BUILDING A SMARTER AI-POWERED SPAM CLASSIFIER:

Problem Definition:

The problem is to build an AI-powered spam classifier that can accurately distinguish between spam and non-spam messages in emails or text messages. The goal is to reduce the number of false positives (classifying legitimate messages as spam) and false negatives (missing actual spam messages) while achieving a high level of accuracy.

Introduction:

The problem is to build an AI-powered spam classifier that can accurately distinguish between spam and non-spam messages in emails or text messages. The goal is to reduce the number of false positives (classifying legitimate messages as spam) and false negatives (missing actual spam messages) while achieving a high level of accuracy.

Steps:

• Data Collection:

Gather a labeled dataset of emails or text messages, with examples of both spam and non-spam (ham) messages.

- Data Preprocessing:
- Tokenization: Break text into words or tokens.
- Stopword Removal: Eliminate common words like "and," "the," etc.
- Lemmatization or Stemming: Reduce words to their base form.
- Feature Extraction:
- Convert text data into numerical features using techniques like TF-IDF (Term Frequency-Inverse Document Frequency) or word embeddings.

• Split Data:

Divide the dataset into training and testing sets.

Model Selection:

• Choose a machine learning algorithm like Naive Bayes, Support Vector Machines, or deep learning techniques (e.g., LSTM or CNN for text classification).

Model Training:

Train the chosen model on the training data.

Model Evaluation:

Evaluate the model's performance on the test data using metrics like accuracy, precision, recall, and F1-score.

Hyperparameter Tuning:

Optimize the model's hyperparameters to improve performance.

Cross-Validation (Optional): Use k-fold cross-validation to ensure the model's robustness.

Deployment: Implement the trained model in a real-world environment for spam classification.

- **Monitoring and Updating**: Continuously monitor the model's performance and update it as needed.
- **User Interface** (Optional): Create a user-friendly interface for users to interact with the spam classifier.

Processing Steps

Understanding the Data:

```
#importing libraries
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorize
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, confusion_matrix,
roc_curve, roc_auc_score
import nltk
from nltk.corpus import stopwords
from collections import Counter
```

#libraries for data visualization import matplotlib.pyplot as plt import seaborn as sns %matplotlib inline

Out [2]:

| | v1 | v2 | Unnamed: 2 | Unnamed: 3 | Unnamed: 4 |
|------|------|--|------------|------------|------------|
| 0 | ham | Go until jurong point, crazy Available only | NaN | NaN | NaN |
| 1 | ham | Ok lar Joking wif u oni | NaN | NaN | NaN |
| 2 | spam | Free entry in 2 a wkly comp to win FA Cup fina | NaN | NaN | NaN |
| 3 | ham | U dun say so early hor U c already then say | NaN | NaN | NaN |
| 4 | ham | Nah I don't think he goes to usf, he lives aro | NaN | NaN | NaN |
| ••• | ••• | ••• | ••• | ••• | ••• |
| 5567 | spam | This is the 2nd time we have tried 2 contact u | NaN | NaN | NaN |
| 5568 | ham | Will i_b going to esplanade fr home? | NaN | NaN | NaN |
| 5569 | ham | Pity, * was in mood for that. Soany other s | NaN | NaN | NaN |

| 5570 | ham | The guy did some bitching but I acted like i'd | NaN | NaN | NaN |
|------|-----|---|-----|-----|-----|
| 5571 | ham | Rofl. Its true to its name | NaN | NaN | NaN |

5572 rows × 5 columns

In [3]:

df.<u>info</u>()

```
2 v2
                  5572 non-null object
  3 Unnamed: 2 50 non-null
                                   object
      Unnamed: 3 12 non-null
                                   object
      Unnamed: 4 6 non-null
                                   object
 dtypes: object(5)
 memory usage: 217.8+ KB
In [4]:
 # Downloading the stopwords dataset
 nltk.download('stopwords')
 [nltk data] Error loading stopwords: <urlopen error [Errno -3]</pre>
 [nltk data]
                 Temporary failure in name resolution>
Out [4]:
False
In [5]:
 # Drop unnecessary columns from the DataFrame
 columns_to_drop = ["Unnamed: 2", "Unnamed: 3", "Unnamed: 4"]
 df.drop(columns=columns to drop, inplace=True)
In [6]:
 Df
Out [6]:
```

| v1 | v2 | |
|----|------|--|
| 0 | ham | Go until jurong point, crazy Available only |
| 1 | ham | Ok lar Joking wif u oni |
| 2 | spam | Free entry in 2 a wkly comp to win FA Cup fina |
| 3 | ham | U dun say so early hor U c already then say |

| 4 | ham | 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 | ham | Pity, * was in mood for that. Soany other s |
| 5570 | ham | The guy did some bitching but I acted like i'd |
| 5571 | ham | Rofl. Its true to its name |

5572 rows × 2 columns

```
In [7]:
```

Rename the columns "v1 and "v2" to new names

```
new_column_names = {"v1":"Category","v2":"Message"}
df.rename(columns = new_column_names,inplace = True)
In [8]:
    df[df.duplicated()]
Out [8]:
```

| Category | Message | |
|----------|---------|---|
| 102 | ham | As per your request 'Melle Melle (Oru Minnamin |
| 153 | ham | As per your request 'Melle Melle (Oru Minnamin |
| 206 | ham | As I entered my cabin my PA said, " Happy B'd |
| 222 | ham | Sorry, I'll call later |
| 325 | ham | No callsmessagesmissed calls |
| | | *** |

| 5524 | spam | You are awarded a SiPix Digital Camera! call 0 |
|------|------|---|
| 5535 | ham | I know you are thinkin malaria. But relax, chi |
| 5539 | ham | Just sleepingand surfing |
| 5553 | ham | Hahahause your brain dear |
| 5558 | ham | Sorry, I'll call later |

403 rows × 2 columns

In [9] :

#Drop duplicated values
df=df.drop duplicates()
df

Out [9]:

| Category | Message | |
|----------|---------|--|
| 0 | ham | Go until jurong point, crazy Available only |
| 1 | ham | Ok lar Joking wif u oni |
| 2 | spam | Free entry in 2 a wkly comp to win FA Cup fina |
| 3 | ham | U dun say so early hor U c already then say |
| 4 | ham | Nah I don't think he goes to usf, he lives aro |
| | | |
| 5570 | ham | The guy did some bitching but I acted like i'd |
| 5571 | ham | Rofl. Its true to its name |

5169 rows × 2 columns

```
In [10]:
```

df.info()

```
<class 'pandas.core.frame.DataFrame'>
Index: 5169 entries, 0 to 5571
Data columns (total 2 columns):
# Column Non-Null Count Dtype
--- 1 Category 5169 non-null object
2 Message 5169 non-null object
dtypes: object(2)
memory usage: 121.1+ KB
In [11]:
```

Out [11]:

df.describe()

| Category | Message | |
|----------|---------|--|
| count | 5169 | 5169 |
| unique | 2 | 5169 |
| top | ham | Go until jurong point, crazy Available only |
| freq | 4516 | 1 |

```
In [12]:
    Df.shape
Out [12]:
    (5169, 2)
In [13]:
    df['Category'].value counts()
Out[13]:
```

```
Category
 ham
          4516
           653
 spam
 Name: count, dtype: int64
Data Visualisation
In [14]:
 sns.countplot(data=df, x='Category')
 plt.xlabel('Category')
 plt.ylabel('count')
 plt.<u>title('Distribution of mails')</u>
 plt.show()
                       Distribution of mails
 4000
 3000
 2000
 1000
    0
                  ham
                                            spam
                             Category
Data Preprocessing
In [15]:
```

Assuming you have a DataFrame named 'df'

df.loc[df["Category"] == "spam", "Category"] = 0

```
df.loc[df["Category"] == "ham", "Category"] = 1
df.head()

/tmp/ipykernel_20/3584819934.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation:
https://pandas.pydata.org/pandas-
docs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
    df.loc[df["Category"] == "spam", "Category"] = 0
/tmp/ipykernel_20/3584819934.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation:
https://pandas.pydata.org/pandas-
docs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
    df.loc[df["Category"] == "ham", "Category"] = 1
```

Out [15]:

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

In [16]:

Separate the feature (X) and target (Y) data

```
X = df["Message"]
Y = df["Category"]
```

In [17]:

Out[17]:

```
1
        Go until jurong point, crazy.. Available only ...
2
                             Ok lar... Joking wif u oni...
3
        Free entry in 2 a wkly comp to win FA Cup fina...
4
        U dun say so early hor... U c already then say...
5
        Nah I don't think he goes to usf, he lives aro...
       This is the 2nd time we have tried 2 contact u...
5567
5568
                    Will I b going to esplanade fr home?
       Pity, * was in mood for that. So...any other s...
5569
5570
       The guy did some bitching but I acted like i'd...
5571
                                Rofl. Its true to its name
Name: Message, Length: 5169, dtype: object
In [18]:
Y
Out [18]:
0
        1
1
        1
2
        0
3
        1
        1
5567
        0
5568
        1
5569
        1
5570
        1
5571
        1
Name: Category, Length: 5169, dtype: object
In [19]:
```

Split the data into training and testing sets

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size =
```

```
0.2, random state = 42)
In [20]:
print (X.shape)
print(X train.shape)
print(X_test.shape)
(5169,)
(4135,)
(1034,)
Feature Extraction: TF-IDF
In [21]:
# Create a TF-IDF vectorizer to convert text messages into numerical
features
feature_extraction = TfidfVectorizer(min df=1, stop words="english",
lowercase=True)
In [22]:
# Convert the training and testing text messages into numerical features
using TF-IDF
X_train_features = feature extraction.fit transform(X train)
X test features = feature extraction.transform(X test)
In [23]:
# Convert the target values into 0 and 1
Y_train = Y_train.astype(int)
Y_test = Y_test.astype(int)
In [24]:
```

print(X train)

```
2228
                           Those were my exact intentions
5529
                                What about this one then.
2149
                       Waaaat?? Lololo ok next time then!
5058
        Free video camera phones with Half Price line ...
5051
        Tick, tick, tick .... Where are you ? I could ...
4740
       Many more happy returns of the day. I wish you...
       Nice line said by a broken heart- Plz don't cu...
474
3266
                        Ok then i come n pick u at engin?
       Eek that's a lot of time especially since Amer...
4016
879
      U have a Secret Admirer who is looking 2 make ...
Name: Message, Length: 4135, dtype: object
```

In [25]:

print(X train features)

```
(0, 3545) 0.7455593142248959
 (0, 2588)
                0.6664392762829205
 (2, 6589)
                0.3136674984299076
 (2, 4696)
                0.29654379102529516
 (2, 4002)
                0.6378379419700079
 (2, 6999)
                0.6378379419700079
 (3, 564) 0.2785767488573773
 (3, 1534)
                0.23384958966251285
 (3, 52) 0.26549489341098675
 (3, 4344)
                0.22076773421612225
 (3, 6770)
                0.2300494583671639
 (3, 251) 0.19582167067522926
 (3, 4299)
                0.18532229917229942
 (3, 4630)
                0.26549489341098675
 (3, 1997)
                0.26549489341098675
 (3, 516) 0.19460402332334106
 (3, 4419)
                0.2562131692599451
 (3, 271) 0.23384958966251285
 (3, 5450)
                0.2300494583671639
 (3, 3941)
                0.18912243046764834
 (3, 5171)
                0.20953002785296104
```

```
(3, 3168) 0.19120469004402674
(3, 4954)
            0.23384958966251285
(3, 1553)
             0.20428654549041733
(3, 6938)
            0.19708708091575408
: :
(4132, 1825) 0.3605065932469792
(4132, 4696) 0.3418197199207224
(4133, 5893) 0.376872105216547
(4133, 4973) 0.36369662422743665
(4133, 2451) 0.376872105216547
(4133, 901) 0.36369662422743665
(4133, 4029) 0.27296922168195425
(4133, 4303) 0.2797666732547047
(4133, 3930) 0.19090886726821316
(4133, 2550) 0.326557029270423
(4133, 4007) 0.2670514851432264
(4133, 6192) 0.21536918062740018
(4133, 6589) 0.19446518344396782
(4134, 45)
             0.36133141627364085
(4134, 6198)
             0.34436343393010593
(4134, 216)
            0.34436343393010593
(4134, 6543) 0.29397934692144273
(4134, 5512)
             0.31535647652238075
(4134, 799)
             0.31535647652238075
(4134, 5715) 0.3033175014581906
(4134, 6069)
             0.2508916342134232
(4134, 4013) 0.26098383065689107
(4134, 1895) 0.2301166472830892
(4134, 4139) 0.20748487401135496
(4134, 6867)
             0.16697204675649222
```

Model Training

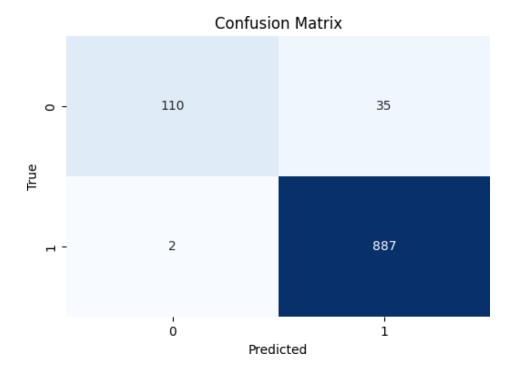
In [26]:

Create a logistic regression model and train it on the training data

```
model = LogisticRegression() #model.fit(X_train_features, Y_train)
```

```
Out[26]:
    LogisticRegression
LogisticRegression()
Model Evaluation and Prediction
In [27]:
# Make predictions on the training data and calculate the accuracy
prediction_on_training_data = model.predict(X train features)
accuracy on training data = accuracy score(Y train,
prediction on training data)
In [28]:
print("Accuracy on training data:", accuracy on training data)
Accuracy on training data: 0.9613059250302297
In [29]:
# Make predictions on the test data and calculate the accuracy
prediction_on_test_data = model.predict(X test features)
accuracy on test data = accuracy_score(Y_test,prediction_on_test_data)
In [30]:
print("Accuracy on test data:",accuracy_on_test_data)
Accuracy on test data: 0.9642166344294004
In [31]:
# Test the model with some custom email messages
```

```
input_mail = ["Congratulations! You've won a free vacation to an exotic
island. Just click on the link below to claim your prize."]
input data features = feature extraction.transform(input mail)
prediction = model.predict(input data features)
if (prediction)[0] == 1:
    print("Ham Mail")
else:
    print("Spam Mail")
Spam Mail
In [32]:
input mail = ["This is a friendly reminder about our meeting scheduled
for tomorrow at 10:00 AM in the conference room. Please make sure to
prepare your presentation and bring any necessary materials."]
input data features = feature extraction.transform(input mail)
prediction = model.predict(input data features)
if (prediction)[0] == 1:
    print("Ham Mail")
else:
    print("Spam Mail")
Ham Mail
In [33]:
# Data visualization - Confusion Matrix
cm = confusion matrix(Y test, prediction on test data)
plt.figure(figsize=(6, 4))
sns.heatmap(cm, annot=True, fmt="d", cmap='Blues', cbar=False)
plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('Confusion Matrix')
plt.show()
```

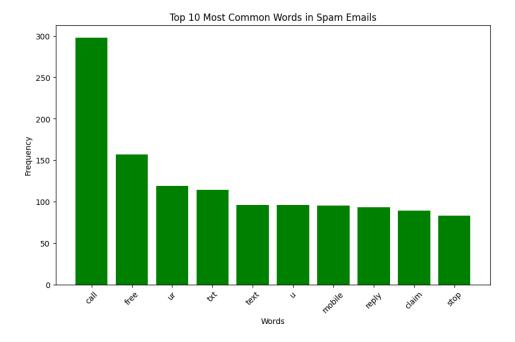


In [34]:
Data visualization - Top 10 Most Common Words in Spam Emails

```
stop_words = set(stopwords.words('english'))
spam_words = " ".join(df[df['Category'] == 0]['Message']).split()
ham_words = " ".join(df[df['Category'] == 1]['Message']).split()

spam_word_freq = Counter([word.lower() for word in spam_words if
word.lower() not in stop_words and word.isalpha()])

plt.figure(figsize=(10, 6))
plt.bar(*zip(*spam_word_freq.most_common(10)), color='g')
plt.xlabel('Words')
plt.ylabel('Frequency')
plt.title('Top 10 Most Common Words in Spam Emails')
plt.xticks(rotation=45)
plt.show()
```



In [35]:
Data visualization - Top 10 Most Common Words in Ham Emails

```
ham_word_freq = Counter([word.lower() for word in ham words if
word.lower() not in stop words and word.isalpha()])
plt.figure(figsize=(10, 6))
plt.bar(*zip(*ham word freq.most common(10)), color='maroon')
plt.xlabel('Words')
plt.ylabel('Frequency')
plt.title('Top 10 Most Common Words in Ham Emails')
plt.xticks(rotation=45)
plt.show()
                 Top 10 Most Common Words in Ham Emails
800
700
600
500
400
300
200
100
 0
                                come
                                     KUOM
                 8
                      go<sup>x</sup>
                           ive
```

Conclusion

An AI-powered smart spam classifier offers significant benefits by efficiently filtering out unwanted messages, emails, or content. It enhances user experience by reducing exposure to spam, phishing attempts, and irrelevant information. In conclusion, AI-driven spam classifiers contribute to a safer and more organized digital environment, ultimately saving time and ensuring a more enjoyable online experience for users.