```
In [103...
          import numpy as np
          import pandas as pd
           import matplotlib.pyplot as plt
           import seaborn as sns
           from sklearn.preprocessing import LabelEncoder
           from nltk import corpus
           from nltk.corpus import stopwords
           #this is for punctuation
          import string
          import nltk
          from nltk.stem.porter import PorterStemmer
          from wordcloud import WordCloud
          from collections import Counter
          from sklearn.feature extraction.text import CountVectorizer, TfidfVectorizer
          from sklearn.model selection import train test split
          from sklearn.naive bayes import GaussianNB, MultinomialNB, BernoulliNB
          from sklearn.metrics import accuracy score, confusion matrix, precision score
          from sklearn.linear model import LogisticRegression
          from sklearn.tree import DecisionTreeClassifier
          from sklearn.ensemble import RandomForestClassifier,AdaBoostClassifier,BaggingClassifier,F
          from sklearn.ensemble import GradientBoostingClassifier
          from sklearn.svm import SVC
          from sklearn.neighbors import KNeighborsClassifier
          from xgboost import XGBClassifier
          from sklearn.preprocessing import MinMaxScaler
           from sklearn.ensemble import VotingClassifier
           import pickle
 In [2]:
          df = pd.read csv('/Users/mac/Desktop/DataScience/NLP/Spam-Ham-Classifier/spam.csv',encodir
 In [3]:
           df.head()
 Out[3]:
               v1
                                                      v2 Unnamed: 2 Unnamed: 3 Unnamed: 4
          0
             ham
                     Go until jurong point, crazy.. Available only ...
                                                                             NaN
                                                                                        NaN
                                                                 NaN
          1
             ham
                                    Ok lar... Joking wif u oni...
                                                                 NaN
                                                                            NaN
                                                                                        NaN
          2 spam
                   Free entry in 2 a wkly comp to win FA Cup fina...
                                                                 NaN
                                                                            NaN
                                                                                        NaN
                    U dun say so early hor... U c already then say...
                                                                            NaN
                                                                                        NaN
             ham
                                                                 NaN
             ham
                     Nah I don't think he goes to usf, he lives aro...
                                                                 NaN
                                                                            NaN
                                                                                        NaN
 In [4]:
          df.shape
          (5572, 5)
 Out[4]:
```

• There are 5572 sms

## 1. Data Cleaning

```
In [5]: df.info()
```

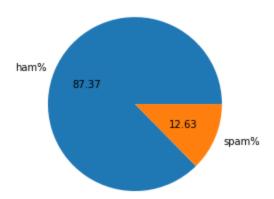
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5572 entries, 0 to 5571

```
Data columns (total 5 columns):
            #
                 Column
                               Non-Null Count
                                                   Dtype
            0
                 v1
                                5572 non-null
                                                   object
            1
                 v2
                                5572 non-null
                                                   object
            2
                 Unnamed: 2 50 non-null
                                                   object
            3
                 Unnamed: 3 12 non-null
                                                   object
                 Unnamed: 4 6 non-null
                                                   object
           dtypes: object(5)
           memory usage: 217.8+ KB
 In [6]:
            #dropping the columns
            df.drop(['Unnamed: 2','Unnamed: 3','Unnamed: 4'],axis=1,inplace=True)
 In [7]:
            df.sample(5)
 Out[7]:
                     ν1
                                                                          ν2
           2802 spam
                                        FreeMsg>FAV XMAS TONES!Reply REAL
           2858
                              Do you know why god created gap between your f...
                   ham
             163 spam
                                  -PLS STOP bootydelious (32/F) is inviting you ...
            1022
                                                        We still on for tonight?
                   ham
            3174
                  spam
                        UR GOING 2 BAHAMAS! CallFREEFONE 08081560665 a...
 In [8]:
            df.rename(columns={'v1':'Target','v2':'Text'},inplace=True)
 In [9]:
            df.head()
 Out [9]:
              Target
                                                             Text
           0
                 ham
                         Go until jurong point, crazy.. Available only ...
           1
                                          Ok lar... Joking wif u oni...
                ham
           2
                spam
                      Free entry in 2 a wkly comp to win FA Cup fina...
           3
                 ham
                        U dun say so early hor... U c already then say...
           4
                        Nah I don't think he goes to usf, he lives aro...
                 ham
In [10]:
            encoder = LabelEncoder()
            df['Target'] = encoder.fit transform(df['Target'])
In [11]:
            df.head()
Out[11]:
              Target
                                                             Text
           0
                   0
                         Go until jurong point, crazy.. Available only ...
           1
                                          Ok lar... Joking wif u oni...
           2
                   1 Free entry in 2 a wkly comp to win FA Cup fina...
           3
                   0
                        U dun say so early hor... U c already then say...
           4
                   0
                        Nah I don't think he goes to usf, he lives aro...
```

```
In [12]:
          #checking missing values
          df.isnull().sum()
         Target
                    0
Out[12]:
          Text
          dtype: int64
In [13]:
           #checking for duplicates
          df.duplicated().sum()
          403
Out[13]:
In [14]:
          df = df.drop duplicates(keep='first')
In [15]:
          df.shape
          (5169, 2)
Out[15]:
```

## 2. EDA

```
In [16]:
          (df['Target'].value counts()/df.shape[0])*100
              87.366996
Out[16]:
              12.633004
         Name: Target, dtype: float64
In [17]:
          plt.pie((df['Target'].value counts()/df.shape[0])*100,labels=['ham%','spam%'],autopct='%0.
          plt.show()
```



Clearly this is a case of imbalanced data.

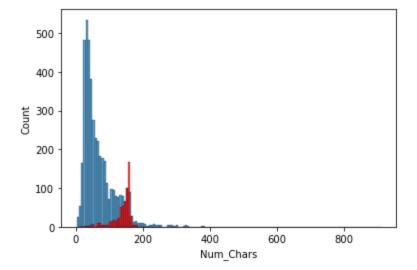
```
In [18]:
          #nltk.download('punkt')
In [19]:
          #numer of char in each sentence
          df['Num Chars'] = df['Text'].apply(len)
In [20]:
          #fetch the number of words
```

```
In [21]:
            #number of sentences
            df['Num Sentences'] = df['Text'].apply(lambda x:len(nltk.sent tokenize(x)))
In [22]:
            df.head(3)
              Target
                                                           Text Num_Chars Num_Words Num_Sentences
Out[22]:
                                                                                                        2
           0
                   0
                        Go until jurong point, crazy.. Available only ...
                                                                         111
                                                                                      24
                                                                                                        2
           1
                   0
                                         Ok lar... Joking wif u oni...
                                                                         29
                                                                                       8
           2
                   1 Free entry in 2 a wkly comp to win FA Cup fina...
                                                                        155
                                                                                      37
                                                                                                        2
In [23]:
            df.describe()
Out[23]:
                        Target
                                Num_Chars
                                             Num_Words Num_Sentences
                  5169.000000
                               5169.000000
                                             5169.000000
                                                              5169.000000
           count
           mean
                     0.126330
                                  78.977945
                                               18.455407
                                                                 1.961308
                     0.332253
                                  58.236293
             std
                                               13.322448
                                                                 1.432583
                     0.000000
                                  2.000000
                                                1.000000
             min
                                                                 1.000000
            25%
                     0.000000
                                  36.000000
                                                9.000000
                                                                 1.000000
            50%
                     0.000000
                                 60.000000
                                               15.000000
                                                                 1.000000
            75%
                     0.000000
                                 117.000000
                                               26.000000
                                                                 2.000000
            max
                     1.000000
                                 910.000000
                                              220.000000
                                                               38.000000
In [24]:
           ham = df[df['Target'] == 0]
           ham.describe()
Out[24]:
                  Target
                          Num_Chars
                                        Num_Words Num_Sentences
                          4516.000000
                                       4516.000000
                                                        4516.000000
           count 4516.0
           mean
                     0.0
                            70.459256
                                          17.123339
                                                           1.815545
                            56.358207
             std
                     0.0
                                          13.491315
                                                           1.364098
                     0.0
                             2.000000
                                          1.000000
                                                           1.000000
             min
            25%
                     0.0
                            34.000000
                                          8.000000
                                                           1.000000
            50%
                            52.000000
                     0.0
                                         13.000000
                                                           1.000000
            75%
                     0.0
                            90.000000
                                         22.000000
                                                           2.000000
            max
                     0.0
                           910.000000
                                        220.000000
                                                          38.000000
In [25]:
            spam = df[df['Target'] == 1]
            spam.describe()
Out [25]:
                  Target Num_Chars Num_Words Num_Sentences
                         653.000000
                                      653.000000
                                                       653.000000
           count
                   653.0
```

df['Num Words'] = df['Text'].apply(lambda x: len(nltk.word tokenize(x)))

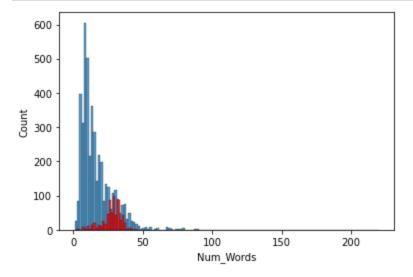
	Target	Num_Chars	Num_Words	Num_Sentences
mean	1.0	137.891271	27.667688	2.969372
std	0.0	30.137753	7.008418	1.488910
min	1.0	13.000000	2.000000	1.000000
25%	1.0	132.000000	25.000000	2.000000
50%	1.0	149.000000	29.000000	3.000000
75%	1.0	157.000000	32.000000	4.000000
max	1.0	224.000000	46.000000	9.000000

```
In [26]: sns.histplot(ham['Num_Chars'])
sns.histplot(spam['Num_Chars'],color='red');
```

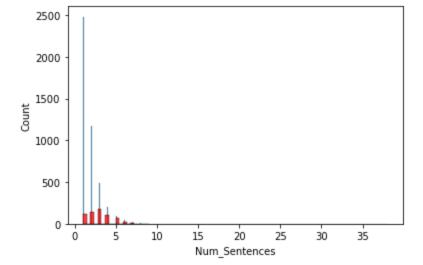


• We can see through this graph that spam messages uses more character than ham

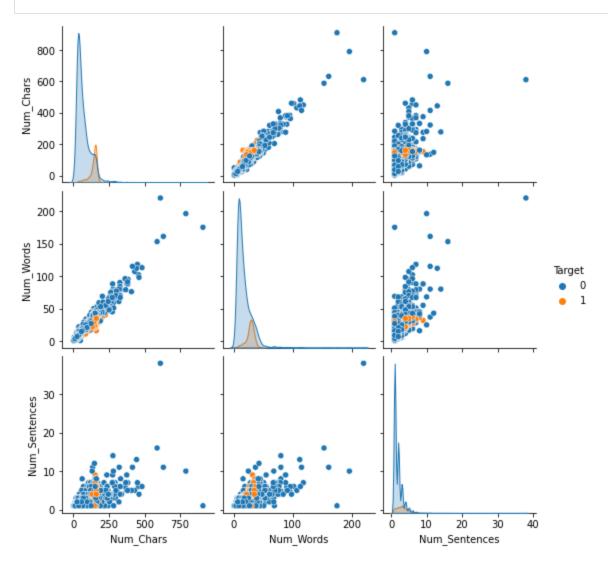
```
In [27]:
    sns.histplot(ham['Num_Words'])
    sns.histplot(spam['Num_Words'],color='red');
```



```
In [28]: sns.histplot(ham['Num_Sentences'])
    sns.histplot(spam['Num_Sentences'],color='red');
```

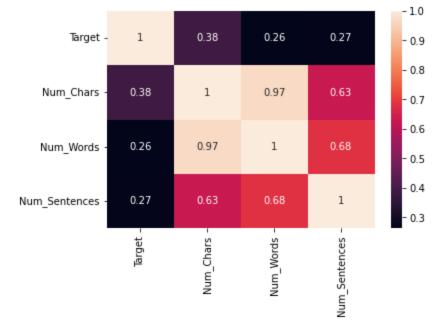


In [29]: sns.pairplot(hue='Target',data=df);



• There are outliers.

```
In [30]: sns.heatmap(df.corr(),annot=True);
```



In [31]:

Out[34]:

def transform text(text):

• Since there is high multi collinearity between feature. Therefore while building model we will only take one of the festures.

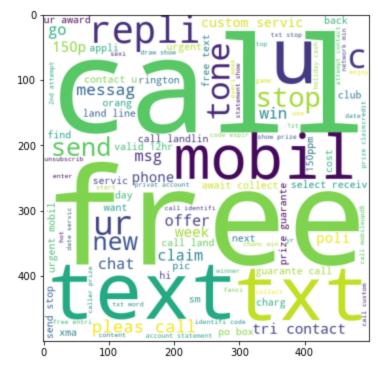
```
text = text.lower()
              text = nltk.word tokenize(text)
              #removing special chars
              y = []
              for i in text:
                  if i.isalnum():
                       y.append(i)
              text = y[:] #clonning of list
              y.clear()
              for i in text:
                  if i not in stopwords.words('english') and i not in string.punctuation:
                       y.append(i)
              text = y[:] #clonning of list
              y.clear()
              ps = PorterStemmer()
              for i in text:
                  y.append(ps.stem(i))
              return ' '.join(y)
In [32]:
          transform text('how are you Annie 3456778 %$^ ?')
          'anni 3456778'
Out[32]:
In [33]:
          df['Transformed Text'] = df['Text'].apply(transform text)
In [34]:
          df.head()
```

	Target	Text	Num_Chars	Num_Words	Num_Sentences	Transformed_Text
0	0	Go until jurong point, crazy Available only	111	24	2	go jurong point crazi avail bugi n great world
1	0	Ok lar Joking wif u oni	29	8	2	ok lar joke wif u oni
2	1	Free entry in 2 a wkly comp to win FA Cup fina	155	37	2	free entri 2 wkli comp win fa cup final tkt 21
3	0	U dun say so early hor U c already then say	49	13	1	u dun say earli hor u c alreadi say
4	0	Nah I don't think he goes to usf, he lives aro	61	15	1	nah think goe usf live around though

```
In [35]: wc = WordCloud(width=500, height=500, min_font_size=10, background_color='white')
In [36]: spam_wc = wc.generate(df[df['Target']==1]['Transformed_Text'].str.cat(sep=' '))
    plt.figure(figsize=(15,6))
```

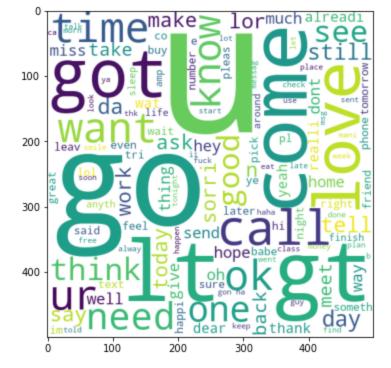
Out[36]: <matplotlib.image.AxesImage at 0x7fa3f9401bb0>

plt.imshow(spam wc)



```
In [37]: ham_wc = wc.generate(df[df['Target']==0]['Transformed_Text'].str.cat(sep=' '))
    plt.figure(figsize=(15,6))
    plt.imshow(ham_wc)
```

Out[37]: <matplotlib.image.AxesImage at 0x7fa3f8de16d0>

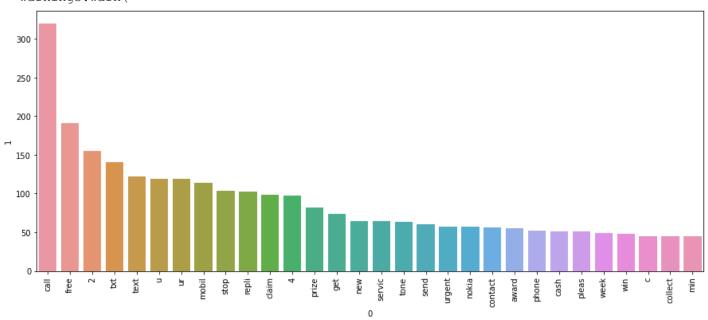


```
In [38]: spam_corpus = []
    for msg in df[df['Target']==1]['Transformed_Text'].tolist():
        for word in msg.split():
            spam_corpus.append(word)

In [39]: sp=pd.DataFrame(Counter(spam_corpus).most_common(30))
    plt.figure(figsize=(15,6))
    sns.barplot(sp[0],sp[1])
    plt.xticks(rotation='vertical')
    plt.show()
```

/Users/mac/opt/anaconda3/lib/python3.8/site-packages/seaborn/\_decorators.py:36: FutureWarn ing: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(



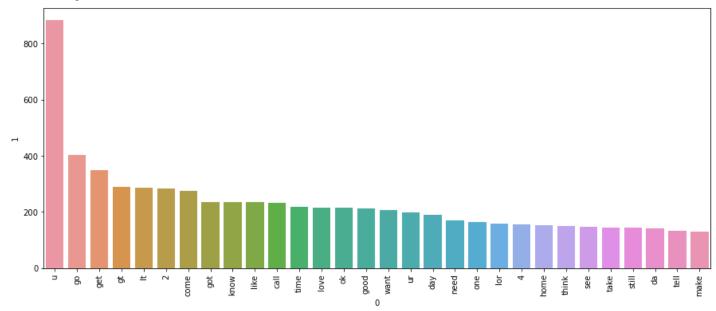
```
In [40]: ham_corpus = []
    for msg in df[df['Target']==0]['Transformed_Text'].tolist():
```

```
for word in msg.split():
    ham_corpus.append(word)
```

```
In [41]:
    h=pd.DataFrame(Counter(ham_corpus).most_common(30))
    plt.figure(figsize=(15,6))
    sns.barplot(h[0],h[1])
    plt.xticks(rotation='vertical')
    plt.show()
```

/Users/mac/opt/anaconda3/lib/python3.8/site-packages/seaborn/\_decorators.py:36: FutureWarn ing: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```



```
Model Building
In [42]:
          cv = CountVectorizer()
          X = cv.fit transform(df['Transformed Text']).toarray()
In [43]:
          X.shape
          (5169, 6708)
Out[43]:
In [44]:
          y = df['Target'].values
In [45]:
          X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=2)
In [46]:
          gng = GaussianNB()
          bnb = BernoulliNB()
          mnb = MultinomialNB()
In [47]:
          gng.fit(X train,y train)
          y pred1 = gng.predict(X test)
          print('Accuracy:',accuracy score(y test,y pred1))
```

```
print('Precision Score:',precision score(y test,y pred1))
         Accuracy: 0.8800773694390716
         Confusion Matrix: [[792 104]
          [ 20 118]]
         Precision Score: 0.5315315315315315
In [48]:
          mnb.fit(X train, y train)
          y pred2 = mnb.predict(X test)
          print('Accuracy:',accuracy score(y test,y pred2))
          print('Confusion Matrix:',confusion matrix(y test,y pred2))
          print('Precision Score:',precision score(y test,y pred2))
         Accuracy: 0.9642166344294004
         Confusion Matrix: [[871 25]
          [ 12 126]]
         Precision Score: 0.8344370860927153
In [49]:
          bnb.fit(X train, y train)
          y pred3 = bnb.predict(X test)
          print('Accuracy:',accuracy_score(y_test,y_pred3))
          print('Confusion Matrix:',confusion matrix(y test,y pred3))
          print('Precision Score:',precision score(y test,y pred3))
         Accuracy: 0.9700193423597679
         Confusion Matrix: [[893
          [ 28 110]]
         Precision Score: 0.9734513274336283
          • Since there is an imbalanced data hence precision matters here most.
         Using tf-idf
In [104...
          tv = TfidfVectorizer()
          X = tv.fit transform(df['Transformed Text']).toarray()
In [105...
          y = df['Target'].values
In [106...
          X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=2)
In [107...
          gng = GaussianNB()
          bnb = BernoulliNB()
          mnb = MultinomialNB()
In [108...
          gng.fit(X train,y train)
          y pred1 = gng.predict(X test)
          print('Accuracy:',accuracy score(y test,y pred1))
          print('Confusion Matrix:',confusion matrix(y test,y pred1))
          print('Precision Score:',precision score(y test,y pred1))
         Accuracy: 0.8762088974854932
         Confusion Matrix: [[793 103]
          [ 25 113]]
          Precision Score: 0.5231481481481481
In [109...
```

mnb.fit(X train, y train)

print('Confusion Matrix:',confusion matrix(y test,y pred1))

```
print('Accuracy:',accuracy score(y test,y pred2))
          print('Confusion Matrix:',confusion matrix(y test,y pred2))
          print('Precision Score:',precision_score(y test,y pred2))
         Accuracy: 0.9593810444874274
         Confusion Matrix: [[896 0]
          [ 42 96]]
         Precision Score: 1.0
In [56]:
          bnb.fit(X train, y train)
          y pred3 = bnb.predict(X test)
          print('Accuracy:',accuracy score(y test,y pred3))
          print('Confusion Matrix:',confusion matrix(y test,y pred3))
          print('Precision Score:',precision score(y test,y pred3))
         Accuracy: 0.9700193423597679
         Confusion Matrix: [[893
          [ 28 110]]
         Precision Score: 0.9734513274336283

    choosen tf-idf and algo- Multinomial

In [57]:
          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)
In [58]:
          clfs = {
              'SVC':svc,
              'KN':knc,
              'NB':mnb,
              'DT': dtc,
              'LR': lrc,
              'RF': rfc,
              'AdaBoost': abc,
              'BgC': bc,
              'ETC': etc,
              'GBDT':gbdt,
              'xgb':xgb
          }
In [59]:
          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
In [60]:
          train classifier(svc, X train, y train, X test, y test)
```

y pred2 = mnb.predict(X test)

```
Out [60]: (0.9729206963249516, 0.9741379310344828)
In [61]:
          accuracy scores = []
          precision scores = []
          for name, clf in clfs.items():
              current accuracy, current precision = train classifier(clf, X train, y train, X test, y te
              print("For ", name)
              print("Accuracy - ", current accuracy)
              print("Precision - ", current precision)
              accuracy scores.append(current accuracy)
              precision scores.append(current precision)
         For SVC
         Accuracy - 0.9729206963249516
         Precision - 0.9741379310344828
         For KN
         Accuracy - 0.9003868471953579
         Precision - 1.0
         For NB
         Accuracy - 0.9593810444874274
         Precision - 1.0
         For DT
         Accuracy - 0.9361702127659575
         Precision - 0.8461538461538461
         For LR
         Accuracy - 0.9516441005802708
         Precision - 0.94
         For RF
         Accuracy - 0.971953578336557
         Precision - 1.0
         For AdaBoost
         Accuracy - 0.9613152804642167
         Precision - 0.9454545454545454
         For BgC
         Accuracy - 0.9584139264990329
         Precision - 0.8625954198473282
         For ETC
         Accuracy - 0.9729206963249516
         Precision - 0.9824561403508771
         For GBDT
         Accuracy - 0.9526112185686654
         Precision - 0.9238095238095239
         For xgb
         Accuracy - 0.9700193423597679
         Precision - 0.9349593495934959
In [62]:
          performance df = pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy':accuracy scores,'Precisi
In [63]:
          performance df
Out[63]:
           Algorithm Accuracy Precision
                  KN 0.900387 1.000000
          2
                  NB 0.959381 1.000000
```

5

8

RF 0.971954 1.000000

ETC 0.972921 0.982456

```
        Algorithm
        Accuracy
        Precision

        0
        SVC
        0.972921
        0.974138

        6
        AdaBoost
        0.961315
        0.945455

        4
        LR
        0.951644
        0.940000

        10
        xgb
        0.970019
        0.934959

        9
        GBDT
        0.952611
        0.923810

        7
        BgC
        0.958414
        0.862595

        3
        DT
        0.936170
        0.846154
```

```
In [64]: performance_df1 = pd.melt(performance_df, id_vars = "Algorithm")
```

In [65]: perf

performance\_df1

Out[65]: Alg

	Algorithm	variable	value
0	KN	Accuracy	0.900387
1	NB	Accuracy	0.959381
2	RF	Accuracy	0.971954
3	ETC	Accuracy	0.972921
4	SVC	Accuracy	0.972921
5	AdaBoost	Accuracy	0.961315
6	LR	Accuracy	0.951644
7	xgb	Accuracy	0.970019
8	GBDT	Accuracy	0.952611
9	BgC	Accuracy	0.958414
10	DT	Accuracy	0.936170
11	KN	Precision	1.000000
12	NB	Precision	1.000000
13	RF	Precision	1.000000
14	ETC	Precision	0.982456
15	SVC	Precision	0.974138
16	AdaBoost	Precision	0.945455
17	LR	Precision	0.940000
18	xgb	Precision	0.934959
19	GBDT	Precision	0.923810
20	BgC	Precision	0.862595
21	DT	Precision	0.846154

```
In [66]:
```

```
# model improve
```

<sup># 1.</sup> Change the max features parameter of TfIdf

```
tv = TfidfVectorizer(max features=3000)
In [67]:
          X = tv.fit transform(df['Transformed Text']).toarray()
          y = df['Target'].values
          X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=2)
In [68]:
          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)
In [69]:
          clfs = {
              'SVC':svc,
              'KN':knc,
              'NB':mnb,
              'DT': dtc,
              'LR': lrc,
              'RF': rfc,
              'AdaBoost': abc,
              'BgC': bc,
              'ETC': etc,
              'GBDT':gbdt,
              'xgb':xgb
In [70]:
          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
In [71]:
          accuracy scores = []
          precision scores = []
          for name, clf in clfs.items():
              current accuracy, current precision = train classifier(clf, X train, y train, X test, y te
              print("For ", name)
              print("Accuracy - ", current accuracy)
              print("Precision - ", current precision)
              accuracy scores.append(current accuracy)
              precision scores.append(current precision)
         For SVC
         Accuracy - 0.9758220502901354
         Precision - 0.9747899159663865
         For KN
         Accuracy - 0.9052224371373307
         Precision - 1.0
         For NB
```

```
For DT
         Accuracy - 0.9303675048355899
         Precision - 0.8367346938775511
         For LR
         Accuracy - 0.9584139264990329
         Precision - 0.9702970297029703
         Accuracy - 0.9758220502901354
         Precision - 0.9829059829059829
         For AdaBoost
         Accuracy - 0.960348162475822
         Precision - 0.9292035398230089
         For BqC
         Accuracy - 0.9584139264990329
         Precision - 0.8682170542635659
         For ETC
         Accuracy - 0.9748549323017408
         Precision - 0.9745762711864406
         For GBDT
         Accuracy - 0.9468085106382979
         Precision - 0.91919191919192
         For xgb
         Accuracy - 0.9671179883945842
         In [74]:
          performance df1 = pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy max ft3000':accuracy sco
In [75]:
          performance df1
             Algorithm Accuracy_max_ft3000 Precision_max_ft3000
Out[75]:
          1
                                 0.905222
                                                     1.000000
                  ΚN
          2
                                                     1.000000
                  NB
                                 0.970986
          5
                  RF
                                 0.975822
                                                     0.982906
                  SVC
          0
                                 0.975822
                                                     0.974790
          8
                  ETC
                                 0.974855
                                                     0.974576
          4
                  LR
                                 0.958414
                                                     0.970297
         10
                  xgb
                                  0.967118
                                                     0.933333
             AdaBoost
                                 0.960348
                                                     0.929204
          9
                GBDT
                                 0.946809
                                                     0.919192
          7
                  BgC
                                 0.958414
                                                     0.868217
          3
                  DT
                                 0.930368
                                                     0.836735
In [83]:
          performance df = pd.merge(on='Algorithm', left=performance df, right=performance df1)
In [84]:
          performance df
Out[84]:
             Algorithm Accuracy Precision Accuracy_max_ft3000 Precision_max_ft3000
          0
                  ΚN
                      0.900387
                                1.000000
                                                    0.905222
                                                                       1.000000
```

Accuracy - 0.9709864603481625

Precision - 1.0

	Algorithm	Accuracy	Precision	Accuracy_max_ft3000	Precision_max_ft3000
1	NB	0.959381	1.000000	0.970986	1.000000
2	RF	0.971954	1.000000	0.975822	0.982906
3	ETC	0.972921	0.982456	0.974855	0.974576
4	SVC	0.972921	0.974138	0.975822	0.974790
5	AdaBoost	0.961315	0.945455	0.960348	0.929204
6	LR	0.951644	0.940000	0.958414	0.970297
7	xgb	0.970019	0.934959	0.967118	0.933333
8	GBDT	0.952611	0.923810	0.946809	0.919192
9	BgC	0.958414	0.862595	0.958414	0.868217
10	DT	0.936170	0.846154	0.930368	0.836735

• The benefit we got is that our accuracy increased in somecases. #### Voting Classifier

```
In [100...
          #taking the models that performed nicely
          rfc = RandomForestClassifier(n estimators=50, random state=2)
          etc = ExtraTreesClassifier(n estimators=50, random state=2)
          svc = SVC(kernel='sigmoid', gamma=1.0,probability=True)
          mnb = MultinomialNB()
          vc = VotingClassifier(estimators = [('svm', svc), ('rfc', rfc), ('etc', etc), ('mnb', mnb)], voting
In [101...
          vc.fit(X train,y train)
         VotingClassifier(estimators=[('svm',
Out [101...
                                         SVC(gamma=1.0, kernel='sigmoid',
                                             probability=True)),
                                         RandomForestClassifier(n estimators=50,
                                                                 random state=2)),
                                        ('etc',
                                         ExtraTreesClassifier(n estimators=50,
                                                               random state=2)),
                                        ('mnb', MultinomialNB())],
                           voting='soft')
In [102...
          y pred = vc.predict(X test)
          print('Accuracy:',accuracy score(y test,y pred))
          print('Precision Score:',precision score(y test,y pred))
         Accuracy: 0.9796905222437138
         Precision Score: 0.9915966386554622

    We are gonna apply Naive Bayes

In [111...
          pickle.dump(tv,open('vectorizer.pkl','wb'))
```

pickle.dump(mnb,open('sms model.pkl','wb'))

In []:

In [ ]:	
In [ ]:	
In [ ]:	