

```
In [1]: import numpy as np
import pandas as pd
```

```
C:\Users\Prave\anaconda3\lib\site-packages\pandas\core\arrays\masked.py:60: UserWarning: Pandas requires version '1.3.6' or newer of 'bottleneck' (version '1.3.5' currently installed).
  from pandas.core import (
```

```
In [2]: df = pd.read_csv("C://Users//Prave//Downloads//spam.csv", encoding='ISO-8859-1')
```

```
In [3]: #df
```

```
In [4]: #1 . data clearning
#2. EDA
#3. text prepossesing
#4. model building
#5. evaluation
#6. improvement
# apply to out data set.
```

1 . data clearning

```
In [5]: # Remove the unwanted columns
df.drop(columns=['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], inplace=True)
```

```
In [6]: #df.head()
```

```
In [7]: df.rename(columns={"v1": 'target', 'v2': 'text'}, inplace=True)
```

```
In [8]: #df.head()
```

```
In [9]: df.shape
```

```
Out[9]: (5572, 2)
```

```
In [10]: from sklearn.preprocessing import LabelEncoder
encoder = LabelEncoder()
```

```
In [11]: df['target'] = encoder.fit_transform(df['target'])
```

```
In [12]: #df.head()
```

```
In [13]: # missing values
df.isnull().sum()
```

```
Out[13]: target      0
text          0
dtype: int64
```

```
In [14]: # check for duplicate
df.duplicated().sum()
```

```
Out[14]: 403
```

```
In [15]: # remove duplicates
df = df.drop_duplicates(keep='first')
```

```
In [16]: df.duplicated().sum()
```

```
Out[16]: 0
```

```
In [17]: df.shape
```

```
Out[17]: (5169, 2)
```

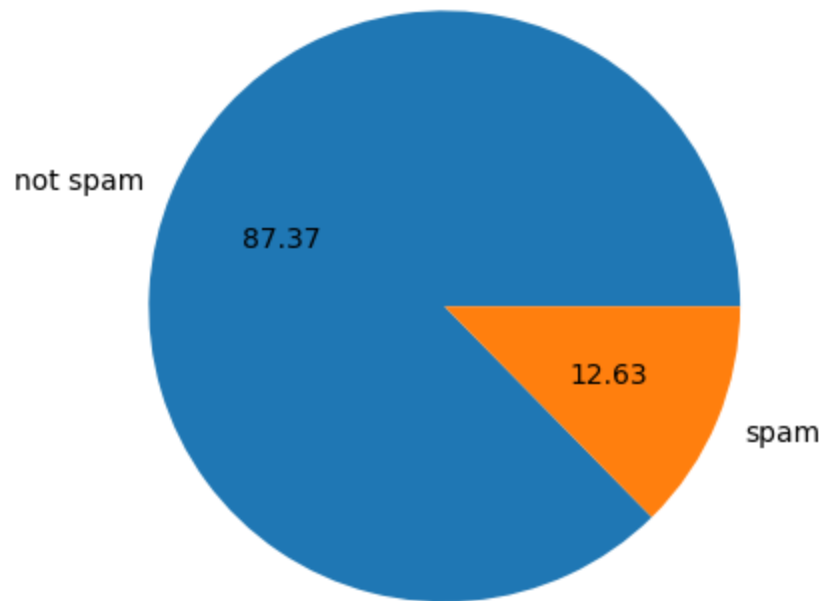
2 EDA

```
In [18]: df['target'].value_counts()
```

```
Out[18]: target
0      4516
1       653
Name: count, dtype: int64
```

```
In [19]: import matplotlib.pyplot as plt
```

```
In [20]: plt.pie(df['target'].value_counts(), labels=['not spam', 'spam'], autopct="%0.2f")
plt.show()
```



```
In [21]: # data is imbalanced
```

```
In [22]: !pip install nltk
import nltk
```

```
Requirement already satisfied: nltk in c:\users\prave\anaconda3\lib\site-packages (3.7)
Requirement already satisfied: tqdm in c:\users\prave\anaconda3\lib\site-packages (from nltk) (4.64.1)
Requirement already satisfied: joblib in c:\users\prave\anaconda3\lib\site-packages (from nltk) (1.1.1)
Requirement already satisfied: regex>=2021.8.3 in c:\users\prave\anaconda3\lib\site-packages (from nltk) (2022.7.9)
Requirement already satisfied: click in c:\users\prave\anaconda3\lib\site-packages (from nltk) (8.0.4)
Requirement already satisfied: colorama in c:\users\prave\anaconda3\lib\site-packages (from click->nltk) (0.4.6)
```

```
In [23]: nltk.download('punkt')
```

```
[nltk_data] Downloading package punkt to
[nltk_data] C:\Users\Prave\AppData\Roaming\nltk_data...
[nltk_data] Package punkt is already up-to-date!
```

```
Out[23]: True
```

```
In [24]: df['num_characters'] = df['text'].apply(len)
```

```
In [25]: #df.head()
```

```
In [26]: # num of words
df['num_words'] = df['text'].apply(lambda x:len(nltk.word_tokenize(x)))
```

```
In [27]: #df.head()
```

```
In [28]: df['num_sentences'] = df['text'].apply(lambda x:len(nltk.sent_tokenize(x)))
```

```
In [29]: #df.head()
```

```
In [30]: #summary stats
df[['num_characters', 'num_words', 'num_sentences']].describe()
```

```
Out[30]:
```

	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

```
In [31]: #only not spam
df[df['target']==0][['num_characters', 'num_words', 'num_sentences']].describe()
```

```
Out[31]:
```

	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
25%	34.000000	8.000000	1.000000
50%	52.000000	13.000000	1.000000
75%	90.000000	22.000000	2.000000
max	910.000000	220.000000	28.000000

```
In [32]: #only not spam
df[df['target']==1][['num_characters', 'num_words', 'num_sentences']].describe()
```

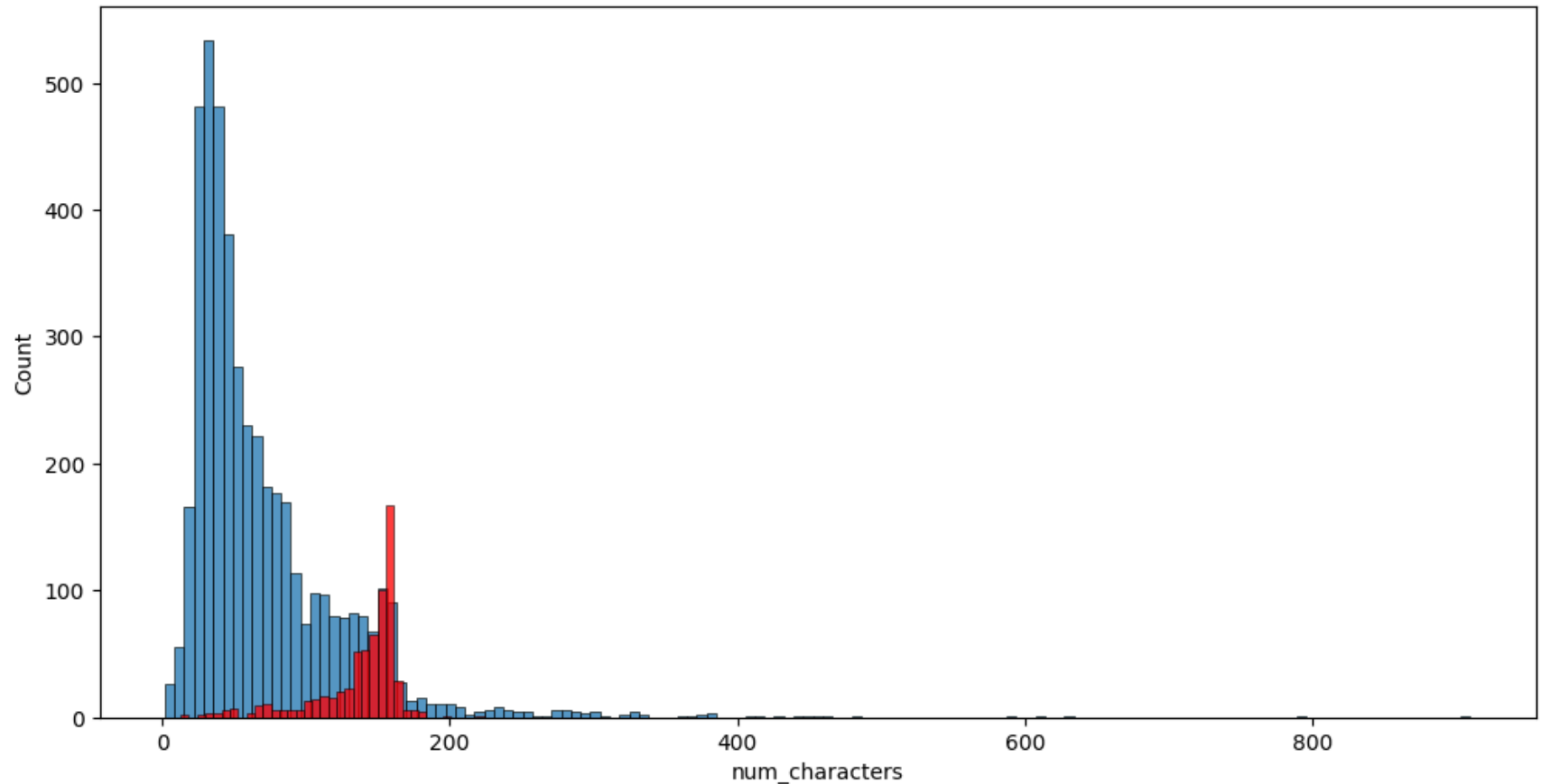
```
Out[32]:
```

	num_characters	num_words	num_sentences
count	653.000000	653.000000	653.000000
mean	137.891271	27.667688	2.967841
std	30.137753	7.008418	1.483201
min	13.000000	2.000000	1.000000
25%	132.000000	25.000000	2.000000
50%	149.000000	29.000000	3.000000
75%	157.000000	32.000000	4.000000
max	224.000000	46.000000	8.000000

```
In [33]: # histogram
import seaborn as sns
```

```
In [34]: plt.figure(figsize=(12,6))
sns.histplot(df[df['target']==0]['num_characters'])
sns.histplot(df[df['target']==1]['num_characters'],color='red')
```

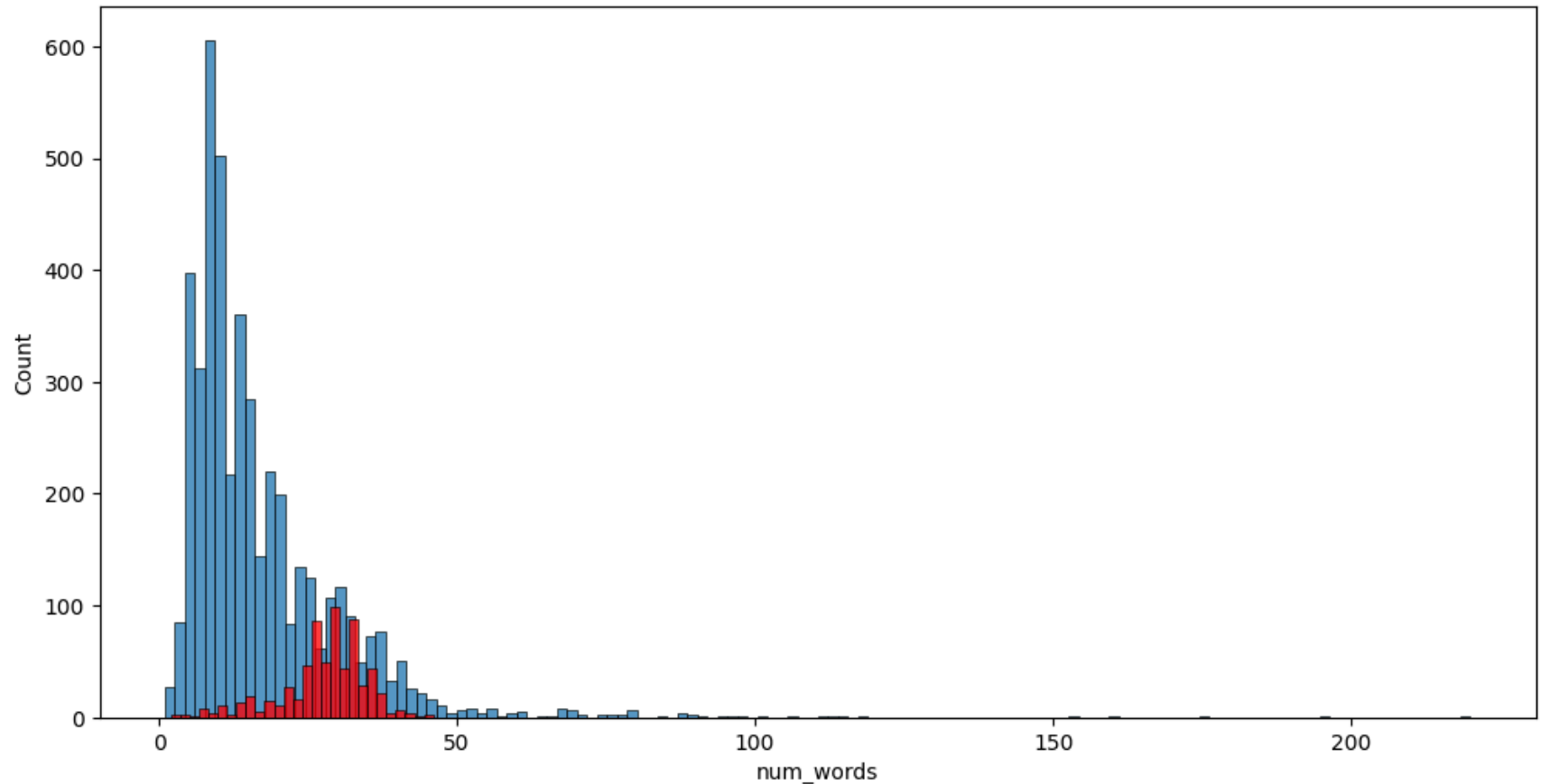
```
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
  with pd.option_context('mode.use_inf_as_na', True):
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
  with pd.option_context('mode.use_inf_as_na', True):
Out[34]: <Axes: xlabel='num_characters', ylabel='Count'>
```



```
In [35]: plt.figure(figsize=(12,6))
sns.histplot(df[df['target']==0]['num_words'])
sns.histplot(df[df['target']==1]['num_words'],color='red')
```

```
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
```

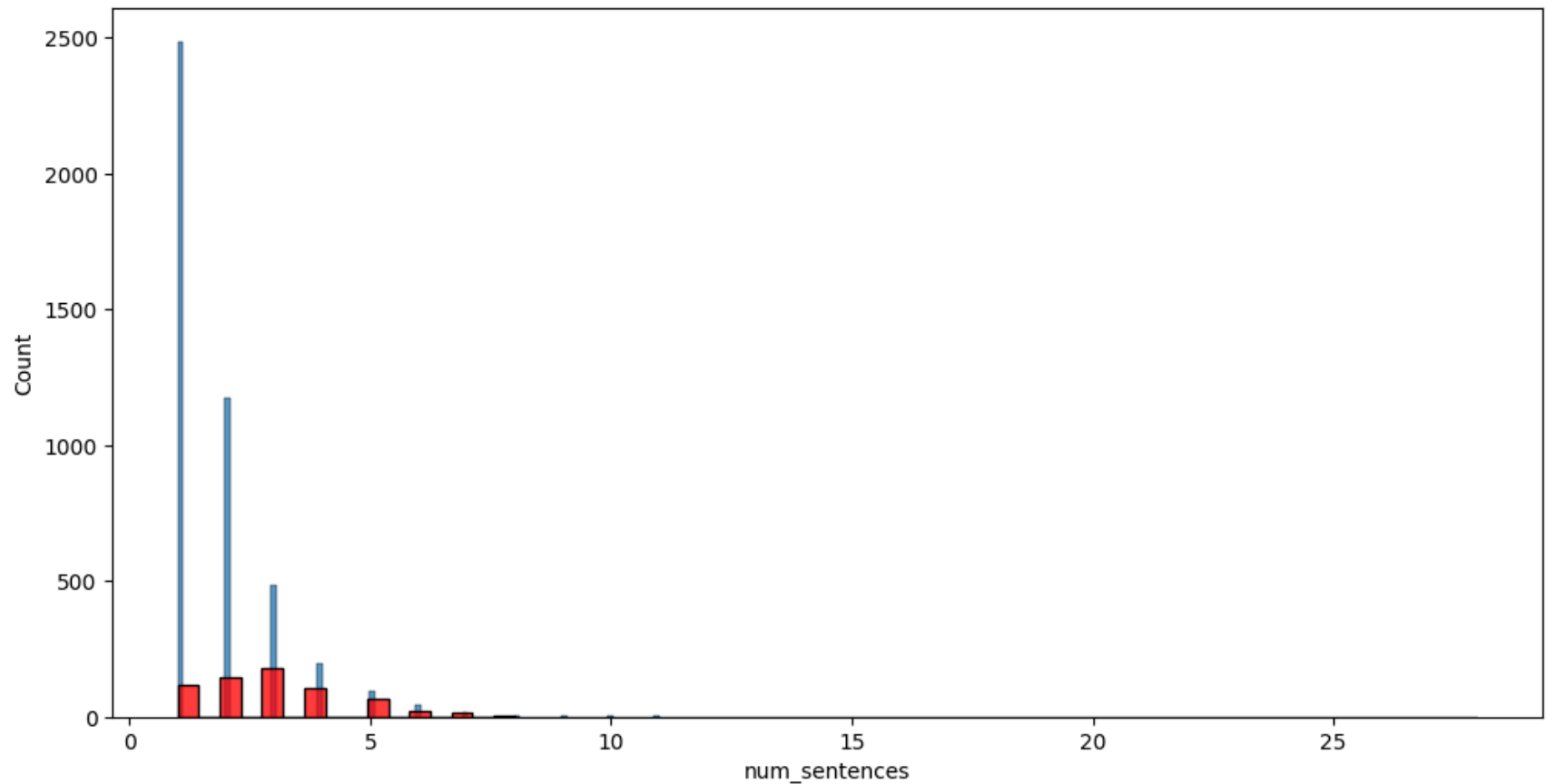
Out[35]: <Axes: xlabel='num_words', ylabel='Count'>



```
In [36]: plt.figure(figsize=(12,6))
sns.histplot(df[df['target']==0]['num_sentences'])
sns.histplot(df[df['target']==1]['num_sentences'],color='red')
```

```
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
```

Out[36]: <Axes: xlabel='num_sentences', ylabel='Count'>

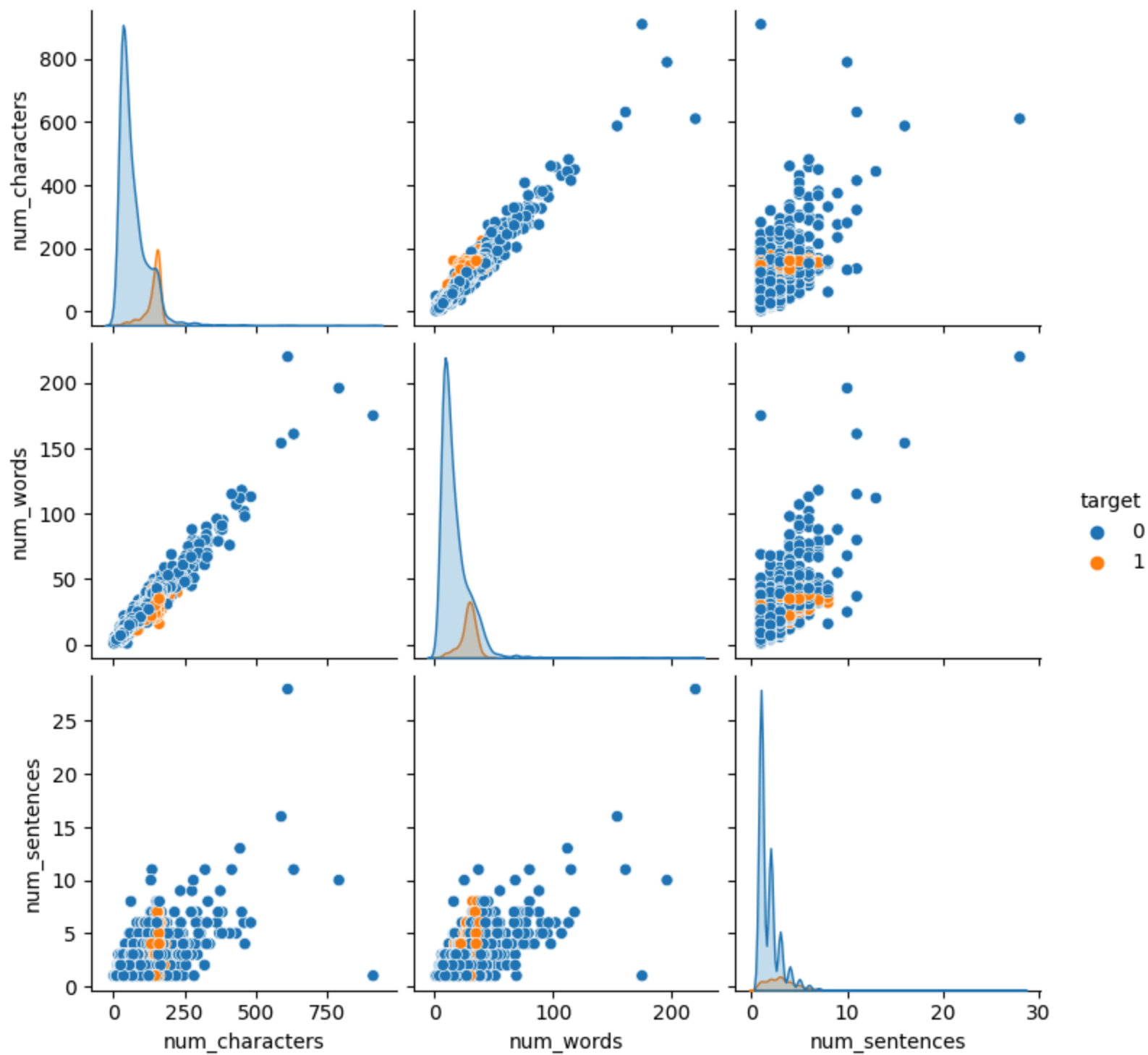


```
In [37]: # relation
sns.pairplot(df, hue='target')
```

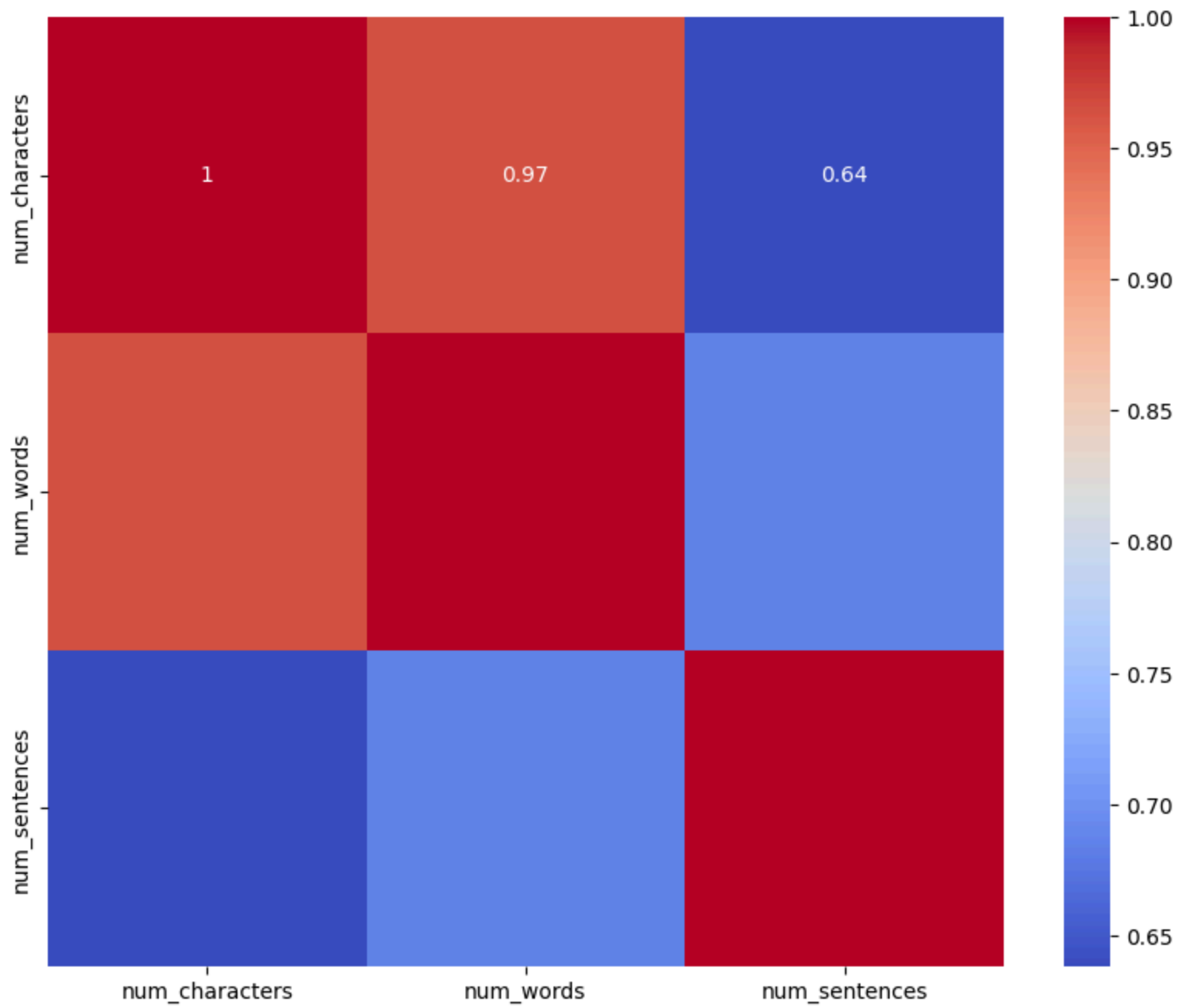


```
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1075: FutureWarning: When grouping with a length-1 list-
like, you will need to pass a length-1 tuple to get_group in a future version of pandas. Pass `(name,)` instead of `nam
e` to silence this warning.
    data_subset = grouped_data.get_group(pd_key)
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1075: FutureWarning: When grouping with a length-1 list-
like, you will need to pass a length-1 tuple to get_group in a future version of pandas. Pass `(name,)` instead of `nam
e` to silence this warning.
    data_subset = grouped_data.get_group(pd_key)
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
C:\Users\Prave\anaconda3\lib\site-packages\seaborn\_oldcore.py:1075: FutureWarning: When grouping with a length-1 list-
like, you will need to pass a length-1 tuple to get_group in a future version of pandas. Pass `(name,)` instead of `nam
e` to silence this warning.
    data_subset = grouped_data.get_group(pd_key)
<seaborn.axisgrid.PairGrid at 0x1e59891d720>
```

Out[37]:



```
In [38]: # Compute the correlation matrix, specifying numeric_only
        #!pip install --upgrade pandas
        df_numeric = df.select_dtypes(include=['float64', 'int64'])
        plt.figure(figsize=(10,8))
        sns.heatmap(df_numeric.corr(), annot=True, cmap='coolwarm')
        plt.show()
```



text preprocessing

```
In [39]: #Lower case
#Tokenization
#Removing special characters
#removing stop words and punctuations
# stemming
```

```
In [40]: # Download the stopwords resource
nlk.download('stopwords')
from nltk.corpus import stopwords
import string
stopwords.words('english')
from nltk.stem.porter import PorterStemmer

ps = PorterStemmer()
```

```
[nltk_data] Downloading package stopwords to
[nltk_data] C:\Users\Prave\AppData\Roaming\nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

```
In [41]: def transform_text(text):
# Convert text to lowercase and tokenize
text = text.lower()
text = nltk.word_tokenize(text)

# Use list comprehension to filter out non-alphanumeric words and stopwords
return [ps.stem(i) for i in text if i.isalnum() and i not in stopwords.words('english') and i not in string.punctua
```

```
In [42]: transform_text('do your like dancing and singing ?')
```

```
Out[42]: ['like', 'danc', 'sing']
```

```
In [43]: df['transformed_text'] = df['text'].apply(transform_text)
```

```
In [96]: # Save the transformed dataset using pickle
with open('transformed_data.pkl', 'wb') as file:
    pickle.dump(df, file)
```

```
In [44]: df.head()
```

Out[44]:	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, crazy, avail, bugi, n, gre...
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, fin...
3	0	U dun say so early hor... U c already then say...	49	13	1	[u, dun, say, earli, hor, u, c, already, 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 [45]: !pip install wordcloud

Requirement already satisfied: wordcloud in c:\users\prave\anaconda3\lib\site-packages (1.9.3)
Requirement already satisfied: pillow in c:\users\prave\anaconda3\lib\site-packages (from wordcloud) (9.4.0)
Requirement already satisfied: numpy>=1.6.1 in c:\users\prave\anaconda3\lib\site-packages (from wordcloud) (1.23.5)
Requirement already satisfied: matplotlib in c:\users\prave\anaconda3\lib\site-packages (from wordcloud) (3.9.1)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib->wordcloud) (1.0.5)
Requirement already satisfied: packaging>=20.0 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib->wordcloud) (22.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib->wordcloud) (3.0.9)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib->wordcloud) (4.25.0)
Requirement already satisfied: cycler>=0.10 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib->wordcloud) (0.11.0)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib->wordcloud) (2.8.2)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib->wordcloud) (1.4.4)
Requirement already satisfied: six>=1.5 in c:\users\prave\anaconda3\lib\site-packages (from python-dateutil>=2.7->matplotlib->wordcloud) (1.16.0)
```

```
In [46]: !pip install wordcloud matplotlib
from wordcloud import WordCloud
wc = WordCloud(width=500,height=500,min_font_size=10,background_color='white')
```

Requirement already satisfied: wordcloud in c:\users\prave\anaconda3\lib\site-packages (1.9.3)
Requirement already satisfied: matplotlib in c:\users\prave\anaconda3\lib\site-packages (3.9.1)
Requirement already satisfied: numpy>=1.6.1 in c:\users\prave\anaconda3\lib\site-packages (from wordcloud) (1.23.5)
Requirement already satisfied: pillow in c:\users\prave\anaconda3\lib\site-packages (from wordcloud) (9.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib) (3.0.9)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib) (4.25.0)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib) (1.0.5)
Requirement already satisfied: packaging>=20.0 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib) (22.0)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib) (1.4.4)
Requirement already satisfied: cycler>=0.10 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib) (0.11.0)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\prave\anaconda3\lib\site-packages (from matplotlib) (2.8.2)
Requirement already satisfied: six>=1.5 in c:\users\prave\anaconda3\lib\site-packages (from python-dateutil>=2.7->matplotlib) (1.16.0)

```
In [47]: # Ensure the 'transformed_text' column has only strings by converting non-string data
#df['transformed_text'] = df['transformed_text'].astype(str)
```

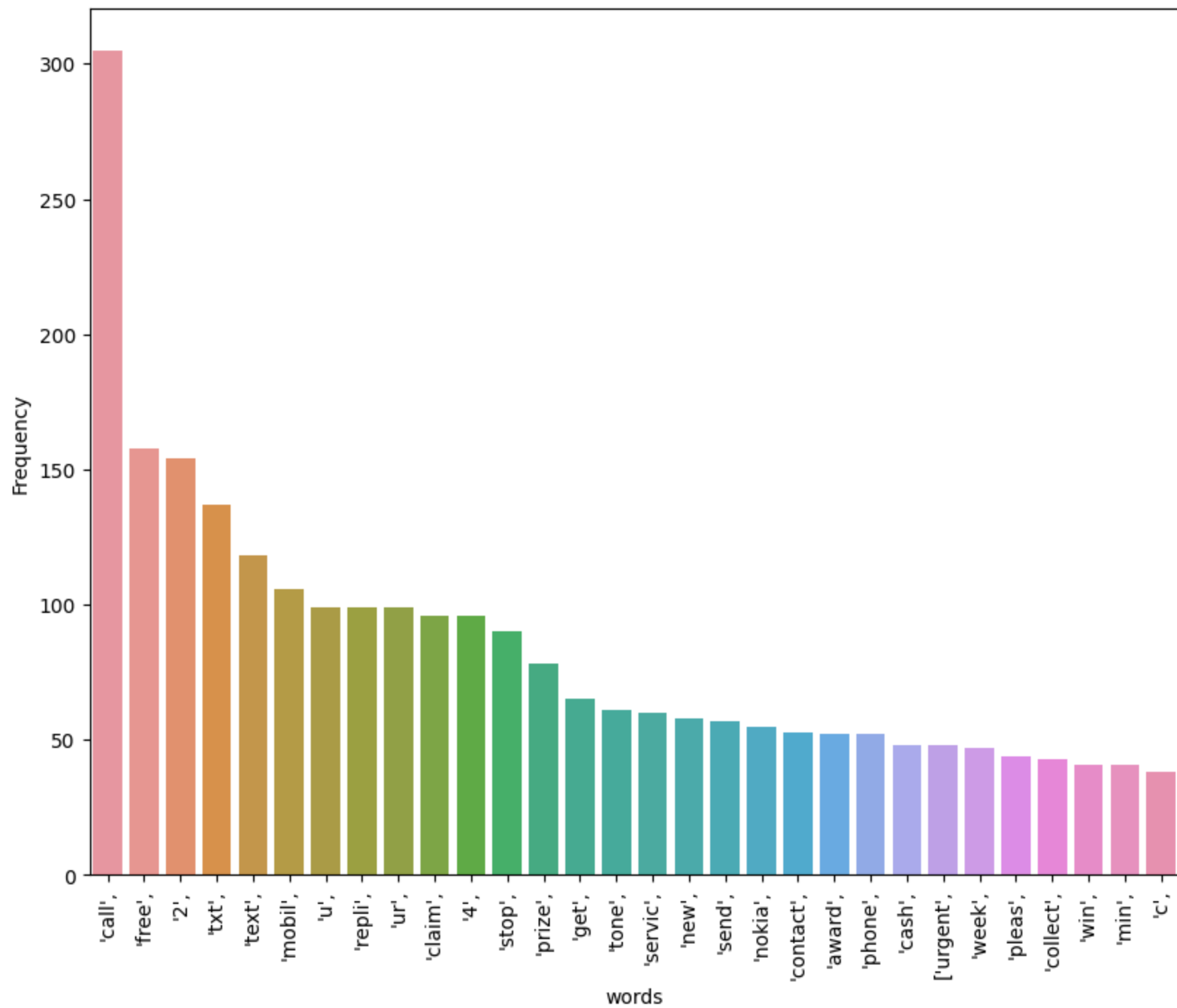
```
In [48]: df.loc[:, 'transformed_text'] = df['transformed_text'].astype(str)
spam_wc = wc.generate(df[df['target']==1]['transformed_text'].str.cat(sep=" "))
```

```
In [49]: # Plot the word cloud
# Plot the WordCloud
plt.figure(figsize=(10, 5))
plt.imshow(spam_wc, interpolation='bilinear')
plt.axis('off')
plt.show()
```



```
In [54]: from collections import Counter
common_words_df = pd.DataFrame(Counter(spam_corpus).most_common(30), columns=['words', 'Frequency'])
```

```
In [55]: # Plot the barplot using Seaborn
plt.figure(figsize=(10,8))
sns.barplot(x='words', y='Frequency', data=common_words_df)
plt.xticks(rotation=90) # Rotate x-axis labels for better readability
plt.show()
```



```
In [56]: df.head(5)
```

```
Out[56]:
```

	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', 'b...
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', ...
3	0	U dun say so early hor... U c already then say...	49	13	1	['u', 'dun', 'say', 'earli', 'hor', 'u', 'c', ...
4	0	Nah I don't think he goes to usf, he lives aro...	61	15	1	['nah', 'think', 'goe', 'usf', 'live', 'around...

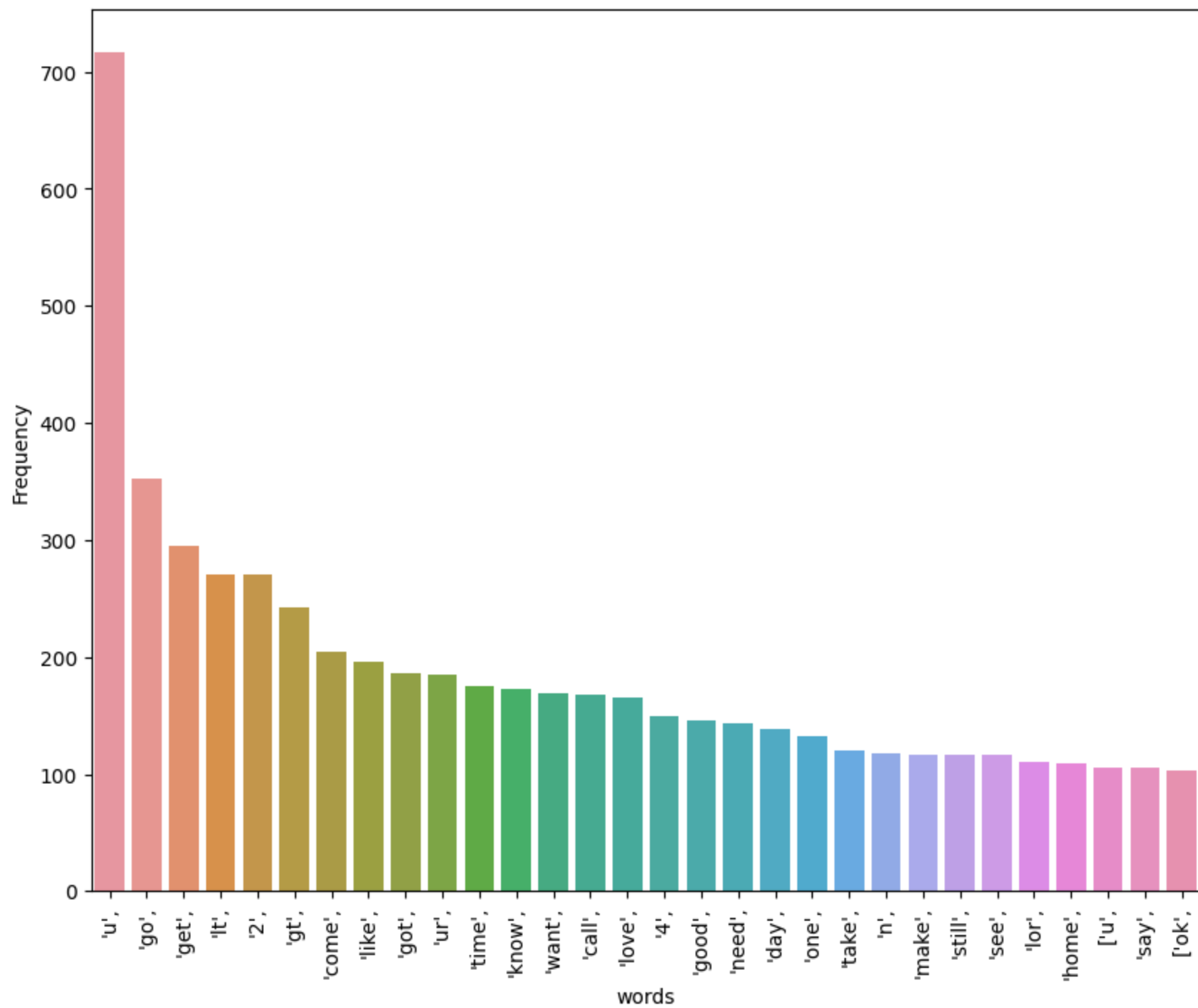
```
In [57]: #ham_corpus = []
#ham_corpus = df[df['target'] == 0]['transformed_text'].apply(lambda x: ' '.join(x)).tolist()
ham_corpus = [word for msg in df[df['target'] == 0]['transformed_text'].tolist() for word in msg.split()]
```

```
In [58]: print(len(ham_corpus))
```

35403

```
In [59]: from collections import Counter
common_words_df = pd.DataFrame(Counter(ham_corpus).most_common(30), columns=['words', 'Frequency'])
```

```
In [60]: # Plot the barplot using Seaborn
plt.figure(figsize=(10,8))
sns.barplot(x='words', y='Frequency', data=common_words_df)
plt.xticks(rotation=90) # Rotate x-axis labels for better readability
plt.show()
```



Model Building

```
In [61]: df.head()
```

```
Out[61]:
```

	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', 'b...
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', ...
3	0	U dun say so early hor... U c already then say...	49	13	1	['u', 'dun', 'say', 'earli', 'hor', 'u', 'c', ...
4	0	Nah I don't think he goes to usf, he lives aro...	61	15	1	['nah', 'think', 'goe', 'usf', 'live', 'around...

```
In [62]: from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
cv = CountVectorizer()
tfidf = TfidfVectorizer(max_features=3000)
```

```
In [63]: X = cv.fit_transform(df['transformed_text']).toarray()
```

```
In [64]: #from sklearn.preprocessing import MinMaxScaler
#scaler = MinMaxScaler()
#X = scaler.fit_transform(X)
```

```
In [65]: # appending the num of charc col to X
#X = np.hstack((X, df['num_characters'].values.reshape(-1,1)))
```

```
In [66]: X.shape
```

```
Out[66]: (5169, 6708)
```

```
In [67]: y = df['target'].values
```

```
In [68]: y
```

```
Out[68]: array([0, 0, 1, ..., 0, 0, 0])
```

```
In [69]: from sklearn.model_selection import train_test_split
        from sklearn.metrics import accuracy_score, confusion_matrix, precision_score
```

```
In [70]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=2)
```

```
In [71]: from sklearn.naive_bayes import GaussianNB, MultinomialNB, BernoulliNB
```

```
In [72]: gnb = GaussianNB()
        mnb = MultinomialNB()
        bnb = BernoulliNB()
```

```
In [73]: gnb.fit(X_train, y_train)
        y_pred_gnb = gnb.predict(X_test)
        print(accuracy_score(y_test, y_pred_gnb))
        print(confusion_matrix(y_test, y_pred_gnb))
        print(precision_score(y_test, y_pred_gnb))
```

```
0.8800773694390716
[[792 104]
 [ 20 118]]
0.5315315315315315
```

```
In [74]: mnb.fit(X_train, y_train)
        y_pred_mnb = mnb.predict(X_test)
        print(accuracy_score(y_test, y_pred_mnb))
        print(confusion_matrix(y_test, y_pred_mnb))
        print(precision_score(y_test, y_pred_mnb))
```

```
0.9642166344294004
[[871  25]
 [ 12 126]]
0.8344370860927153
```

```
In [75]: bnb.fit(X_train, y_train)
        y_pred_bnb = bnb.predict(X_test)
        print(accuracy_score(y_test, y_pred_bnb))
        print(confusion_matrix(y_test, y_pred_bnb))
        print(precision_score(y_test, y_pred_bnb))
```

```
0.9700193423597679
[[893   3]
 [ 28 110]]
0.9734513274336283
```

```
In [76]: # bernouli in performing well
```

```
In [77]: # Tfidf
```

```
In [78]: X = tfidf.fit_transform(df['transformed_text']).toarray()
```

```
In [79]: # appending the num of charc col to X
#X = np.hstack((X,df['num_characters'].values.reshape(-1,1)))
```

```
In [98]: # Save the TfidfVectorizer and X_tfidf using pickle
with open('tfidf_vectorizer.pkl', 'wb') as file:
    pickle.dump(tfidf, file)
with open('X_tfidf.pkl', 'wb') as file:
    pickle.dump(X, file)
```

```
In [80]: y = df['target'].values
```

```
In [99]: # Save the target variable
with open('target.pkl', 'wb') as file:
    pickle.dump(y, file)
```

```
In [81]: X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=2)
```

```
In [82]: gnb.fit(X_train,y_train)
y_pred_gnb = gnb.predict(X_test)
print(accuracy_score(y_test,y_pred_gnb))
print(confusion_matrix(y_test,y_pred_gnb))
print(precision_score(y_test,y_pred_gnb))
```

```
0.8694390715667312
[[788 108]
 [ 27 111]]
0.5068493150684932
```

```
In [83]: mnb.fit(X_train,y_train)
y_pred_mnb = mnb.predict(X_test)
print(accuracy_score(y_test,y_pred_mnb))
print(confusion_matrix(y_test,y_pred_mnb))
print(precision_score(y_test,y_pred_mnb))
```



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

```
In [84]: bnb.fit(X_train,y_train)
y_pred_bnb = bnb.predict(X_test)
print(accuracy_score(y_test,y_pred_bnb))
print(confusion_matrix(y_test,y_pred_bnb))
print(precision_score(y_test,y_pred_bnb))
```

```
0.9835589941972921
[[895   1]
 [ 16 122]]
0.991869918699187
```

```
In [100... # Save models using pickle
with open('gaussian_nb_model.pkl', 'wb') as file:
    pickle.dump(gnb, file)
with open('multinomial_nb_model.pkl', 'wb') as file:
    pickle.dump(mnb, file)
with open('bernoulli_nb_model.pkl', 'wb') as file:
    pickle.dump(bnb, file)
```

```
In [101... # Save predictions
with open('y_pred_gnb.pkl', 'wb') as file:
    pickle.dump(y_pred_gnb, file)
with open('y_pred_mnb.pkl', 'wb') as file:
    pickle.dump(y_pred_mnb, file)
with open('y_pred_bnb.pkl', 'wb') as file:
    pickle.dump(y_pred_bnb, file)
```

```
In [85]: # mnb performing well interms of precision with TfIdf
```

```
In [86]: # pipeline creation
```

```
In [87]: # Calculate accuracy scores
accuracy_gnb = accuracy_score(y_test, y_pred_gnb)
accuracy_mnb = accuracy_score(y_test, y_pred_mnb)
accuracy_bnb = accuracy_score(y_test, y_pred_bnb)
```

```
In [88]: # Calculate precision scores (set average to 'weighted' for multi-class classification)
precision_gnb = precision_score(y_test, y_pred_gnb, average='weighted')
```

```
precision_mnb = precision_score(y_test, y_pred_mnb, average='weighted')
precision_bnb = precision_score(y_test, y_pred_bnb, average='weighted')
```

```
In [89]: # Store models with their accuracy and precision
model_results = [
    ("GaussianNB", gnb, accuracy_gnb, precision_gnb),
    ("MultinomialNB", mnb, accuracy_mnb, precision_mnb),
    ("BernoulliNB", bnb, accuracy_bnb, precision_bnb)
]
```

```
In [90]: # Sort models by accuracy and precision
sorted_by_accuracy = sorted(model_results, key=lambda x: x[2], reverse=True) # Sort by accuracy
sorted_by_precision = sorted(model_results, key=lambda x: x[3], reverse=True) # Sort by precision
```

```
In [91]: # Get the best model by accuracy and precision
best_model_accuracy = sorted_by_accuracy[0]
best_model_precision = sorted_by_precision[0]
```

```
In [92]: # Print sorted results
print("Models sorted by Accuracy:")
for name, model, accuracy, precision in sorted_by_accuracy:
    print(f"Model: {name}, Accuracy: {accuracy}, Precision: {precision}")
```

```
Models sorted by Accuracy:
Model: BernoulliNB, Accuracy: 0.9835589941972921, Precision: 0.9836958370815164
Model: MultinomialNB, Accuracy: 0.9709864603481625, Precision: 0.971926423835803
Model: GaussianNB, Accuracy: 0.8694390715667312, Precision: 0.9054755995131818
```

```
In [93]: print("\nModels sorted by Precision:")
for name, model, accuracy, precision in sorted_by_precision:
    print(f"Model: {name}, Precision: {precision}, Accuracy: {accuracy}")
```

```
Models sorted by Precision:
Model: BernoulliNB, Precision: 0.9836958370815164, Accuracy: 0.9835589941972921
Model: MultinomialNB, Precision: 0.971926423835803, Accuracy: 0.9709864603481625
Model: GaussianNB, Precision: 0.9054755995131818, Accuracy: 0.8694390715667312
```

```
In [94]: # Print the best models
print(f"\nBest model by Accuracy: {best_model_accuracy[0]} with Accuracy: {best_model_accuracy[2]}")
print(f"Best model by Precision: {best_model_precision[0]} with Precision: {best_model_precision[3]}")
```

```
Best model by Accuracy: BernoulliNB with Accuracy: 0.9835589941972921
Best model by Precision: BernoulliNB with Precision: 0.9836958370815164
```

In [102...

```
# Save model results using pickle
with open('model_results.pkl', 'wb') as file:
    pickle.dump(model_results, file)

# Save sorted results by accuracy
with open('sorted_by_accuracy.pkl', 'wb') as file:
    pickle.dump(sorted_by_accuracy, file)

# Save sorted results by precision
with open('sorted_by_precision.pkl', 'wb') as file:
    pickle.dump(sorted_by_precision, file)

# Save best model by accuracy
with open('best_model_accuracy.pkl', 'wb') as file:
    pickle.dump(best_model_accuracy, file)

# Save best model by precision
with open('best_model_precision.pkl', 'wb') as file:
    pickle.dump(best_model_precision, file)
```

In []: