```
import numpy as np
import pandas as pd
import sklearn
print("numpy",np.__version__)
print("pandas",pd.__version__)
print("sklearn", sklearn.__version__)
numpy 1.26.4
pandas 2.2.2
sklearn 1.5.1
df = pd.read csv('emails.csv')
df.sample(5)
                                                   text
                                                         spam
      Subject: re : enron offsite hi steve : liste...
                                                             0
3160
     Subject: customer profiling meeting bob shult...
                                                             0
3744 Subject: tiger team info vince , here is the...
                                                             0
2916 Subject: re : meeting re : wharton strategy j...
                                                             0
2617 Subject: non - firm power curve building hi v...
                                                             0
df.shape
(5728, 2)
# 1. Data Cleaning
# 2. EDA
# 3. Text preprocessing
# 4. Model Building
# 5. Evaluation
# 6. Improvement
# 7. Website
# 8. Deploy
```

## 1. Data Cleaning

```
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5728 entries, 0 to 5727
Data columns (total 2 columns):
    Column Non-Null Count Dtype
#
- - -
0
     text
             5728 non-null
                             object
             5728 non-null
                             int64
1
     spam
dtypes: int64(1), object(1)
memory usage: 89.6+ KB
```

```
# removal of columns not needed as their is already two columns and
there is no null values present
# renaming the cols
df.rename (columns = {'text': 'messages','spam': 'target'},
inplace=True)
df.sample(5)
                                               messages target
      Subject: re : test vince : candice ' s conta...
4030
860
      Subject: perfect visual solution for your busi...
                                                              1
233
                                                              1
      Subject: in the heart of your business! corp...
2301 Subject: organizational announcement in case ...
                                                              0
4117 Subject: larry thorpe hi vince, been meanin...
# In target we already have mapping 1-spam and 0-Not Spam so no need
to encode this
#missing values
df.isnull().sum()
            0
messages
target
dtype: int64
#check for duplicate values
df.duplicated().sum()
33
#remove duplicates
df = df.drop duplicates(keep='first')
df.duplicated().sum()
0
df.shape
(5695, 2)
```

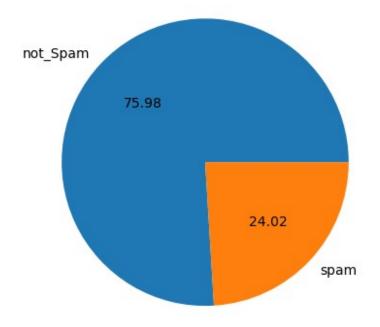
## 2. EDA

```
3 Subject: 4 color printing special request add... 1
4 Subject: do not have money , get software cds ... 1

df['target'].value_counts()

target
0    4327
1    1368
Name: count, dtype: int64

import matplotlib.pyplot as plt
plt.pie(df['target'].value_counts(),labels=['not_Spam', 'spam'], autopct="%0.2f")
plt.show()
```



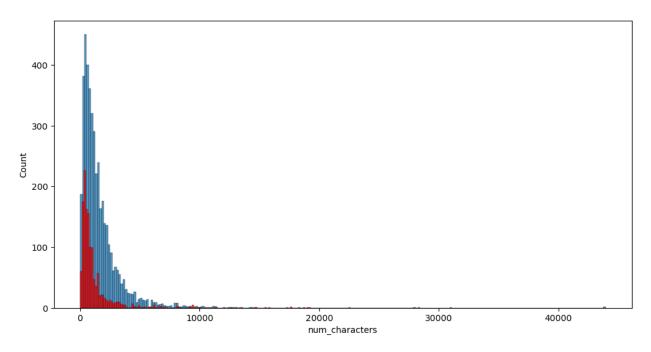
```
#Data is slightly imbalanced
import nltk
!pip install nltk

Requirement already satisfied: nltk in c:\users\sevan\anaconda3\lib\
site-packages (3.9.1)
Requirement already satisfied: click in c:\users\sevan\anaconda3\lib\
site-packages (from nltk) (8.1.7)
Requirement already satisfied: joblib in c:\users\sevan\anaconda3\lib\
site-packages (from nltk) (1.4.2)
```

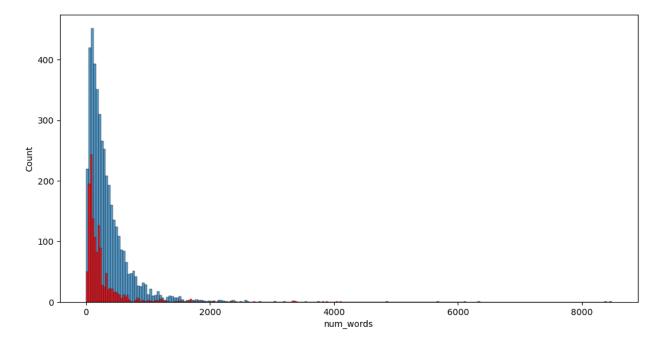
```
Requirement already satisfied: regex>=2021.8.3 in c:\users\sevan\
anaconda3\lib\site-packages (from nltk) (2024.9.11)
Requirement already satisfied: tqdm in c:\users\sevan\anaconda3\lib\
site-packages (from nltk) (4.66.5)
Requirement already satisfied: colorama in c:\users\sevan\anaconda3\
lib\site-packages (from click->nltk) (0.4.6)
nltk.download('punkt')
[nltk data] Downloading package punkt to
[nltk data]
               C:\Users\sevan\AppData\Roaming\nltk data...
              Package punkt is already up-to-date!
[nltk data]
True
df['num characters']= df['messages'].apply(len)
df.head()
                                            messages target
num characters
O Subject: naturally irresistible your corporate...
1 Subject: the stock trading gunslinger fanny i...
598
2 Subject: unbelievable new homes made easy im ...
448
3 Subject: 4 color printing special request add...
500
4 Subject: do not have money , get software cds ...
235
#num of words
nltk.download('punkt tab')
[nltk data] Downloading package punkt tab to
[nltk data]
               C:\Users\sevan\AppData\Roaming\nltk data...
[nltk data]
              Package punkt tab is already up-to-date!
True
df['num words'] = df['messages'].apply(lambda x:len
(nltk.word tokenize(x)))
df.head()
                                           messages target
num characters \
O Subject: naturally irresistible your corporate...
1484
1 Subject: the stock trading gunslinger fanny i...
598
```

```
2 Subject: unbelievable new homes made easy im ...
448
3 Subject: 4 color printing special request add...
                                                            1
500
4 Subject: do not have money , get software cds ...
235
   num words
0
         325
1
          90
2
          88
3
          99
          53
df['num sentences'] = df['messages'].apply(lambda
x:len(nltk.sent tokenize(x)))
df.head()
                                             messages target
num characters \
O Subject: naturally irresistible your corporate...
1484
1 Subject: the stock trading gunslinger fanny i...
                                                            1
598
2 Subject: unbelievable new homes made easy im ...
                                                            1
448
3 Subject: 4 color printing special request add...
500
4 Subject: do not have money , get software cds ...
235
   num words
              num sentences
0
         325
                         11
                          1
1
          90
2
          88
                          4
3
                          5
          99
                          9
          53
df[['num characters','num words','num sentences']].describe()
                                     num_sentences
       num characters
                         num words
          5695.000000
                       5695.000000
                                       5695.000000
count
          1558.067076
                        328,214047
mean
                                         19.462511
          2047.078711
                        419.654234
                                         35.981993
std
            13.000000
                          3.000000
                                          1.000000
min
25%
           508,500000
                        102,000000
                                          7,000000
                                         12,000000
50%
           979.000000
                        211.000000
75%
          1893.000000
                        403,000000
                                         22,000000
         43952.000000
                       8479.000000
                                       1565,000000
max
```

```
# for ham messages separately
df[df['target'] == 0 ]
[['num_characters','num_words','num_sentences']].describe()
       num characters
                                     num sentences
                          num words
          4327.000000
                                       4327,000000
                       4327.000000
count
          1634.200139
                         347.283799
                                         19.838225
mean
          1965.016383
                         407.227757
                                         38.124564
std
min
            13.000000
                           3.000000
                                          1.000000
25%
           577.500000
                         120.000000
                                          7.000000
50%
          1122.000000
                         240.000000
                                         13.000000
75%
          2037.500000
                         440.500000
                                         22,000000
         43952.000000
                       8479.000000
                                       1565.000000
max
# for spam messages separately
df[df['target'] == 1 ]
[['num_characters','num_words','num_sentences']].describe()
       num characters
                          num words
                                     num sentences
count
          1368.000000
                       1368.000000
                                       1368.000000
          1317.257310
                         267.896199
                                         18.274123
mean
          2271.372893
                                         28.130434
std
                         451.623124
            18.000000
                           5.000000
                                          1.000000
min
           401.500000
25%
                          80,000000
                                          6.000000
50%
           693,500000
                         141.000000
                                         11.000000
75%
          1250.250000
                         252.000000
                                         18.000000
         28432.000000
                       6131.000000
                                        438.000000
max
import seaborn as sns
plt.figure(figsize=(12, 6))
sns.histplot(df[df['target']== 0]['num characters'])
sns.histplot(df[df['target']== 1]['num characters'], color='red')
<Axes: xlabel='num characters', ylabel='Count'>
```

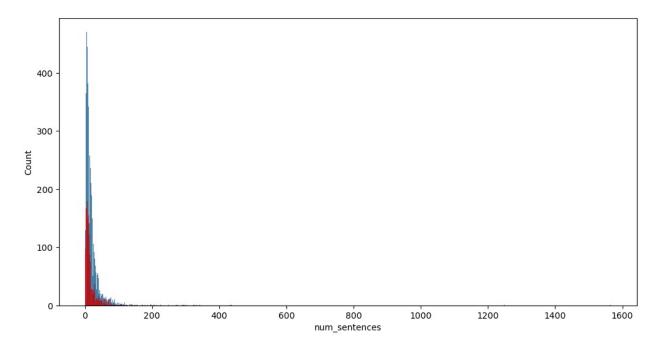


```
plt.figure(figsize=(12, 6))
sns.histplot(df[df['target']== 0]['num_words'])
sns.histplot(df[df['target']== 1]['num_words'], color='red')
<Axes: xlabel='num_words', ylabel='Count'>
```

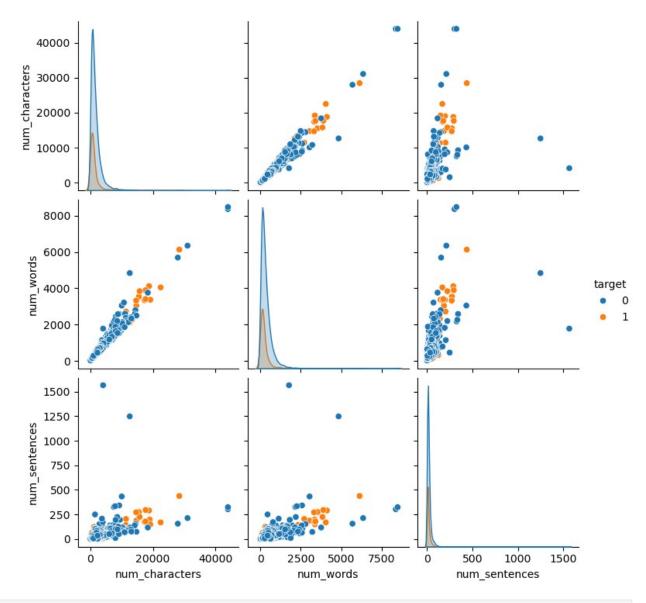


```
plt.figure(figsize=(12, 6))
sns.histplot(df[df['target']== 0]['num_sentences'])
sns.histplot(df[df['target']== 1]['num_sentences'], color='red')
```

## <Axes: xlabel='num\_sentences', ylabel='Count'>



sns.pairplot(df, hue='target')
<seaborn.axisgrid.PairGrid at 0x25f9a10a1e0>



sns.heatmap(df.select\_dtypes(include=['number']).corr(), annot=True)
<Axes: >



## 3. Data Preprocessing

lowercase Tokenization Removing special characters Removing stop words and Punctuations Stemming

```
import string
import nltk
from nltk.corpus import stopwords
stopwords.words('english')
['a',
 'about',
'above',
 'after',
 'again',
 'against',
 'ain',
 'all',
 'am',
 'an',
 'and',
 'any',
 'are',
```

```
'aren',
"aren't",
'as',
'at',
'be',
'because',
'been',
'before',
'being',
'below',
'between',
'both',
'but',
'by',
'can',
'couldn',
"couldn't",
'd',
'did',
'didn',
"didn't",
'do',
'does',
'doesn',
"doesn't",
'doing',
'don',
"don't",
'down',
'during',
'each',
'few',
'for',
'from',
'further',
'had',
'hadn',
"hadn't",
'has',
'hasn',
"hasn't",
'have',
'haven',
"haven't",
'having',
'he',
"he'd",
"he'll",
'her',
```

```
'here',
'hers',
'herself',
"he's",
'him',
'himself',
'his',
'how',
'i',
"i'd",
'if',
"i'll",
"i'm",
'in',
'into',
'is',
'isn',
"isn't",
'it',
"it'd",
"it'll",
"it's",
'its',
'itself',
"i've",
'just',
'11',
'm',
'ma',
'me',
'mightn',
"mightn't",
'more',
'most',
'mustn',
"mustn't",
'my',
'myself',
'néedn',
"needn't",
'no',
'nor',
'not',
'now',
'o',
'of',
'off',
'on',
'once',
```

```
'only',
'or',
'other',
'our',
'ours',
'ourselves',
'out',
'over',
'own',
're',
's',
'same',
'shan',
"shan't",
'she',
"she'd",
"she'll<sup>"</sup>,
"she's",
'should',
'shouldn',
"shouldn't",
"should've",
'so',
'some',
'such',
't',
'than',
'that',
"that'll",
'the',
'their',
'theirs',
'them',
'themselves',
'then',
'there',
'these',
'they',
"they'd",
"they'll",
"they're",
"they've",
'this',
'those',
'through',
'to',
'too',
'under',
'until',
```

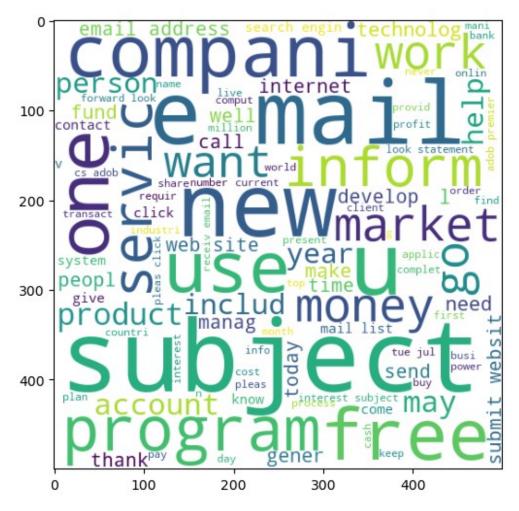
```
'up',
 've',
 'very',
 'was',
 'wasn',
 "wasn't",
 'we',
 "we'd",
 "we'll",
 "we're",
 'were',
 'weren',
 "weren't",
 "we've",
 'what',
 'when',
 'where',
 'which',
 'while',
 'who',
 'whom',
 'why',
'will',
 'with',
 'won',
 "won't",
 'wouldn',
 "wouldn't",
 'y',
 'you',
 "you'd",
 "you'll",
 'your',
 "you're",
 'yours',
 'yourself',
 'yourselves',
 "you've"]
from nltk.stem.porter import PorterStemmer
ps = PorterStemmer()
def transform messages(messages):
    messages = messages.lower()
    messages = nltk.word tokenize(messages)
    y = []
    for i in messages:
        if i.isalnum():
             y.append(i)
    text = y[:]
```

```
v.clear()
   for i in text:
       if i not in stopwords.words('english') and i not in
string.punctuation:
           y.append(i)
   text = v[:]
   y.clear()
   for i in text:
       y.append(ps.stem(i))
    return " ".join(y)
transform messages(" Subject: naturally irresistible your corporate
identity It is really hard to recollect a company: the market is
full of suggestions and the information isoverwhelming; but a good
catchy logo , stylish statlonery and outstanding website will make
the task much easier . we do not promise that having ordered a iogo
your company will automatically become a world leader : it isquite
ciear that without good products , effective business organization
and practicable aim it will be hotat nowadays market; but we do
promise that your marketing efforts will become much more effective .
here is the list of clear benefits : creativeness : hand - made ,
original logos, specially done to reflect your distinctive company
image . convenience : logo and stationery are provided in all formats
; easy - to - use content management system letsyou change your
website content and even its structure . promptness : you will see
logo drafts within three business days . affordability : your
marketing break - through shouldn ' t make gaps in your budget . 100 %
satisfaction quaranteed : we provide unlimited amount of changes with
no extra fees for you to be surethat you will love the result of this
collaboration . have a look at our portfolio _ _ _ _ _ _ _
```

'subject natur irresist corpor ident lt realli hard recollect compani market full suggest inform isoverwhelming good catchi logo stylish statloneri outstand websit make task much easier promis having order iogo compani automaticaili becom world ieader isguit ciear without good product effect busi organ practic aim hotat nowaday market promis market effort becom much effect list clear benefit creativ hand made origin logo special done reflect distinct compani imag conveni logo stationeri provid format easi use content manag system letsyou chang websit content even structur prompt see logo draft within three busi day afford market break make gap budget 100 satisfact guarante provid unlimit amount chang extra fee surethat love result collabor look portfolio interest'

```
df['messages'][0]
"Subject: naturally irresistible your corporate identity lt is really
hard to recollect a company : the market is full of suggestions and
the information isoverwhelming; but a good catchy logo, stylish
statlonery and outstanding website will make the task much easier .
we do not promise that having ordered a iogo your company will
automatically become a world leader: it isguite clear that without
good products, effective business organization and practicable aim it
will be hotat nowadays market; but we do promise that your marketing
efforts will become much more effective . here is the list of clear
benefits : creativeness : hand - made , original logos , specially
done to reflect your distinctive company image . convenience : logo
and stationery are provided in all formats; easy - to - use content
management system letsyou change your website content and even its
structure . promptness : you will see logo drafts within three
business days . affordability : your marketing break - through
shouldn ' t make gaps in your budget . 100 % satisfaction
guaranteed : we provide unlimited amount of changes with no extra fees
for you to be surethat you will love the result of this collaboration
. have a look at our portfolio _ _ _ _ _ _ _ _ _ _
df['transformed messages'] = df['messages'].apply(transform messages)
df.head()
                                         messages target
num characters \
O Subject: naturally irresistible your corporate...
1484
1 Subject: the stock trading gunslinger fanny i...
                                                        1
598
2 Subject: unbelievable new homes made easy im ...
448
3 Subject: 4 color printing special request add...
500
4 Subject: do not have money , get software cds ...
235
   num words num sentences
transformed messages
        325
                          subject natur irresist corpor ident lt
                        11
realli ...
1
         90
                           subject stock trade gunsling fanni merril
muzo...
         88
                        4 subject unbeliev new home made easi im
want sh...
```

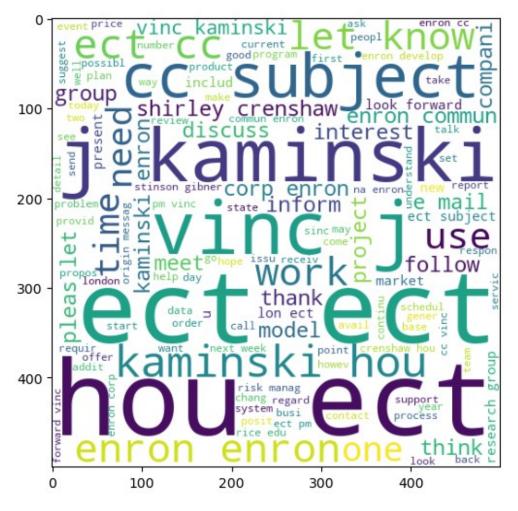
```
99
                            subject 4 color print special request
addit in...
          53
                             subject money get softwar cd softwar
compat gr...
!pip install wordcloud
Requirement already satisfied: wordcloud in c:\users\sevan\anaconda3\
lib\site-packages (1.9.4)
Requirement already satisfied: numpy>=1.6.1 in c:\users\sevan\
anaconda3\lib\site-packages (from wordcloud) (1.26.4)
Requirement already satisfied: pillow in c:\users\sevan\anaconda3\lib\
site-packages (from wordcloud) (10.4.0)
Requirement already satisfied: matplotlib in c:\users\sevan\anaconda3\
lib\site-packages (from wordcloud) (3.9.2)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\sevan\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (1.2.0)
Requirement already satisfied: cycler>=0.10 in c:\users\sevan\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\sevan\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (4.51.0)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\sevan\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (1.4.4)
Requirement already satisfied: packaging>=20.0 in c:\users\sevan\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (24.1)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\sevan\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (3.1.2)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\sevan\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (2.9.0.post0)
Requirement already satisfied: six>=1.5 in c:\users\sevan\anaconda3\
lib\site-packages (from python-dateutil>=2.7->matplotlib->wordcloud)
(1.16.0)
from wordcloud import WordCloud
wc =
WordCloud(width=500, height=500, min font size=10, background color='whit
e')
spam wc = wc.generate(df[df['target'] == 1]
['transformed_messages'].str.cat(sep=" "))
# I have to change this to messages -> Transformed messages
plt.figure(figsize = (15,6))
plt.imshow(spam wc)
<matplotlib.image.AxesImage at 0x25f9ad77800>
```



```
ham_wc = wc.generate(df[df['target'] == 0]
['transformed_messages'].str.cat(sep=" "))

plt.figure(figsize = (15,6))
plt.imshow(ham_wc)

<matplotlib.image.AxesImage at 0x25f9da62870>
```



```
df.head()
                                            messages target
num characters
O Subject: naturally irresistible your corporate...
1484
1 Subject: the stock trading gunslinger fanny i...
598
2 Subject: unbelievable new homes made easy im ...
448
3 Subject: 4 color printing special request add...
500
4 Subject: do not have money , get software cds ...
235
   num words num sentences
transformed messages
         325
                             subject natur irresist corpor ident lt
                         11
realli ...
                             subject stock trade gunsling fanni merril
          90
```

```
muzo...
          88
                              subject unbeliev new home made easi im
2
want sh...
          99
                           5
                               subject 4 color print special request
addit in...
          53
                           9
                              subject money get softwar cd softwar
compat gr...
spam corpus = []
for msg in df[df['target']==1]['transformed_messages'].tolist():
    for word in msg.split():
        spam corpus.append(word)
spam corpus
['subject',
 'natur',
 'irresist',
 'corpor',
 'ident',
 'lt',
 'realli',
 'hard',
 'recollect',
 'compani',
 'market',
 'full',
 'suggest',
 'inform',
 'isoverwhelming',
 'good',
 'catchi',
 'logo',
 'stylish',
 'statloneri',
 'outstand',
 'websit',
 'make',
 'task',
 'much',
 'easier',
 'promis',
 'havinq',
 'order',
 'iogo',
 'compani',
 'automaticaili',
 'becom',
 'world',
 'ieader',
```

```
'isguit',
'ciear',
'without',
'good',
'product',
'effect',
'busi',
'organ',
'practic',
'aim',
'hotat',
'nowaday',
'market',
'promis',
'market',
'effort',
'becom',
'much',
'effect',
'list',
'clear',
'benefit',
'creativ',
'hand',
'made',
'origin',
'logo',
'special',
'done',
'reflect',
'distinct',
'compani',
'imag',
'conveni',
'logo',
'stationeri',
'provid',
'format',
'easi',
'use',
'content',
'manag',
'system',
'letsyou',
'chang',
'websit',
'content',
'even',
'structur',
```

```
'prompt',
'see',
'logo<sup>'</sup>,
'draft',
'within',
'three',
'busi',
'day',
'afford',
'market',
'break',
'make',
'gap',
'budget',
'100',
'satisfact',
'guarante',
'provid',
'unlimit',
'amount',
'chang',
'extra',
'fee',
'surethat',
'love',
'result',
'collabor',
'look',
'portfolio',
'interest',
'subject',
'stock',
'trade',
'gunsling',
'fanni',
'merril',
'muzo',
'colza',
'attaind'
'penultim',
'like',
'esmark',
'perspicu',
'rambl',
'segovia',
'group',
'tri',
'slung',
'kansa',
```

```
'tanzania',
'ye',
'chameleon',
'continu',
'clothesman',
'libretto',
'chesapeak',
'tight',
'waterway',
'herald',
'hawthorn',
'like',
'chisel',
'morristown',
'superior',
'deoxyribonucl',
'clockwork',
'tri',
'hall',
'incred',
'mcdougal',
'ye',
'hepburn',
'einsteinian',
'earmark',
'sapl',
'boar',
'duan',
'plain',
'palfrey',
'inflex',
'like',
'huzzah',
'pepperoni',
'bedtim',
'nameabl<sup>'</sup>,
'attir',
'tri',
'edt',
'chronographi',
'optima',
'ye',
'pirogu',
'diffus',
'albeit',
'subject'
'unbeliev',
'new',
'home',
```

```
'made',
'easi',
'im',
'want',
'show',
'homeown',
'pre',
'approv',
'454',
'169',
'home',
'loan',
'3',
'72',
'fix',
'rate',
'offer',
'extend<sup>'</sup>,
'uncondit',
'credit',
'way',
'factor',
'take',
'advantag',
'limit',
'time',
'opportun',
'ask',
'visit',
'websit'
'complet',
'1',
'minut',
'post',
'approv',
'form',
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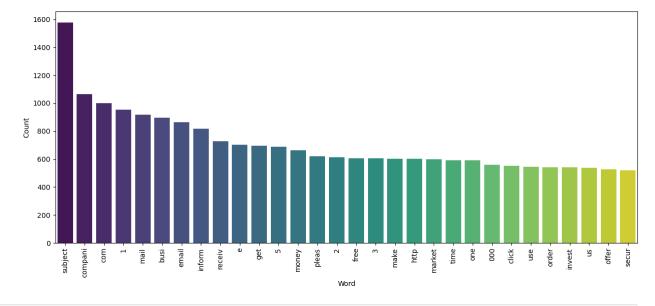
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len(spam corpus)
174630
# from collections import Counter
# sns.barplot(pd.DataFrame(Counter(spam corpus).most common(30))
[0],pd.DataFrame(Counter(spam corpus).most common(30))[1])
# plt.xticks(rotation='vertical')
from collections import Counter
# Convert Counter to DataFrame with named columns
df spam = pd.DataFrame(Counter(spam corpus).most common(30),
columns=['Word', 'Count'])
# Create the barplot with different colors
```

```
plt.figure(figsize=(15, 6))
sns.barplot(x=df_spam['Word'], y=df_spam['Count'], palette="viridis")
# You can try "magma", "coolwarm", etc.
# Rotate x-axis labels for better readability
plt.xticks(rotation='vertical')
plt.show()

C:\Users\sevan\AppData\Local\Temp\ipykernel_2808\3513459698.py:11:
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=df_spam['Word'], y=df_spam['Count'], palette="viridis") # You can try "magma", "coolwarm", etc.
```



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'disast',
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'sop',
'dpc',
'buy',
'naphtha',
'ioc',
'sanjay',
'jog',
'thursday',
'jan',
'18',
```

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'reduc',
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'deficit',
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'tax',
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'dabhol',
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'dpc',
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'naphtha',
'state',
'run',
'indian',
'oil',
'corpor',
'ioc',
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'told',
'financi',
'express',
'wednesday',
'dpc',
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'pay',
'4',
'per',
'cent',
'sale',
'tax',
'govern',
'way',
'back',
'1995',
'modifi',
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'tax',
'rate',
'4',
'per',
'cent',
'discourag',
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'naphtha',
'gujarat',
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'compani',
'oper',
'maharashtra',
'decis',
'taken',
'view',
'encourag',
'electr',
'compani',
'procur',
'naphtha',
'reduc',
'rate',
'within',
'state',
'govern',
'sourc',
'ad',
'sourc',
'said',
'compani',
'pay',
'nearli',
'15',
'3',
'per',
```

```
'cent',
'sale',
'tax',
'naphtha',
'procur',
'gujarat',
'howev',
'follow',
'represent',
'govern',
'slash',
'sale',
'tax',
'rate',
'4',
'per',
'cent<sup>'</sup>,
'state',
'financ',
'depart',
'opinion',
'would',
'present',
'state',
'cabinet',
'shortli',
'order',
'take',
'final',
'decis',
'deserv',
'special<sup>'</sup>,
'signific',
'especi',
'state',
'energi',
'depart',
'loss',
'make',
'maharashtra',
'state',
'electr',
'board',
'mseb',
'wholeheartedli',
'support',
'dpc',
'caus',
'recommend',
```

```
'sale',
'tax',
'waiver',
'dpc',
'ask',
'union',
'ministri',
'oil',
'petroleum',
'procur',
'naphtha',
'within',
'countri',
'view',
'excess',
'avail',
'present',
'state',
'govern',
'mseb',
'made',
'clear',
'would',
'left',
'altern',
'pass',
'addit',
'burden',
'mseb',
'would',
'ultim',
'pass',
'consum',
'dpc',
'also',
'told',
'state',
'govern',
'paid',
'sale',
'tax',
'procur',
'naphtha',
'glencor',
'calend',
'year',
'2000',
'sourc',
'state',
```

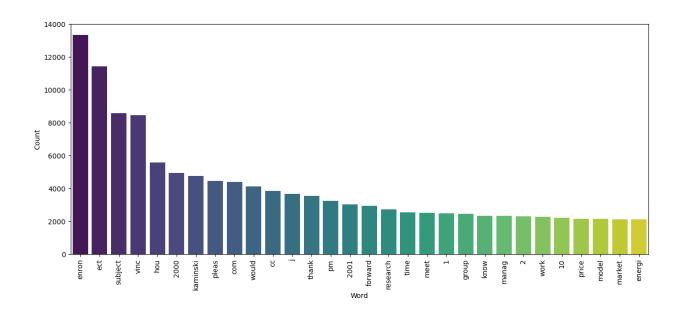
```
'energi',
'depart',
'mseb',
'stress',
'need',
'waiver',
'express',
'inabl',
'bear',
'addit',
'burden',
'suggest',
'state',
'reciproc',
'offer',
'sale',
'tax',
'exempt',
'dpc',
'ioc',
'behest',
'centr',
'tri',
'match',
'intern',
'land',
'price',
'naphtha',
'recent',
'sign',
'memorandum',
'agreement',
'dpc',
'state',
'financ',
'depart',
'stick',
'view',
'may',
'hurt',
'state',
'whole',
'sourc',
'state',
'energi',
'depart',
'mseb',
'said',
'dpc',
```

```
'procur',
'naphtha',
'rs<sup>'</sup>,
'050',
'per',
'ton',
'ioc',
'calend',
'year',
'2001',
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'price',
'quot',
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'price',
'compris',
'175',
'per',
'tonn',
'free',
'board',
'fob',
'21',
'8',
'per',
'cent',
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'duti',
'5',
'4',
'per',
'cent',
'sale',
'tax',
'18',
'87',
'premium',
'dpc',
'senior',
'vice',
'presid',
'mukesh',
```

```
'tyagi',
'reiter',
'compani',
'alreadi',
'made',
'appeal',
'state',
'govern',
'sale',
'tax',
'waiver',
'naphtha',
'larger',
'interest',
'consum',
'sale',
'tax',
'pass',
'mseb',
'bear',
'addit',
'burden',
'pass',
'consum',
'ad',
'centr',
'yet',
'receiv',
'propos',
'enron',
'thursday',
'jan',
'ĺ8',
'2001',
'centr',
'wednesday',
'said',
'recevi',
'propos',
'maharashtra',
'govern',
'seek',
'help',
'solv',
'tangl',
'enron',
'promot',
'dhabol',
'power',
```

```
'project',
'relat',
'cost',
'surplu',
'power',
'ask',
'report',
'maharashtra',
'govern',
'send',
'propos',
'centr',
'buy',
'surplu',
'power',
'dhabol',
'power',
'compani',
'power',
'trade',
'corpor',
'power',
'minist',
'suresh',
'prabhu',
'said',
'receiv',
'propos',
'care',
'watch',
'situat',
'await',
'concret',
'propos',
'interven',
'matter',
'parbhu',
'said',
'sidelin',
'greentech',
'environ',
'excel',
'award',
'ceremoni',
'ask',
'whether',
'possibl',
'govern',
'ask',
```

```
'power',
 'trade',
 'corpor',
 'buy',
 'power'
 'dhabol',
 'power',
 'corpor',
 'prabhu',
 'repli',
 'ptc',
 . . . ]
len(ham corpus)
702991
from collections import Counter
# Convert Counter to DataFrame with named columns
df ham = pd.DataFrame(Counter(ham corpus).most common(30),
columns=['Word', 'Count'])
# Create the barplot with different colors
plt.figure(figsize=(15, 6))
sns.barplot(x=df ham['Word'], y=df ham['Count'], palette="viridis") #
You can try "magma", "coolwarm", etc.
# Rotate x-axis labels for better readability
plt.xticks(rotation='vertical')
plt.show()
C:\Users\sevan\AppData\Local\Temp\ipykernel_2808\1297375888.py:8:
FutureWarning:
Passing `palette` without assigning `hue` is deprecated and will be
removed in v0.14.0. Assign the `x` variable to `hue` and set
`legend=False` for the same effect.
  sns.barplot(x=df_ham['Word'], y=df_ham['Count'], palette="viridis")
# You can try "magma", "coolwarm", etc.
```



## 4. Model Buiding

```
# Text Vectorization
# 1. Using Bag of Words
df.head()
                                            messages target
num characters \
O Subject: naturally irresistible your corporate...
1484
1 Subject: the stock trading gunslinger fanny i...
598
2 Subject: unbelievable new homes made easy im ...
448
3 Subject: 4 color printing special request add...
500
4 Subject: do not have money , get software cds ...
   num words num sentences
transformed messages
         325
                         11
                             subject natur irresist corpor ident lt
realli ...
          90
                             subject stock trade gunsling fanni merril
1
                          1
muzo...
                             subject unbeliev new home made easi im
          88
                          4
want sh...
                             subject 4 color print special request
addit in...
                             subject money get softwar cd softwar
compat gr...
```

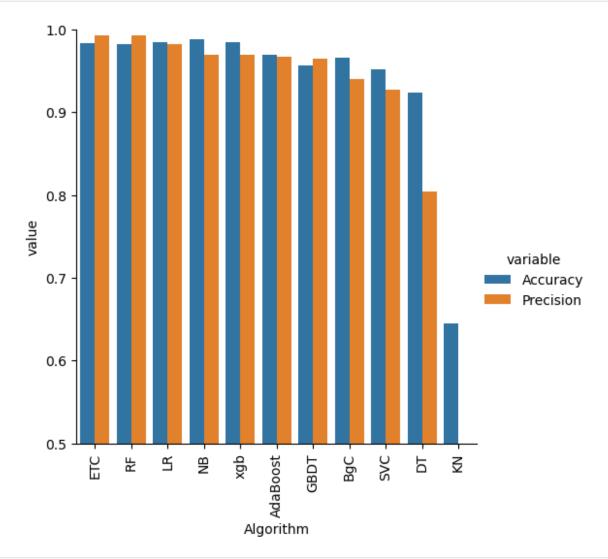
```
from sklearn.feature extraction.text import
CountVectorizer, TfidfVectorizer
cv = CountVectorizer()
tfidf = TfidfVectorizer(max features=3000)
# X = cv.fit transform(df['transformed messages']).toarray()
X = tfidf.fit transform(df['transformed messages']).toarray()
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
X = scaler.fit transform(X)
# # appending the num character col to X
\# X = np.hstack((X,df['num characters'].values.reshape(-1,1)))
X.shape
(5695, 3000)
y = df['target'].values
array([1, 1, 1, ..., 0, 0, 0], dtype=int64)
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, random state=2)
from sklearn.naive bayes import GaussianNB, MultinomialNB, BernoulliNB
from sklearn.metrics import
accuracy score, confusion matrix, precision score
qnb = GaussianNB()
mnb = MultinomialNB()
bnb = BernoulliNB()
gnb.fit(X train,y train)
y pred1 = gnb.predict(X test)
print(accuracy_score(y_test,y_pred1))
print(confusion matrix(y test,y pred1))
print (precision score(y test,y pred1))
0.9525899912203687
[[837 12]
 [ 42 24811
0.9538461538461539
mnb.fit(X train,y train)
y pred2 = mnb.predict(X test)
print(accuracy score(y test,y pred2))
```

```
print(confusion matrix(y_test,y_pred2))
print (precision score(y test,y pred2))
0.9877085162423178
[[840
      9]
[ 5 28511
0.9693877551020408
bnb.fit(X_train,y_train)
y pred3 = mnb.predict(X test)
print(accuracy score(y test,y pred3))
print(confusion_matrix(y_test,y_pred3))
print (precision_score(y_test,y_pred3))
0.9877085162423178
[[840
       91
 [ 5 285]]
0.9693877551020408
# tfidf-->MNB
!pip install xgboost
Requirement already satisfied: xgboost in c:\users\sevan\anaconda3\
lib\site-packages (3.0.0)
Requirement already satisfied: numpy in c:\users\sevan\anaconda3\lib\
site-packages (from xgboost) (1.26.4)
Requirement already satisfied: scipy in c:\users\sevan\anaconda3\lib\
site-packages (from xgboost) (1.13.1)
from sklearn.linear model import LogisticRegression
from sklearn.svm import SVC
from sklearn.naive bayes import MultinomialNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import AdaBoostClassifier
from sklearn.ensemble import BaggingClassifier
from sklearn.ensemble import ExtraTreesClassifier
from sklearn.ensemble import GradientBoostingClassifier
from xgboost import XGBClassifier
svc = SVC(kernel = 'sigmoid', gamma = 1.0)
knc = KNeighborsClassifier()
mnb = MultinomialNB()
dtc = DecisionTreeClassifier(max depth=5)
lrc =LogisticRegression(solver = 'liblinear',penalty='l1')
rfc = RandomForestClassifier(n estimators=50, random state=2)
abc = AdaBoostClassifier(n estimators=50, random state=2)
bc = BaggingClassifier(n estimators=50, random state=2)
etc = ExtraTreesClassifier(n estimators=50, random state=2)
```

```
gbdt = GradientBoostingClassifier(n estimators=50, random state=2)
xgb = XGBClassifier(n estimators=50, random state=2)
clfs = {
    'SVC' : svc,
    'KN' : knc.
    'NB' : mnb,
    'DT' : dtc,
    'LR' : lrc,
    'RF' : rfc,
    'AdaBoost' : abc,
    'BgC' : bc,
    'ETC' : etc,
    'GBDT' : gbdt,
    'xqb' : xqb
}
def train_classifier(clf,X_train,y_train,X_test,y_test):
    clf.fit(X train,y train)
    y pred = clf.predict(X test)
    accuracy = accuracy score(y test,y pred)
    precision = precision score(y test,y pred)
    return accuracy, precision
train_classifier(svc,X_train,y_train,X_test,y_test)
(0.95171202809482, 0.9272727272727272)
accuracy scores = []
precision scores = []
for name,clf in clfs.items():
    current accuracy, current precision =
train classifier(clf,X train,y train,X test,y test)
    print("For ",name)
    print("Accuracy - ",current_accuracy)
    print("Precision - ", current_precision)
    accuracy scores.append(current accuracy)
    precision_scores.append(current_precision)
For SVC
Accuracy - 0.95171202809482
Precision - 0.92727272727272
For KN
Accuracy - 0.6453028972783144
Precision - 0.41786743515850144
For NB
Accuracy - 0.9877085162423178
Precision - 0.9693877551020408
For DT
```

```
Accuracy - 0.9236172080772608
Precision - 0.8048048048048
For LR
Accuracy - 0.9841966637401229
Precision - 0.9822695035460993
For
    RF
Accuracy - 0.9824407374890255
Precision - 0.9927007299270073
C:\Users\sevan\anaconda3\Lib\site-packages\sklearn\ensemble\
weight boosting.py:527: FutureWarning: The SAMME.R algorithm (the
default) is deprecated and will be removed in 1.6. Use the SAMME
algorithm to circumvent this warning.
 warnings.warn(
For AdaBoost
Accuracy - 0.9692712906057945
Precision - 0.967032967032967
For BqC
Accuracy - 0.9657594381035997
Precision - 0.9403508771929825
For ETC
Accuracy - 0.9833187006145742
Precision - 0.99272727272727
For
    GBDT
Accuracy - 0.9561018437225637
Precision - 0.9651162790697675
For
    xgb
Accuracy - 0.9850746268656716
Precision - 0.9690721649484536
performance df = pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy':
accuracy scores,
'Precision':precision scores}).sort values('Precision',ascending=False
)
performance df
   Algorithm
             Accuracy
                       Precision
8
         ETC
             0.983319
                        0.992727
5
         RF
             0.982441
                        0.992701
4
         LR
             0.984197
                        0.982270
2
         NB
             0.987709
                        0.969388
10
        xgb
             0.985075
                        0.969072
             0.969271
                        0.967033
6
   AdaBoost
9
       GBDT
             0.956102
                        0.965116
7
        BqC
             0.965759
                        0.940351
0
        SVC
             0.951712
                        0.927273
3
         DT
             0.923617
                        0.804805
1
         KN
             0.645303
                        0.417867
```

```
performance_df1 = pd.melt(performance_df,id_vars = "Algorithm")
sns.catplot(x = 'Algorithm', y='value', hue = 'variable',
data=performance_df1,kind='bar',height=5)
plt.ylim(0.5,1.0)
plt.xticks(rotation = 'vertical')
plt.show()
```



```
#model improve
# 1. Change the max_features parameter of TFIDf

temp_df = pd.DataFrame({'Algorithm':clfs.keys(),
   'Accuracy_max_ft_3000':accuracy_scores,
   'Precision_max_ft_3000':precision_scores})

new_df = performance_df.merge(temp_df,on='Algorithm')
new_df
```

```
Algorithm Accuracy
                        Precision Accuracy max ft 3000
Precision max ft 3000
         ETC 0.983319
                         0.992727
                                                0.983319
0.992727
          RF
              0.982441
                         0.992701
                                                0.982441
0.992701
2
          LR
              0.984197
                         0.982270
                                                0.984197
0.982270
              0.987709
                         0.969388
          NB
                                                0.987709
0.969388
              0.985075
                         0.969072
                                                0.985075
         xgb
0.969072
    AdaBoost
              0.969271
                         0.967033
                                                0.969271
0.967033
        GBDT
              0.956102
                         0.965116
                                                0.956102
0.965116
7
         BgC
              0.965759
                         0.940351
                                                0.965759
0.940351
              0.951712
                         0.927273
                                                0.951712
         SVC
0.927273
          DT
              0.923617
                         0.804805
                                                0.923617
0.804805
              0.645303
                         0.417867
                                                0.645303
10
          KN
0.417867
temp_df = pd.DataFrame({'Algorithm':clfs.keys(),
'Accuracy scalling':accuracy scores, 'Precision scalling':precision sco
res})
# temp df = pd.DataFrame({'Algorithm':clfs.keys(),
'Accurary num chars':accuracy scores, 'Precision num chars':precision s
cores}).sort values('Precision num chars',ascending=False)
new df scaled = new df.merge(temp df,on='Algorithm')
new df scaled
   Algorithm
              Accuracy
                        Precision
                                    Accuracy_max_ft_3000
0
              0.983319
                         0.992727
                                                0.983319
         ETC
1
          RF
              0.982441
                         0.992701
                                                0.982441
2
          LR
              0.984197
                         0.982270
                                                0.984197
3
          NB
              0.987709
                         0.969388
                                                0.987709
4
         xab
              0.985075
                         0.969072
                                                0.985075
5
    AdaBoost
              0.969271
                         0.967033
                                                0.969271
6
        GBDT
              0.956102
                         0.965116
                                                0.956102
7
         BqC
              0.965759
                         0.940351
                                                0.965759
8
         SVC
              0.951712
                         0.927273
                                                0.951712
9
          DT
              0.923617
                         0.804805
                                                0.923617
10
          KN
              0.645303
                         0.417867
                                                0.645303
```

```
Precision max ft 3000 Accuracy scalling
                                               Precision scalling
0
                 0.992727
                                     0.983319
                                                         0.992727
1
                 0.992701
                                     0.982441
                                                         0.992701
2
                 0.982270
                                     0.984197
                                                         0.982270
3
                 0.969388
                                     0.987709
                                                         0.969388
4
                 0.969072
                                     0.985075
                                                         0.969072
5
                 0.967033
                                     0.969271
                                                         0.967033
6
                 0.965116
                                                         0.965116
                                     0.956102
7
                 0.940351
                                     0.965759
                                                         0.940351
8
                 0.927273
                                     0.951712
                                                         0.927273
9
                 0.804805
                                     0.923617
                                                         0.804805
10
                 0.417867
                                     0.645303
                                                         0.417867
#Voting Classifier
# svc = SVC(kernel='sigmoid', gamma=1.0,probability=True)
etc =ExtraTreesClassifier(n estimators=50, random state=2)
rf = RandomForestClassifier()
lr = LogisticRegression()
mnb = MultinomialNB()
from sklearn.ensemble import VotingClassifier
# voting = VotingClassifier(estimators=[('svm',svc),('nb',mnb),
('et',etc)], voting='soft')
voting = VotingClassifier(estimators=[('nb',mnb),('et',etc),('RF',rf),
('LR',lr)],voting='soft')
voting.fit(X train,y train)
VotingClassifier(estimators=[('nb', MultinomialNB()),
                              ('et',
                              ExtraTreesClassifier(n estimators=50,
                                                    random state=2)),
                              ('RF', RandomForestClassifier()),
                              ('LR', LogisticRegression())],
                 voting='soft')
y pred = voting.predict(X test)
print("Accuracy",accuracy_score(y_test,y_pred))
print("Precision", precision score(y test, y pred))
Accuracy 0.9912203687445127
Precision 0.9929577464788732
#Apply stacking
estimators =[('LR',lr),('nb',mnb),('ETC',etc)]
final estimator=RandomForestClassifier()
from sklearn.ensemble import StackingClassifier
clf=StackingClassifier(estimators=estimators,final estimator=final est
imator)
```

```
clf.fit(X_train,y_train)
y_pred = clf.predict(X_test)
print("Accuracy",accuracy_score(y_test,y_pred))
print("Precision",precision_score(y_test,y_pred))

Accuracy 0.9920983318700615
Precision 0.9929824561403509
import pickle
pickle.dump(tfidf,open('vectorizer.pkl','wb'))
pickle.dump(mnb,open('model.pkl','wb'))
```