```
import matplotlib.pyplot as plt
            6 from nltk.stem.porter import PorterStemmer
               ps = PorterStemmer()
            8 from nltk.corpus import stopwords
           10 from wordcloud import WordCloud
               wc = WordCloud(width=500,height=500,min_font_size=10,background_color='white')
           13 from collections import Counter
               from sklearn.feature_extraction.text import CountVectorizer,TfidfVectorizer
               cv = CountVectorizer()
           17 tfidf = TfidfVectorizer()
           19 import string
           20 import seaborn as sns
               from sklearn.preprocessing import LabelEncoder
           22 encoder = LabelEncoder()
In [2]:
            df = pd.read_csv('spam_classifier\\spam.csv',encoding='ISO-8859-1')
            2 df
Out[2]:
                                                            v2 Unnamed: 2 Unnamed: 3 Unnamed: 4
                                                                                               NaN
             0 ham
                          Go until jurong point, crazy.. Available only ...
                                                                      NaN
                                                                                   NaN
                                         Ok lar... Joking wif u oni...
             1 ham
                                                                      NaN
                                                                                   NaN
                                                                                               NaN
             2 spam Free entry in 2 a wkly comp to win FA Cup fina...
                                                                                               NaN
             3 ham
                       U dun say so early hor... U c already then say...
                                                                      NaN
                                                                                   NaN
                                                                                               NaN
                 ham
                         Nah I don't think he goes to usf, he lives aro...
                                                                       NaN
                                                                                   NaN
                                                                                               NaN
                                                                                   NaN
                                                                                               NaN
           5567 spam
                       This is the 2nd time we have tried 2 contact u...
                                                                       NaN
                               Will I b going to esplanade fr home?
           5568 ham
                                                                      NaN
                                                                                   NaN
                                                                                               NaN
           5569 ham
                         Pity, * was in mood for that. So...any other s...
                                                                       NaN
                                                                                   NaN
                                                                                               NaN
                        The guy did some bitching but I acted like i'd...
                                                                                   NaN
                                                                                               NaN
           5571 ham
                                           Rofl. Its true to its name
                                                                                   NaN
                                                                                               NaN
                                                                       NaN
          5572 rows × 5 columns
          1. Data Cleaning
In [3]: 1 df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 5572 entries, 0 to 5571
          Data columns (total 5 columns):
                              Non-Null Count Dtype
           # Column
           0
              v1
                              5572 non-null
                                                object
                              5572 non-null
           1
               v2
                                                object
               Unnamed: 2 50 non-null
                                                object
               Unnamed: 3 12 non-null
               Unnamed: 4 6 non-null
          dtypes: object(5)
          memory usage: 217.8+ KB
In [4]:
            1 # droping columns
              df.drop(columns = ['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'],inplace=True )
Out[4]:
                                                            v2
                 ham
                          Go until jurong point, crazy.. Available only ...
                                         Ok lar... Joking wif u oni..

    ham

             2 spam Free entry in 2 a wkly comp to win FA Cup fina..
                       U dun say so early hor... U c already then say..
                         Nah I don't think he goes to usf, he lives aro..
           5567 spam
                       This is the 2nd time we have tried 2 contact u..
                                Will i_ b going to esplanade fr home?
                        Pity, * was in mood for that. So...any other s...
           5569
                 ham
                 ham The guy did some bitching but I acted like i'd...
           5570
           5571 ham
                                           Rofl. Its true to its name
          5572 rows × 2 columns
            1 # renaming columns
               df.rename(columns = {'v1':'target','v2':'text'},inplace=True)
            3 df
Out[5]:
                 target
             0
                 ham
                          Go until jurong point, crazy.. Available only ...
                                          Ok lar... Joking wif u oni...
                 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...
                        Nah I don't think he goes to usf, he lives aro...
           5567 spam This is the 2nd time we have tried 2 contact u...
           5568
                 ham
                                Will i b going to esplanade fr home?
           5569
                         Pity, * was in mood for that. So...any other s...
                        The guy did some bitching but I acted like i'd...
           5570
                 ham
           5571 ham
                                           Rofl. Its true to its name
          5572 rows × 2 columns
In [6]: | 1 | # changing ham = 0 and spam = 1
            2 | df['target'] = encoder.fit_transform(df['target'])
            3 df.head()
Out[6]:
                 0
          0
                       Go until jurong point, crazy.. Available only ...
                0
                                      Ok lar... Joking wif u oni...
                 1 Free entry in 2 a wkly comp to win FA Cup fina...
                 0 U dun say so early hor... U c already then say...
                 0 Nah I don't think he goes to usf, he lives aro...
```

In [ ]: 1

import numpy as np
import pandas as pd

In [69]:

```
In [7]:
         1 #missing values or not
          2 df.isnull().sum()
Out[7]: target
        text
         dtype: int64
          1 # check for duplicate values
In [8]:
          2 df.duplicated().sum()
Out[8]: 403
         # removing duplicates
df = df.drop_duplicates(keep='first')
In [9]:
          3 df.duplicated().sum()
Out[9]: 0
In [10]: 1 df.shape
Out[10]: (5169, 2)
        2. EDA
In [11]: 1 df['target'].value_counts()
Out[11]: 0
            4516
        Name: target, dtype: int64
In [12]:
          plt.pie(df['target'].value_counts(),labels=['ham','spam'],autopct="%0.2f")
          plt.show()
In [14]:
          1 # num of characters
            df['num_characters'] = df['text'].apply(len)
          3 df.head()
        C:\Users\Admin\AppData\Local\Temp/ipykernel_21892/4041876216.py:2: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas.docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexi
         ng.html#returning-a-view-versus-a-copy)
          df['num_characters'] = df['text'].apply(len)
Out[14]:
           target
              0
                   Go until jurong point, crazy.. Available only ...
                                                            111
         0
              0
                                Ok lar... Joking wif u oni...
                                                             29
              1 Free entry in 2 a wkly comp to win FA Cup fina...
                                                            155
                                                             49
         3
              0 U dun say so early hor... U c already then say...
               0 Nah I don't think he goes to usf, he lives aro...
                                                             61
In [15]: 1 # num of words
          2 df['num_words'] = df['text'].apply(lambda x:len(nltk.word_tokenize(x)))
        A value is trying to be set on a copy of a slice from a \mathsf{DataFrame}.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas.docs/stable/user_guide/indexi
         ng.html#returning-a-view-versus-a-copy)
           \label{eq:df['num_words'] = df['text'].apply(lambda x:len(nltk.word_tokenize(x)))} \\
Out[15]:
           target
                                              text num_characters num_words
         0
              0
                   Go until jurong point, crazy.. Available only ...
                                                            111
                                                                      24
                                Ok lar... Joking wif u oni...
                                                                       8
              1 Free entry in 2 a wkly comp to win FA Cup fina...
                                                            155
                                                                      37
         2
```

0 U dun say so early hor... U c already then say... 49 13 Nah I don't think he goes to usf, he lives aro... 15

In [16]: | 1 | # num of sentence df['num\_sentences'] = df['text'].apply(lambda x:len(nltk.sent\_tokenize(x))) 3 df.head()

> A value is trying to be set on a copy of a slice from a  $\mathsf{DataFrame}$ . Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas.pydata.pyd ng.html#returning-a-view-versus-a-copy)

df['num\_sentences'] = df['text'].apply(lambda x:len(nltk.sent\_tokenize(x))) Out[16]:

text num characters num words num sentences target Go until jurong point, crazy.. Available only ... 24 2 0 1 0 Ok lar... Joking wif u oni... 29 8 2 37 2 155 2 1 Free entry in 2 a wkly comp to win FA Cup fina... 0 U dun say so early hor... U c already then say... 3 49 13 0 Nah I don't think he goes to usf, he lives aro...

In [17]: 1 | df[['num\_characters', 'num\_words', 'num\_sentences']].describe()

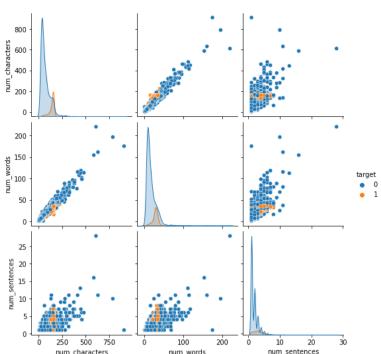
num\_characters num\_words num\_sentences 5169.000000 5169.000000 count 5169.000000 78.977945 18.453279 1.947185 1.362406 58.236293 13.324793 std 1.000000 min 2.000000 1.000000 25% 36.000000 9.000000 1.000000 50% 60.000000 15.000000 1.000000 75% 117.000000 26.000000 2.000000 910.000000 220.000000 28.000000

Out[17]:

```
In [18]:
            1 # ham msg desc
             df[df['target'] == 0][['num_characters','num_words','num_sentences']].describe()
Out[18]:
                      4516.000000 4516.000000
                                                   4516.000000
            count
                                     17.120903
            mean
                        70.459256
                                                      1.799601
                         56.358207
                                                      1.278465
                                                      1.000000
                         2.000000
                                      1.000000
             25%
                                                      1.000000
                        34.000000
                                      8.000000
             50%
                        52.000000
                                     13.000000
                                                      1.000000
                                     22.000000
                        90.000000
                                                      2.000000
             75%
                                                     28.000000
                       910.000000 220.000000
             max
In [19]:
             1 # spam msg desc
             2 df[df['target'] == 1][['num_characters','num_words','num_sentences']].describe()
Out[19]:
                   num_characters num_words num_sentences
            count
                       653.000000
                                   653.000000
                                                    653.000000
                       137.891271
                                                      1.483201
                        30.137753
                                     7.008418
              std
                         13.000000
                                     2.000000
                                                      1.000000
                                    25.000000
                       132.000000
                                                     2.000000
                       149.000000
                                                     3.000000
             50%
                       157.000000
                                    32.000000
                                                     4.000000
             75%
             max
                       224.000000
                                    46.000000
                                                     8.000000
In [20]:
             plt.figure(figsize=(12,6))
               sns.histplot(df[df['target'] == 0]['num_characters'])
             sns.histplot(df[df['target'] == 1]['num_characters'],color='red')
Out[20]: <AxesSubplot:xlabel='num_characters', ylabel='Count'>
              500
            Sount 300
              200
              100
                                                                                                         800
                                                                                    600
                                                               400
                                                                num_characters
In [21]:
            plt.figure(figsize=(12,6))
sns.histplot(df[df['target'] == 0]['num_words'])
sns.histplot(df[df['target'] == 1]['num_words'],color='red')
Out[21]: <AxesSubplot:xlabel='num_words', ylabel='Count'>
              600
              500
              400
           300
300
              200
              100
                                                                100
                                                                                      150
                                                                                                           200
                                                                  num_words
            plt.figure(figsize=(12,6))
sns.histplot(df[df['target'] == 0]['num_sentences'])
sns.histplot(df[df['target'] == 1]['num_sentences'],color='red')
In [22]:
Out[22]: <AxesSubplot:xlabel='num_sentences', ylabel='Count'>
              2500
              2000
              1500
              1000
               500
                                                                 15
num_sentences
                                                                                                           25
```

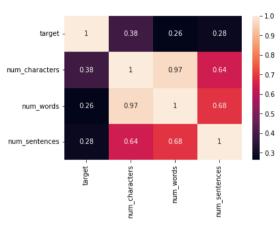
In [23]: 1 sns.pairplot(df,hue='target')

Out[23]: <seaborn.axisgrid.PairGrid at 0x21cc1b136d0>



In [24]: 1 sns.heatmap(df.corr(),annot=True)

### Out[24]: <AxesSubplot:>



# 3. Data Prepocessing

Lower Case

Tokenization

Removing special characters

Removing stop words and punctuations

## Steming

In [26]:

```
In [25]:
               def transform_text(text):
                     text = text.lower()
                     text = nltk.word_tokenize(text)
                     y = []
for i in text:
                          if i.isalnum():
                                y.append(i)
            10
11
12
                    text = y[:]
y.clear()
           13
14
15
16
17
18
19
20
                     for i in text:
                         if i not in {\sf stopwords.words('english')} and i not in {\sf string.punctuation:}
                              y.append(i)
                    text = y[:]
y.clear()
                     for i in text:
                         y.append(ps.stem(i))
            21
            22
           23
                     return " ".join(y)
```

C:\Users\Admin\AppData\Local\Temp/ipykernel\_21892/1835954565.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas.pydata.pyd ng.html#returning-a-view-versus-a-copy)
df['transformed\_text'] = df['text'].apply(transform\_text)

df['transformed\_text'] = df['text'].apply(transform\_text)

Out[26]:	target		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
	5567	1	This is the 2nd time we have tried 2 contact $u$	161	35	4	2nd time tri 2 contact u pound prize 2 claim e
	5568	0	Will i_ b going to esplanade fr home?	37	9	1	b go esplanad fr home
	5569	0	Pity, * was in mood for that. Soany other s	57	15	2	piti mood suggest
	5570	0	The guy did some bitching but I acted like i'd	125	27	1	guy bitch act like interest buy someth els nex
	5571	0	Rofl. Its true to its name	26	7	2	rofl true name

5169 rows × 6 columns

```
In [27]: 1 # Spam wordcLoud
spam_wc = wc.generate(df[df['target'] == 1]['transformed_text'].str.cat(sep=" "))
plt.figure(figsize=(15,6))
plt.imshow(spam_wc)
```

### Out[27]: <matplotlib.image.AxesImage at 0x21cc5ccf6d0>

```
pleas call collect

win charge rington club

toneur award prise club

toneur award prise club

toneur award prise club

stop callect

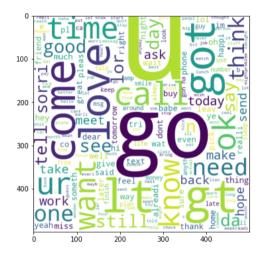
whose prise club

stop charce win

call identifit to de

call id
```

### Out[28]: <matplotlib.image.AxesImage at 0x21cc5c9d870>



In [29]: 1 df.head()

```
Out[29]:
                 target
                                                                     text num_characters num_words num_sentences
                                                                                                                                                          transformed_text
                             Go until jurong point, crazy.. Available only ...
                                                                                                                            2 go jurong point crazi avail bugi n great world...
                                                                                                                                                         ok lar joke wif u oni
                      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
                                                                                                                                free entri 2 wkli comp win fa cup final tkt 21...
                         U dun say so early hor... U c already then say...
                                                                                          49
                                                                                                        13
                                                                                                                                           u dun say earli hor u c alreadi say
                           Nah I don't think he goes to usf, he lives aro...
                                                                                          61
                                                                                                        15
                                                                                                                                         nah think goe usf live around though
```

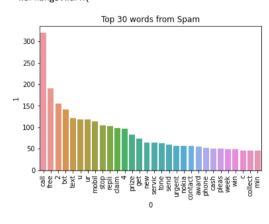
Spam corpus : 9939 Ham corpus : 35394

In [31]: 1 # Spam Corpus bar

```
# Spam Corpus barplot for top 30 words
sns.barplot(pd.DataFrame(Counter(spam_corpus).most_common(30))[0],pd.DataFrame(Counter(spam_corpus).most_common(30))[1]).set(title='Top 30 words from Spam')
plt.xticks(rotation="vertical")
plt.show()
```

C:\Python310\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and pas sing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

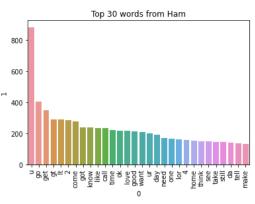


```
In [32]:

# Ham Corpus barplot for top 30 words
sns.barplot(pd.DataFrame(Counter(ham_corpus).most_common(30))[0],pd.DataFrame(Counter(ham_corpus).most_common(30))[1]).set(title='Top 30 words from Ham')
plt.xticks(rotation="vertical")
plt.show()
```

C:\Python310\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and pas sing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(



```
4. Model Building
 In [70]: 1 df.head()
 Out[70]:
              target
                                                       text num_characters num_words num_sentences
                                                                                                                         transformed_text
                  0
                        Go until jurong point, crazy.. Available only ...
                                                                      111
                                                                                  24
                                                                                                  2 go jurong point crazi avail bugi n great world..
                  0
                                      Ok lar... Joking wif u oni...
                                                                       29
                                                                                   8
                                                                                                  2
                                                                                                                         ok lar joke wif u oni
                                                                                  37
                  1 Free entry in 2 a wkly comp to win FA Cup fina...
                                                                      155
                                                                                                 2 free entri 2 wkli comp win fa cup final tkt 21...
                  0 U dun say so early hor... U c already then say...
                                                                       49
                                                                                  13
                                                                                                             u dun say earli hor u c alreadi say
                     Nah I don't think he goes to usf, he lives aro...
                                                                                                            nah think goe usf live around though
 In [71]:    1    X = tfidf.fit_transform(df['transformed_text']).toarray()
             2 X.shape
 Out[71]: (5169, 6708)
 In [72]:
            1 y = df['target'].values
 Out[72]: array([0, 0, 1, ..., 0, 0, 0])
 In [73]: 1 from sklearn.model_selection import train_test_split
             1 X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=2)
 In [75]:
                from sklearn.naive_bayes import GaussianNB,MultinomialNB,BernoulliNB
                from sklearn.metrics import accuracy_score,confusion_matrix,precision_score
 In [76]:
                gnb = GaussianNB()
                mnb = MultinomialNB()
             3 bnb = BernoulliNB()
               gnb.fit(X_train,y_train)
 In [77]:
               y_pred_1 = gnb.predict(X_test)
                print(f"Accuracy_score : {accuracy_score(y_test,y_pred_1)} ")
                print(f"Confusion_matrix : \n{confusion_matrix(y_test,y_pred_1)} ")
                print(f"Precision_score : {precision_score(y_test,y_pred_1)} ")
           Accuracy_score : 0.8762088974854932
           Confusion_matrix :
           [[793 103]
            [ 25 113]]
           Precision_score : 0.5231481481481481
 In [78]:
            1 mnb.fit(X_train,y_train)
             y_pred_2 = mnb.predict(X_test)
print(f"Accuracy_score : {accuracy_score(y_test,y_pred_2)} ")
print(f"Confusion_matrix : \n{confusion_matrix(y_test,y_pred_2)} ")
               print(f"Precision_score : {precision_score(y_test,y_pred_2)} ")
           Accuracy_score : 0.9593810444874274
           Confusion_matrix :
           [[896 0]
[ 42 96]]
           Precision_score : 1.0
In [100]:
             1 rfc.fit(X_train,y_train)
               y_pred_4 = rfc.predict(X_test)
               print(f"Accuracy_score : {accuracy_score(y_test,y_pred_4)} ")
                print(f"Confusion_matrix : \n{confusion_matrix(y_test,y_pred_4)} ")
             5 print(f"Precision_score : {precision_score(y_test,y_pred_4)} ")
           Accuracy_score : 0.9738878143133463
           Confusion_matrix :
            [ 27 111]]
           Precision_score : 1.0
 In [80]: 1 # tfidf --> MNB
 In [81]:
             1 from sklearn.linear_model import LogisticRegression
               from sklearn.svm import SVC
                from sklearn.naive_bayes import MultinomialNB
                from sklearn.tree import DecisionTreeClassifier
             5 from sklearn.neighbors import KNeighborsClassifier
             6 from sklearn.ensemble import RandomForestClassifier
             7 from sklearn.ensemble import AdaBoostClassifier
             8 from sklearn.ensemble import BaggingClassifier
             9 from sklearn.ensemble import ExtraTreesClassifier
            10 from sklearn.ensemble import GradientBoostingClassifier
            11 from xgboost import XGBClassifier
            svc = SVC(kernel='sigmoid', gamma=1.0)
knc = KNeighborsClassifier()
 In [82]:
               mnb = MultinomialNB()
               dtc = DecisionTreeClassifier(max_depth=5)
             5 lrc = LogisticRegression(solver='liblinear', penalty='l1')
             6 rfc = RandomForestClassifier(n_estimators=50, random_state=2)
            7 abc = AdaBoostClassifier(n_estimators=50, random_state=2)
8 bc = BaggingClassifier(n_estimators=50, random_state=2)
             9 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 [83]:
         1 clfs = {
               'SVC' : svc,
               'KN' : knc,
               'NB': mnb,
               'DT': dtc,
              'LR': lrc,
'RF': rfc,
'AdaBoost': abc,
         6
7
               'BgC': bc,
        10
               'ETC': etc,
        11
12
               'GBDT':gbdt,
               'xgb':xgb
        13 }
         def train_classifier(clf,X_train,y_train,X_test,y_test):
In [84]:
              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 [85]: 1 train_classifier(svc,X_train,y_train,X_test,y_test)
Out[85]: (0.9729206963249516, 0.9741379310344828)
In [86]:
         1 accuracy_scores = []
           precision_scores = []
         4 for name,clf in clfs.items():
              current_accuracy,current_precision = train_classifier(clf, X_train,y_train,X_test,y_test)
              print("For ",name)
         8
              print("Accuracy - ",current_accuracy)
print("Precision - ",current_precision)
        10
              print("/////////")
        12
        13
              accuracy_scores.append(current_accuracy)
        14
        15
              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
       Accuracy - 0.9352030947775629
Precision - 0.8380952380952381
        Accuracy - 0.9516441005802708
        Precision - 0.94
        For RF
       Accuracy - 0.9738878143133463
        Precision - 1.0
        For AdaBoost
       Accuracy - 0.9613152804642167
Precision - 0.9454545454545454
        Accuracy - 0.9584139264990329
Precision - 0.8625954198473282
        Accuracy - 0.9758220502901354
       Precision - 0.9829059829059829
        For GBDT
        Accuracy - 0.9526112185686654
        Precision - 0.9238095238095239
        For xgb
       Accuracy - 0.9690522243713733
        Precision - 0.9344262295081968
        In [87]:
         performance_df = pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy':accuracy_scores,
                                      'Precision':precision_scores}).sort_values('Precision',ascending=False)
           performance_df
Out[87]:
           Algorithm Accuracy Precision
```

KN 0.900387 1.000000 NB 0.959381

**9** GBDT 0.952611 0.923810 BgC 0.958414 0.862595

DT 0.935203 0.838095

10

7

3

0.973888 1.000000 ETC 0.975822 0.982906 SVC 0.972921 0.974138 AdaBoost 0.961315 0.945455 LR 0.951644 0.940000 xgb 0.969052 0.934426

1.000000

```
Out[89]:
              Algorithm variable
                                  value
                    KN Accuracy 0.900387
                   NB Accuracy 0.959381
           2
                    RF Accuracy 0.973888
                   ETC Accuracy 0.975822
                  SVC Accuracy 0.972921
           5 AdaBoost Accuracy 0.961315
                    LR Accuracy 0.951644
                   xgb Accuracy 0.969052
                 GBDT Accuracy 0.952611
           8
                   BgC Accuracy 0.958414
          10
                   DT Accuracy 0.935203
          11
                   KN Precision 1.000000
          12
                   NB Precision 1.000000
           13
          14
                   ETC Precision 0.982906
          15
                  SVC Precision 0.974138
              AdaBoost Precision 0.945455
          16
          17
                    LR Precision 0.940000
                   xgb Precision 0.934426
          18
          19
                 GBDT Precision 0.923810
          20
                   BgC Precision 0.862595
          21
                   DT Precision 0.838095
In [90]:
          1 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()
```

0.7 Accuracy AdaBoost LR xgb GBDT BgC -KN NB RF ETC Algorithm

performance\_df1 = pd.melt(performance\_df, id\_vars = "Algorithm")

In [68]:

In [89]:

performance\_df1

1 # model improve 2 # 1. Change the max\_features parameter of TfIdf

Rofl. Its true to its name

3 **df** 

Out[68]: target text num\_characters num\_words num\_sentences transformed\_text 111 24 Go until jurong point, crazy.. Available only  $\dots$ 2 go jurong point crazi avail bugi n great world... 1 Ok lar... Joking wif u oni... 29 8 2 ok lar joke wif u oni 1 Free entry in 2 a wkly comp to win FA Cup fina... 2 free entri 2 wkli comp win fa cup final tkt 21... 0 U dun say so early hor... U c already then say... 49 13 u dun say earli hor u c alreadi say 61 15 0 Nah I don't think he goes to usf, he lives aro... nah think goe usf live around though 5567 1 This is the 2nd time we have tried 2 contact u... 161 4 2nd time tri 2 contact u pound prize 2 claim e... 9 1 37 b go esplanad fr home 5568 0 Will I b going to esplanade fr home? 5569 0 Pity, \* was in mood for that. So...any other s... 57 piti mood suggest 0 The guy did some bitching but I acted like i'd... 1 guy bitch act like interest buy someth els nex...

7

0 5169 rows × 6 columns

5571

In [67]: 1 | temp\_df = pd.DataFrame({'Algorithm':clfs.keys(), 'Accuracy\_max\_ft\_3000':accuracy\_scores, 'Precision\_max\_ft\_3000':precision\_scores}).sort\_values('Precision\_max\_ft\_3000', ascending=False) 2 temp\_df

rofl true name

Out[67]: Algorithm Accuracy\_max\_ft\_3000 Precision\_max\_ft\_3000 0.905222 1.000000 NB 0.970986 1.000000 2 0.974855 0.982759 0 SVC 0.975822 0.974790 8 ETC 0.974855 0.974576 LR 0.958414 0.970297 4 10 xgb 0.971954 0.943089 0.960348 0.929204 6 AdaBoost 0.920000 **GBDT** 0.947776 9 0.957447 0.867188 7 BgC 3 DT 0.927466 0.811881

```
temp_df
 Out[66]:
               Algorithm Accuracy_scaling Precision_scaling
                                 0.905222
                                                 1.000000
                     NB
                                 0.970986
                                                 1.000000
                     RF
                                 0.974855
                                                 0.982759
                                                 0.974790
                    SVC
                                 0.975822
                    ETC
                                 0.974855
                                                 0.974576
                     LR
                                 0.958414
                                                 0.970297
            10
                     xgb
                                 0.971954
                                                 0.943089
                                 0.960348
                                                 0.929204
                AdaBoost
                   GBDT
                                 0.947776
                                                 0.920000
                    BgC
                                 0.957447
                                                 0.867188
                     DT
                                 0.927466
                                                 0.811881
 In [62]:
             1 new_df =
                         performance_df.merge(temp_df,on='Algorithm')
               new_df
 Out[62]:
                Algorithm
                         Accuracy
                                   Precision Accuracy_scaling Precision_scaling
             0
                     KN
                          0.905222
                                    1.000000
                                                    0.905222
                                                                    1.000000
                     NB 0.970986
                                   1.000000
                                                    0.970986
                                                                    1.000000
                                                    0.974855
                                                                    0.982759
                                                    0.975822
                    SVC
                                   0.974790
                                                                    0.974790
                         0.975822
                    ETC
                          0.974855
                                   0.974576
                                                    0.974855
                                                                    0.974576
                          0.958414
                                   0.970297
                                                    0.958414
                                                                    0.970297
                          0.971954
                                                    0.971954
                                                                    0.943089
                     xgb
                AdaBoost
                         0.960348
                                   0.929204
                                                    0.960348
                                                                    0.929204
                   GBDT
                         0.947776
                                   0.920000
                                                    0.947776
                                                                    0.920000
                    BgC
                         0.957447
                                                    0.957447
                                                                    0.867188
            10
                                                    0.927466
                                                                    0.811881
                     DT 0.927466 0.811881
 In [57]:
             new_df_scaled = new_df.merge(temp_df,on='Algorithm')
                temp_df = pd.DataFrame({'Algorithm':clfs.keys(), 'Accuracy_num_chars':accuracy_scores, 'Precision_num_chars':precision_scores}).sort_values('Precision_num_chars', ascending=False)
 In [58]:
 In [59]:
               new_df_scaled.merge(temp_df,on='Algorithm')
 Out[59]:
                                                                                                                                         Precision_num_chars
                Algorithm Accuracy
                                  Precision Accuracy_scaling_x Precision_scaling_x Accuracy_scaling_y Precision_scaling_y
                                                                                                                     Accuracy_num_chars
                          0.905222
                                    1.000000
                                                     0.905222
                                                                        1.000000
                                                                                          0.905222
                                                                                                             1.000000
                                                                                                                                0.905222
                                                                                                                                                    1.000000
                                                                                          0.970986
                         0.970986
                                   1.000000
                                                     0.970986
                                                                        1.000000
                                                                                                             1.000000
                                                                                                                                0.970986
                                                                                                                                                    1.000000
                     NB
                     RF
                         0.974855
                                   0.982759
                                                     0.974855
                                                                        0.982759
                                                                                          0.974855
                                                                                                             0.982759
                                                                                                                                0.974855
                                                                                                                                                    0.982759
                    SVC 0.975822
                                                      0.975822
                                                                        0.974790
                                                                                          0.975822
                                                                                                             0.974790
                                                                                                                                0.975822
                                                                                                                                                    0.974790
                                                                        0.974576
                                                                                          0.974855
                                                                                                                                0.974855
                    ETC 0.974855
                                   0.974576
                                                     0.974855
                                                                                                             0.974576
                                                                                                                                                    0.974576
                     LR 0.958414
                                   0.970297
                                                     0.958414
                                                                        0.970297
                                                                                          0.958414
                                                                                                             0.970297
                                                                                                                                0.958414
                                                                                                                                                    0.970297
                          0.971954
                                                     0.971954
                                                                        0.943089
                                                                                          0.971954
                                                                                                             0.943089
                                                                                                                                0.971954
                                                                                                                                                    0.943089
                     xgb
                                                                        0.929204
                                                                                          0.960348
                                                                                                             0.929204
                                                                                                                                0.960348
                                                                                                                                                    0.929204
                          0.960348
                                   0.929204
                   GBDT 0.947776
                                                     0.947776
                                                                        0.920000
                                                                                          0.947776
                                                                                                             0.920000
                                                                                                                                0.947776
                                                                                                                                                    0.920000
                                   0.920000
                    BgC 0.957447 0.867188
                                                     0.957447
                                                                        0.867188
                                                                                          0.957447
                                                                                                             0.867188
                                                                                                                                0.957447
                                                                                                                                                    0.867188
            10
                     DT 0.927466 0.811881
                                                     0.927466
                                                                        0.811881
                                                                                          0.927466
                                                                                                             0.811881
                                                                                                                                0.927466
                                                                                                                                                    0.811881
 In [91]:
            1 # Voting Classifier
                knc = KNeighborsClassifier()
                     = MultinomialNB()
                rfc = RandomForestClassifier(n_estimators=50, random_state=2)
                from sklearn.ensemble import VotingClassifier
 In [92]:
             voting = VotingClassifier(estimators=[('kn', knc), ('nb', mnb), ('rf', rfc)],voting='soft')
 In [93]:
             voting.fit(X_train,y_train)
 Out[93]: VotingClassifier(estimators=[('kn', KNeighborsClassifier()),
                                           ('nb', MultinomialNB()),
                                           ('rf'
                                            RandomForestClassifier(n_estimators=50,
                                                                      random_state=2))],
                              voting='soft')
             1 y_pred = voting.predict(X_test)
                print("Accuracy", accuracy_score(y_test,y_pred))
print("Precision", precision_score(y_test,y_pred))
           Accuracy 0.9410058027079303
           Precision 1.0
 In [95]:
             1 # Applying stacking
                estimators=[('kn', knc), ('nb', mnb), ('rf', rfc)]
             3 final_estimator=RandomForestClassifier()
 In [96]:
             1 from sklearn.ensemble import StackingClassifier
             1 clf = StackingClassifier(estimators=estimators, final_estimator=final_estimator)
 In [97]:
 In [98]:
               clf.fit(X_train,y_train)
               y_pred = clf.predict(X_test)
               print("Accuracy",accuracy_score(y_test,y_pred))
             4 print("Precision", precision_score(y_test,y_pred))
           Accuracy 0.971953578336557
           Precision 0.8865248226950354
In [102]:
             1 import pickle
                pickle.dump(tfidf,open('vectorizer.pkl','wb'))
               pickle.dump(rfc,open('model.pkl','wb'))
 In [ ]: 1
 In [ ]: 1
```

temp\_df = pd.DataFrame({'Algorithm':clfs.keys(),'Accuracy\_scaling':accuracy\_scores,'Precision\_scaling':precision\_scores}).sort\_values('Precision\_scaling', ascending=False)

In [66]: