
  with 872946 stored elements in Compressed Sparse Row format>
y=df['Label']
y.shape
(4101,)
```

Splitting the dataset into the Training set and Test set

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.3,
                                                    stratify=y, random_state = 1)
X_train.shape, X_test.shape, y_train.shape, y_test.shape
((2870, 48527), (1231, 48527), (2870,), (1231,))
```

Model

```
from sklearn.ensemble import RandomForestClassifier
rf=RandomForestClassifier()
rf.fit(X_train,y_train)
RandomForestClassifier()
```

Model Prediction

```
y_pred=rf.predict(X_test)
```

```
y_pred.shape
```

```
(1231,)
```

Get Model Evaluation

```
from sklearn.metrics import accuracy_score, confusion_matrix,  
classification_report
```

```
accuracy_score(y_test,y_pred)
```

```
0.5174654752233956
```

```
confusion_matrix(y_test,y_pred)
```

```
array([[168, 413],  
       [181, 469]])
```

```
print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
0	0.48	0.29	0.36	581
1	0.53	0.72	0.61	650
accuracy			0.52	1231
macro avg	0.51	0.51	0.49	1231
weighted avg	0.51	0.52	0.49	1231

```
rf.predict_proba(X_test)
```

```
array([[0.58, 0.42],  
       [0.49, 0.51],  
       [0.6 , 0.4 ],  
       ...,  
       [0.6 , 0.4 ],  
       [0.5 , 0.5 ],  
       [0.57, 0.43]])
```