Team No:-PNT2022TMID08666

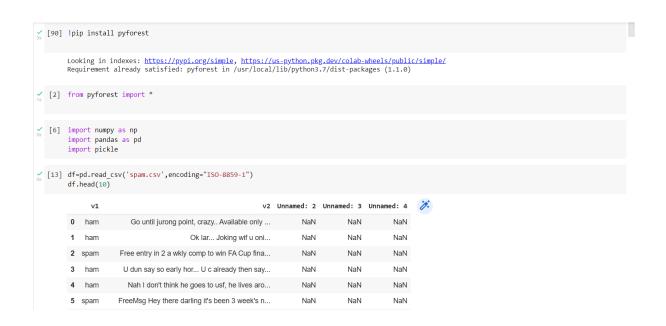
Assignment 4

Student Name	Sreeleka
Student Roll Number	19BCS315
Maximum Marks	2 Marks

Problem Statement :- SMS SPAM Classification

Problem Statement: Over recent years, as the popularity of mobile phone devices has increased, Short Message Service (SMS) has grown into a multibillion dollar industry. At the same time, reduction in the cost of messaging services has resulted in growth in unsolicited commercial advertisements (spams) being sent to mobile phones. Due to Spam SMS, Mobile service providers suffer from some sort of financial problems as well as it reduces calling time for users. Unfortunately, if the user accesses such Spam SMS they may face the problem of virus or malware. When SMS arrives at mobile it will disturb mobile user privacy and concentration. It may lead to frustration for the user. So Spam SMS is one of the major issues in the wireless communication world and it grows day by day.

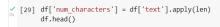
Screenshots:-











t	arget	text	num_characters
0	0	Go until jurong point, crazy Available only	111
1	0	Ok lar Joking wif u oni	29
2	1	Free entry in 2 a wkly comp to win FA Cup fina	155
3	0	U dun say so early hor U c already then say	49
4	0	Nah I don't think he goes to usf, he lives aro	61

[30] df['num_words'] = df['text'].apply(lambda x:len(nltk.word_tokenize(x))) df.head()

	target	text	num_characters	num_words
0	0	Go until jurong point, crazy Available only	111	24
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
3	0	U dun say so early hor U c already then say	49	13
4	0	Nah I don't think he goes to usf, he lives aro	61	15

%

[31] df['num_sentences'] = df['text'].apply(lambda x:len(nltk.sent_tokenize(x))) df[['num_characters','num_words','num_sentences']].describe()

	num_characters	num_words	num_sentences
count	5169.000000	5169.000000	5169.000000
mean	78.977945	18.453279	1.947185
std	58.236293	13.324793	1.362406
min	2.000000	1.000000	1.000000
25%	36.000000	9.000000	1.000000
50%	60.000000	15.000000	1.000000
75%	117.000000	26.000000	2.000000
max	910.000000	220.000000	28.000000

// [32] df[df['target'] == 0][['num_characters','num_words','num_sentences']].describe()
// [32] df['target'] == 0][['num_characters','num_words

%

	num_characters	num_words	num_sentences
count	4516.000000	4516.000000	4516.000000
mean	70.459256	17.120903	1.799601
std	56.358207	13.493725	1.278465
min	2.000000	1.000000	1.000000

```
[33] 75% 157.00000 32.00000 4.00000

max 224.00000 46.00000 8.000000

[80] plt.figure(figsize=(12,6))
sns.histplot(df[df['target'] == 0]['num_characters'])
sns.histplot(df[df['target'] == 1]['num_characters'],color='red')
print("Team No:- PNTZ02ZTMID08666")

Team No:- PNTZ02ZTMID08666

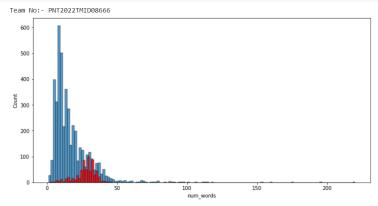
500

400

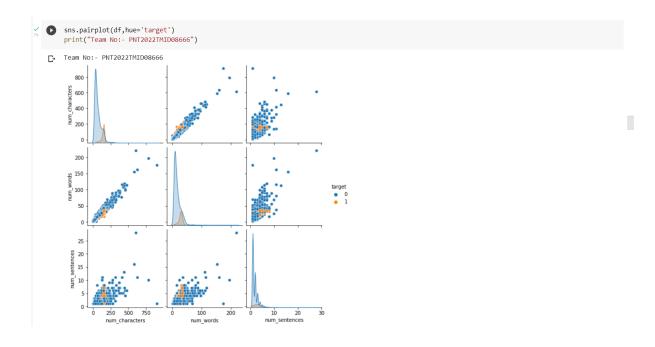
200

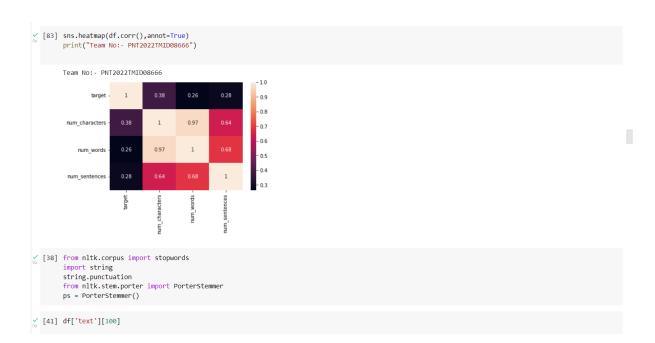
100
```

```
[81] plt.figure(figsize=(12,6))
    sns.histplot(df[df['target'] == 0]['num_words'])
    sns.histplot(df[df['target'] == 1]['num_words'],color='red')
    print("Team No:- PNT2022TMID08666")
```



400 num_characters





[41] df['text'][100]

'Okay name ur price as long as its legal! Wen can I pick them up? Y u ave x ams xx'

[46] import nltk

nltk.download('stopwords')

 $\begin{tabular}{ll} [nltk_data] & Downloading package stopwords to /root/nltk_data... \\ [nltk_data] & Unzipping corpora/stopwords.zip. \\ True & \begin{tabular}{ll} True & \begin{tab$

ta	arget	text	num_characters	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

%

[84] from wordcloud import wordcloud

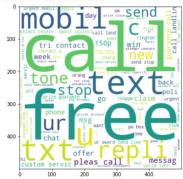
wc = Wordcloud(width=500,height=500,min_font_size=10,background_color='white')

spam_wc = wc.generate(df[df['target'] == 1]['transformed_text'].str.cat(sep=" "))

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

plt.imshow(spam_wc)
print("Team No:- PNT2022TMID08666")

[84] Team No:- PNT2022TMID08666



[85] ham_wc = wc.generate(df[df['target'] == 0]['transformed_text'].str.cat(sep=" ")) plt.figure(figsize=(15,6)) plt.imshow(ham_wc)
print("Team No:- PNT2022TMID08666")



```
[50] # For Spam messages
spam_corpus = []
                 spam_corpus = []
for msg in df[df['target'] == 1]['transformed_text'].tolist():
    for word in msg.split():
                                         spam_corpus.append(word)
             len(spam_corpus)
                9939
[86] 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')
                plt.show()
print("Team No:- PNT2022TMID08666")
                /usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version (FutureWarning
                          300
                          250
                          200
                   150
                          100
                                                                 wobill stopic of the control of the 
                 Team No:- PNT2022TMID08666
[52] # For ham messages
              len(ham_corpus)
[53] from sklearn.feature_extraction.text import CountVectorizer,TfidfVectorizer
                cv = CountVectorizer()
                tfidf = TfidfVectorizer(max_features=3000)
x = tfidf.fit_transform(df['transformed_text']).toarray()
               x.shape
y = df['target'].values
                 from sklearn.model_selection import train_test_split
[54] 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
                mnb = MultinomialNB()
                 # MultinomialNB
              mnb.fit(X_train,y_train)
y_pred1 = mnb.predict(X_test)
print(accuracy_score(y_test,y_pred1))
print(confusion_matrix(y_test,y_pred1))
                print(precision_score(y_test,y_pred1))
```

0.9709864603481625 [[896 0] [30 108]] 1.0

```
[55] 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
 [56] svc = SVC(kernel='sigmoid', gamma=1.0)
knc = KNeighborsClassifier()
          mnb = MultinomialNB()
dtc = DecisionTreeClassifier(max_depth=5)
          Trc = 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)
[57] clfs = {
    'SVC' : svc,
    'KN' : knc,
    'NB': mnb,
    'DT': dtc,
    'LR': lrc,
    'se': rfc,
                 'RF': rfc,
'AdaBoost': abc,
                 'BgC': bc,
  [58] train_classifier(svc,X_train,y_train,X_test,y_test)
               (0.9758220502901354, 0.9747899159663865)
 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 - 0.9758220502901354
Precision - 0.9747899159663865
              For KN
Accuracy - 0.9052224371373307
Precision - 1.0
              For NB
Accuracy - 0.9709864603481625
Precision - 1.0
              For DT
Accuracy - 0.9332688588007737
Precision - 0.8415841584158416
              For LR
              Accuracy - 0.9584139264990329
Precision - 0.9702970297029703
               For RF
```

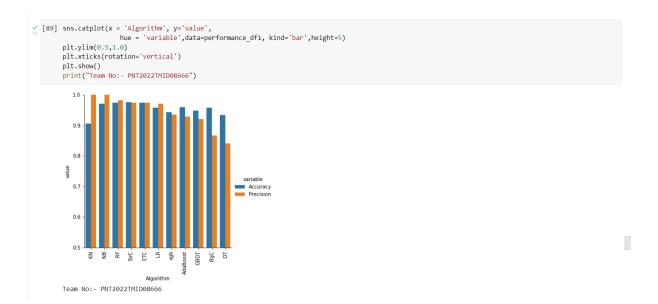
Accuracy - 0.9748549323017408

// [65] performance_df = pd.DataFrame({'Algorithm':clfs.keys(), 'Accuracy':accuracy_scores, 'Precision':precision_scores}).sort_values('Precision', ascending performance_df)

	Algorithm	Accuracy	Precision
1	KN	0.905222	1.000000
2	NB	0.970986	1.000000
5	RF	0.974855	0.982759
0	SVC	0.975822	0.974790
8	ETC	0.974855	0.974576
4	LR	0.958414	0.970297
10	xgb	0.943907	0.934783
6	AdaBoost	0.960348	0.929204
9	GBDT	0.947776	0.920000
7	BgC	0.957447	0.867188
3	DT	0.933269	0.841584

[66] performance_df1 = pd.melt(performance_df, id_vars = "Algorithm")
 performance_df1

	Algorithm	variable	value	7.
0	KN	Accuracy	0.905222	
1	NB	Accuracy	0.970986	
2	RF	Accuracy	0.974855	
3	SVC	Accuracy	0.975822	
4	ETC	Accuracy	0.974855	
5	LR	Accuracy	0.958414	
6	xgb	Accuracy	0.943907	
7	AdaBoost	Accuracy	0.960348	
8	GBDT	Accuracy	0.947776	
9	BgC	Accuracy	0.957447	
10	DT	Accuracy	0.933269	
11	KN	Precision	1.000000	
12	NB	Precision	1.000000	
13	RF	Precision	0.982759	
14	SVC	Precision	0.974790	
15	ETC	Precision	0.974576	
16	LR	Precision	0.970297	
17	xgb	Precision	0.934783	



// [68] temp_df = pd.DataFrame({'Algorithm':clfs.keys(), 'Accuracy_max_ft_3000':accuracy_scores, 'Precision_max_ft_3000':precision_scores}).sort_values('Protemp_df)

temp_df

	Algorithm	Accuracy_max_ft_3000	Precision_max_ft_3000
1	KN	0.905222	1.000000
2	NB	0.970986	1.000000
5	RF	0.974855	0.982759
0	SVC	0.975822	0.974790
8	ETC	0.974855	0.974576
4	LR	0.958414	0.970297
10	xgb	0.943907	0.934783
6	AdaBoost	0.960348	0.929204
9	GBDT	0.947776	0.920000
7	BgC	0.957447	0.867188
3	DT	0.933269	0.841584

[69] temp_df = pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy_scaling':accuracy_scores,'Precision_scaling':precision_scores}).sort_values('Precision_scores')

%

	Algorithm	Accuracy_scaling	Precision_scaling
1	KN	0.905222	1.000000
2	NB	0.970986	1.000000
5	RF	0.974855	0.982759

[70] new_df = performance_df.merge(temp_df,on='Algorithm') new_df

	Algorithm	Accuracy	Precision	Accuracy_scaling	Precision_scaling
0	KN	0.905222	1.000000	0.905222	1.000000
1	NB	0.970986	1.000000	0.970986	1.000000
2	RF	0.974855	0.982759	0.974855	0.982759
3	SVC	0.975822	0.974790	0.975822	0.974790
4	ETC	0.974855	0.974576	0.974855	0.974576
5	LR	0.958414	0.970297	0.958414	0.970297
6	xgb	0.943907	0.934783	0.943907	0.934783
7	AdaBoost	0.960348	0.929204	0.960348	0.929204
8	GBDT	0.947776	0.920000	0.947776	0.920000
9	BgC	0.957447	0.867188	0.957447	0.867188
10	DT	0.933269	0.841584	0.933269	0.841584

/ [71] new_df_scaled = new_df.merge(temp_df,on='Algorithm') new_df_scaled

	Algorithm	Accuracy	Precision	Accuracy_scaling_x	Precision_scaling_x	Accuracy_scaling_y	Precision_scaling_y
0	KN	0.905222	1.000000	0.905222	1.000000	0.905222	1.000000
1	NB	0.970986	1.000000	0.970986	1.000000	0.970986	1.000000
2	RF	0.974855	0.982759	0.974855	0.982759	0.974855	0.982759

1.

1.

√ 0s [71]	5	LR	0.958414	0.970297	0.958414	0.970297	0.958414	0.970297
	6	xgb	0.943907	0.934783	0.943907	0.934783	0.943907	0.934783
	7	AdaBoost	0.960348	0.929204	0.960348	0.929204	0.960348	0.929204
	8	GBDT	0.947776	0.920000	0.947776	0.920000	0.947776	0.920000
	9	BgC	0.957447	0.867188	0.957447	0.867188	0.957447	0.867188
	10	DT	0.933269	0.841584	0.933269	0.841584	0.933269	0.841584

os	0	<pre>new_df_scaled = new_df.merge(temp_df,on='Algorithm')</pre>
		new_df_scaled

₽		Algorithm	Accuracy	Precision	Accuracy_scaling_x	Precision_scaling_x	Accuracy_scaling_y	Precision_scaling_y
	0	KN	0.905222	1.000000	0.905222	1.000000	0.905222	1.000000
	1	NB	0.970986	1.000000	0.970986	1.000000	0.970986	1.000000
	2	RF	0.974855	0.982759	0.974855	0.982759	0.974855	0.982759
	3	SVC	0.975822	0.974790	0.975822	0.974790	0.975822	0.974790
	4	ETC	0.974855	0.974576	0.974855	0.974576	0.974855	0.974576
	5	LR	0.958414	0.970297	0.958414	0.970297	0.958414	0.970297
	6	xgb	0.943907	0.934783	0.943907	0.934783	0.943907	0.934783
	7	AdaBoost	0.960348	0.929204	0.960348	0.929204	0.960348	0.929204
	8	GBDT	0.947776	0.920000	0.947776	0.920000	0.947776	0.920000
	9	BgC	0.957447	0.867188	0.957447	0.867188	0.957447	0.867188
	10	DT	0.933269	0.841584	0.933269	0.841584	0.933269	0.841584

```
(73] # Voting Classifier
svc = SVC(kernel='sigmoid', gamma=1.0,probability=True)
mnb = MultinomialNB()
                          etc = ExtraTreesClassifier(n_estimators=50, random_state=2)
                          from sklearn.ensemble import VotingClassifier
                          voting = VotingClassifier(estimators=[('svm', svc), ('nb', mnb), ('et', etc)],voting='soft')
                         voting
                         VotingClassifier(estimators=[('svm',
                                                                                                                   probability=True)),
('nb', MultinomialNB()),
('et',
                                                                                                                       SVC(gamma=1.0, kernel='sigmoid',
                                                                                                                      voting='soft')
  [74] voting.fit(X_train,y_train)
                         \label{eq:votingClassifier} VotingClassifier (estimators = [ ('svm', SVC(gamma = 1.0, kernel = 'sigmoid', SVC(gamma = 1.0, kernel = 1.0, kernel = 'sigmoid', 
                                                                                                                   probability=True)),
('nb', MultinomialNB()),
('et',
                                                                                                                      ExtraTreesClassifier(n_estimators=50,
                                                                                                                                                                                        random_state=2))],
   [75] y_pred = voting.predict(X_test)
                         print("Accuracy",accuracy_score(y_test,y_pred))
print("Precision",precision_score(y_test,y_pred))
                          Accuracy 0.9816247582205029
                          Precision 0.9917355371900827
[76] # Applying stacking estimators=[('svm', svc), ('nb', mnb), ('et', etc)]
                       final_estimator=RandomForestClassifier()
from sklearn.ensemble import StackingClassifier
clf = StackingClassifier(estimators=estimators, final_estimator=final_estimator)
                        clf.fit(X_train,y_train)
                       pred = clf.predict(X_test)
print("Accuracy",accuracy_score(y_test,y_pred))
print("Precision",precision_score(y_test,y_pred))
                        Accuracy 0.9787234042553191
[77] import pickle pickle.dump(tfidf,open('vectorizer.pkl','wb'))
                        pickle.dump(mnb,open('model.pkl','wb'))
```

Code:-

```
!pip install pyforest
from pyforest import *
import numpy as np
import pandas as pd
import pickle
df=pd.read csv('spam.csv',encoding="ISO-8859-1")
df.head(10)
df.shape
df.info()
df.drop(columns=['Unnamed: 2','Unnamed: 3','Unnamed: 4'],inplace=True)
df.head()
df.rename(columns={'v1':'target','v2':'text'},inplace=True)
from sklearn.preprocessing import LabelEncoder
encoder = LabelEncoder()
df['target'] = encoder.fit transform(df['target'])
```

```
df.head()
df.isnull().sum()
df.duplicated().sum()
df=df.drop duplicates(keep='first')
df.duplicated().sum()
df.shape
df['target'].value counts()
plt.pie(df['target'].value counts(), labels=['ham','spam'],autopct="%0.2f")
plt.show()
print("Team No:- PNT2022TMID08666")
import nltk
%pip install nltk
nltk.download('punkt')
df['num characters'] = df['text'].apply(len)
df.head()
df['num words'] = df['text'].apply(lambda x:len(nltk.word tokenize(x)))
df.head()
df['num sentences'] = df['text'].apply(lambda x:len(nltk.sent tokenize(x)))
df[['num characters','num words','num sentences']].describe()
df[df['target'] == 0][['num characters','num words','num sentences']].describe()
df[df['target'] == 1][['num_characters', 'num_words', 'num_sentences']].describe()
plt.figure(figsize=(12,6))
sns.histplot(df[df['target'] == 0]['num_characters'])
sns.histplot(df[df['target'] == 1]['num characters'],color='red')
print("Team No:- PNT2022TMID08666")
plt.figure(figsize=(12,6))
sns.histplot(df[df['target'] == 0]['num words'])
sns.histplot(df[df['target'] == 1]['num words'],color='red')
print("Team No:- PNT2022TMID08666")
sns.pairplot(df,hue='target')
print("Team No:- PNT2022TMID08666")
sns.heatmap(df.corr(),annot=True)
print("Team No:- PNT2022TMID08666")
from nltk.corpus import stopwords
import string
string.punctuation
from nltk.stem.porter import PorterStemmer
ps = PorterStemmer()
df['text'][100]
import nltk
nltk.download('stopwords')
df['transformed text'] = df['text'].apply(transform text)
df.head()
from wordcloud import WordCloud
```

```
wc = WordCloud(width=500,height=500,min font size=10,background color='white')
spam wc = wc.generate(df[df['target'] == 1]['transformed text'].str.cat(sep=" "))
plt.figure(figsize=(15,6))
plt.imshow(spam wc)
print("Team No:- PNT2022TMID08666")
ham wc = wc.generate(df[df['target'] == 0]['transformed text'].str.cat(sep=" "))
plt.figure(figsize=(15,6))
plt.imshow(ham wc)
print("Team No:- PNT2022TMID08666")
# For Spam messages
spam corpus = []
for msg in df[df['target'] == 1]['transformed_text'].tolist():
 for word in msg.split():
spam corpus.append(word)
len(spam corpus)
from collections import Counter
sns.barplot(pd.DataFrame(Counter(spam corpus).most common(30))[0],pd.DataFrame(Cou
nter(spam corpus).most common(30))[1])
plt.xticks(rotation='vertical')
plt.show()
print("Team No:- PNT2022TMID08666")
# For ham messages
ham corpus = []
for msg in df[df['target'] == 0]['transformed text'].tolist():
 for word in msg.split():
ham corpus.append(word)
len(ham corpus)
from sklearn.feature extraction.text import CountVectorizer,TfidfVectorizer
cv = CountVectorizer()
tfidf = TfidfVectorizer(max features=3000)
X = tfidf.fit transform(df['transformed text']).toarray()
X.shape
y = df['target'].values
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
mnb = MultinomialNB()
# MultinomialNB
mnb.fit(X train,y train)
y pred1 = mnb.predict(X test)
print(accuracy score(y test,y pred1))
print(confusion_matrix(y_test,y_pred1))
print(precision score(y test,y pred1))
```

```
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': Irc,
  'RF': rfc,
  'AdaBoost': abc,
  'BgC': bc,
  'ETC': etc,
  'GBDT':gbdt,
  'xgb':xgb
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)
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)
performance df
pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy':accuracy scores,'Precision':precision scores
}).sort_values('Precision',ascending=False)
performance df
performance_df1 = pd.melt(performance_df, id_vars = "Algorithm")
performance df1
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()
print("Team No:- PNT2022TMID08666")
temp df
pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy_max_ft_3000':accuracy_scores,'Precision_m
ax ft 3000':precision scores}).sort values('Precision max ft 3000',ascending=False)
temp df
temp df
pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy_scaling':accuracy_scores,'Precision_scaling':p
recision scores}).sort values('Precision scaling',ascending=False)
temp df
new df = performance df.merge(temp df,on='Algorithm')
new df
new df scaled = new df.merge(temp df,on='Algorithm')
new df scaled
new_df_scaled = new_df.merge(temp_df,on='Algorithm')
new df scaled
# Voting Classifier
svc = SVC(kernel='sigmoid', gamma=1.0,probability=True)
mnb = MultinomialNB()
etc = ExtraTreesClassifier(n_estimators=50, random_state=2)
from sklearn.ensemble import VotingClassifier
voting = VotingClassifier(estimators=[('svm', svc), ('nb', mnb), ('et', etc)],voting='soft')
voting
voting.fit(X_train,y_train)
y pred = voting.predict(X test)
print("Accuracy",accuracy_score(y_test,y_pred))
print("Precision",precision score(y test,y pred))
```

```
# Applying stacking
estimators=[('svm', svc), ('nb', mnb), ('et', etc)]
final_estimator=RandomForestClassifier()
from sklearn.ensemble import StackingClassifier
clf = StackingClassifier(estimators=estimators, final_estimator=final_estimator)
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))
import pickle
pickle.dump(tfidf,open('vectorizer.pkl','wb'))
pickle.dump(mnb,open('model.pkl','wb'))
```