

Python Programming (Project Report)

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Spam Email Detection Using Data Visualisation

• **Aim**: The primary aim of this project is to leverage data visualization techniques to uncover patterns inherent in spam emails. By analyzing these patterns, we aim to:

Identify distinguishing characteristics of spam emails: This will involve exploring various email features within a dataset of labeled spam and ham emails. Data visualization will be used to identify trends and relationships between features such as:

- Word frequency: Analyze the frequency of specific words or phrases commonly found in spam emails (e.g., "free", "urgent", "win a million dollars").
- **Email length:** Visualize the distribution of email lengths across spam and legitimate emails.
- **Presence of urgency triggers:** Identify the use of words or phrases that create a sense of urgency, often used in phishing attempts.
- Special characters and symbols: Explore the prevalence of uncommon symbols or excessive use of punctuation in spam emails.

Concepts Used:

- Variables: Variables are used to store data. Examples are: df, styled df, values, etc.
- **Functions:** Functions are blocks of reusable code. Examples are: read_csv, head(), info(), isnull(), etc
- **Importing Library:** importing certain libraries like numpy, pandas, matplotlib.pyplot, seaborn, NLTK, wordcloud, etc.
- OOPs: Using Object Oriented Programming to create a parent class i.e WordDistributionPlot.

PROCEDURE:

Importing all libraries:

First of all in this jupyter file we create a cell in which we import all the necessary libraries required.

```
# Importing necessary libraries
import numbry as np  # For numerical operations
import pands as pd  # For date manipulation and analysis
import pands as pd  # For date manipulation and analysis
import matplottib.pyplet as plt  # For data visualization
%matplotlib inline

# Importing WordCloud for text visualization
from wordcloud import WordCloud

# Importing NLTK for natural language processing
import nltk
from nltk.corpus import stopwords  # For stopwords

# Downloading NLTK data
nltk.download('stopwords')  # Downloading stopwords data
nltk.download('all')

* Downloading tokenizer data
* Ntk.download('all')

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* Ntk.download('all')
```

Reading the data set:

We read our data set for the project and displaying starting data from the whole data set.



Data Cleaning:

Displaying the information of the data set

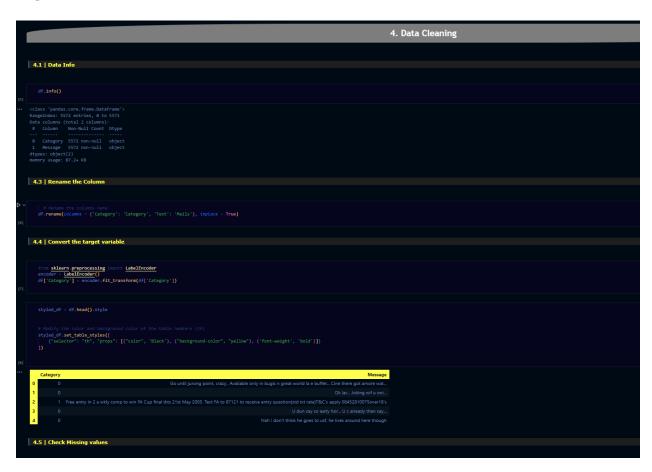
Renaming the column

Checking missing values

Checking duplicate values

Remove duplicate values

Shape of the dataset

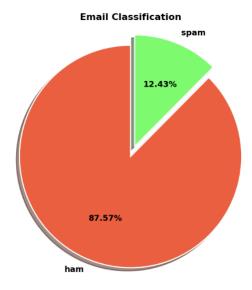


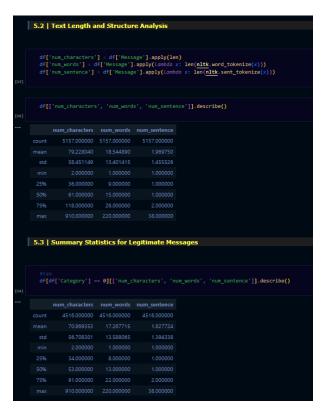
```
# 4.5 | Check Missing values

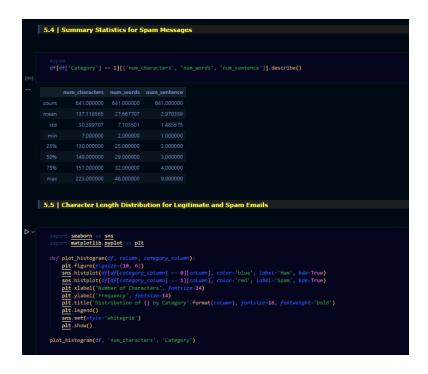
# ## Classing | Classing
```

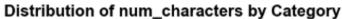
Data Analysis:

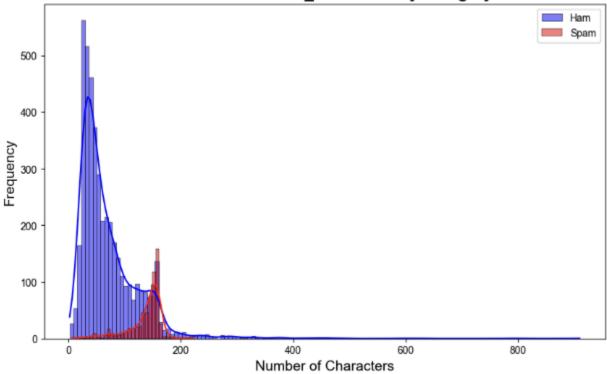
Performing the data analysis on the data set.



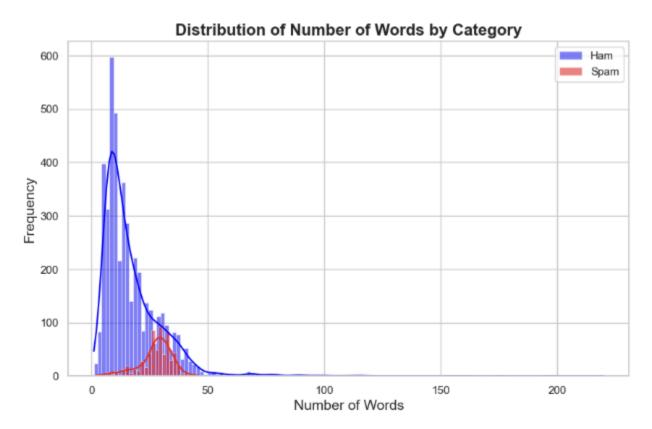








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| import seaborn as and import seaborn as seaborn as and import seaborn as se
```



5.7 | Pairplot for Data Visualization

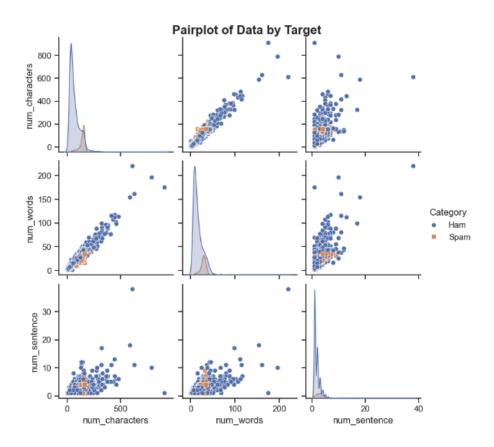
```
import seaborn as sns
import matplotlib.pyplot as plt

# Create a pairplot with custom styling
sns.set(style='ticks', color_codes=True)
g = sns.pairplot(df, hue='Category', diag_kind='kde', markers=["o", "s"])

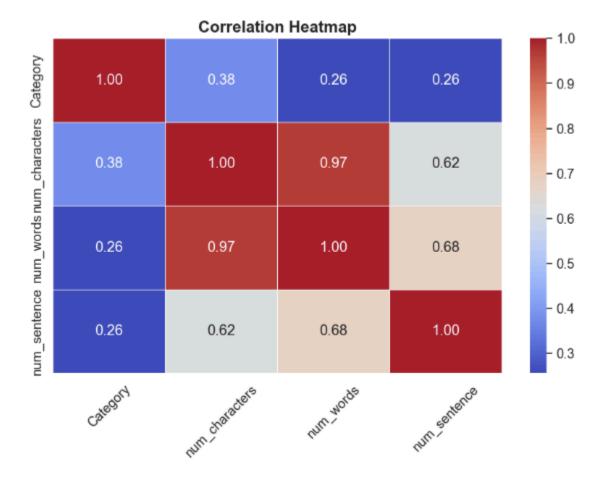
# Set a title for the pairplot
g.fig.suptitle("Pairplot of Data by Target", fontsize=16, fontweight='bold')
plt.subplots_adjust(top=0.95) # Adjust the position of the title

# Customize the legend
g._legend.set_title('Category')
for t, 1 in zip(g._legend.texts, ["Ham", "Spam"]):
    t.set_text(1)

# Show the pairplot
plt.show()
```



	5.8 Coorelation						
	<pre>df[['Category','num_characters', 'num_words', 'num_sentence']].corr()</pre>						
[22]							
•••		Category 1	num_characters	num_words	num_sentence		
	Category						
			0.966006	1.000000			
D ~							
	<pre>import seaborn as sns import matplotlib.pyplot as plt</pre>						
	import improving pyprot as pit						
	# Select the columns for the correlation matrix						
	<pre>correlation_matrix = df[['Category', 'num_characters', 'num_words', 'num_sentence']].corr()</pre>						
	<pre>plt.figure(figsize=(10, 6))</pre>						
	<pre>sns.set(font_scale=1.2) # Adjust font scale for better readability sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', linewidths=0.5, fmt=".2f")</pre>						
	<pre># Set a title for the heatmap plt.title("Correlation Heatmap", fontsize=16, fontweight='bold')</pre>						
	particular terresponding points are 10, juntaneight out)						
	<pre>plt.xticks(rotation=45)</pre>						
	plt.show()						



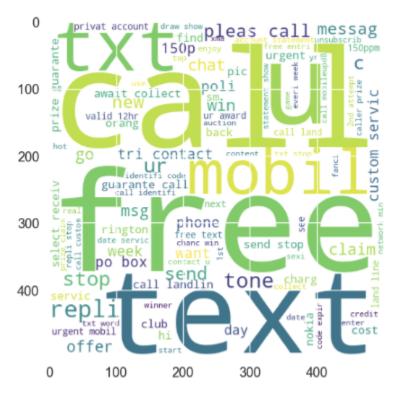
Data Preprocessing:

```
transform_text( to certal person point, crass, available only in logist in great serial in a buffet cite get acce wet.")

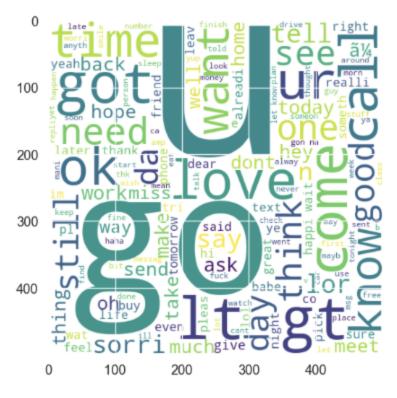
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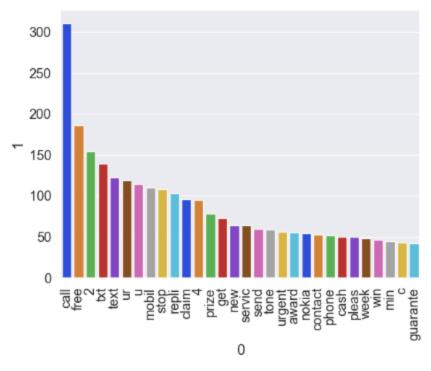
```
ham_wc = wc.generate(df[df['Category'] == 0]['transformed_text'].str.cat(sep = " "))
plt.figure(figsize = (15,6))
plt.imshow(ham_wc)
plt.show()
```



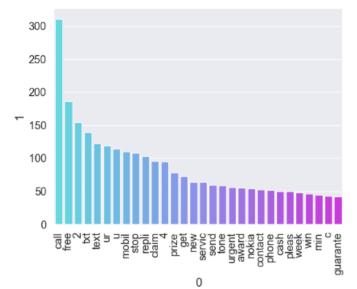
```
spam_carpos = []
for sentence in df[df['Category'] == 1]['transformed_text'].tolist():
    for word in sentence.split():
        spam_carpos.append(word)

from collections import Counter
filter_df = pd.DataFrame(Counter(spam_carpos).most_common(30))

sns.barplot(data = filter_df, x = filter_df[0], y = filter_df[1], palette = 'bright')
    plt.xticks(rotation = 90)
    plt.show()
```







Conclusion:

Conclusion: In our evaluation of various classification algorithms, we observed the following key insights:

- Support Vector Classifier (SVC) and Random Forest (RF) demonstrated the highest accuracy, both achieving approximately 97.58%.
 Naive Bayes (NB) achieved a perfect precision score, indicating zero false positives.
 Other models, including Gradient Boosting, Adaboost, Logistic Regression, and Bagging Classifier, displayed competitive performance with accuracy scores ranging from 94.68% to 96.03%.

The selection of the optimal model should consider factors beyond just accuracy, such as computational efficiency and the specific requirements of the application. It is advisable to perform further model fine-tuning and validation before making a final choice.