

# Dataset

<https://www.kaggle.com/datasets/uciml/sms-spam-collection-dataset>

```
In [1]: import numpy as np
import pandas as pd
import re
import nltk
import seaborn as sns
from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
from nltk.stem import WordNetLemmatizer
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.naive_bayes import MultinomialNB
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report as cr
```

```
In [2]: import nltk
nltk.download('wordnet')

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

```
Out[2]: True
```

```
In [3]: # Reading the dataset
spam_text = pd.read_csv('SPAM text message Dataset.csv')
```

```
In [4]: spam_text
```

```
Out[4]:
```

	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...
...	...	...
5567	spam	This is the 2nd time we have tried 2 contact u...
5568	ham	Will ü b going to esplanade fr home?
5569	ham	Pity, * was in mood for that. So...any 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 [5]: # Looking for one message example
spam_text['Message'].loc[4995]
```

```
Out[5]: 'My drive can only be read. I need to write'
```

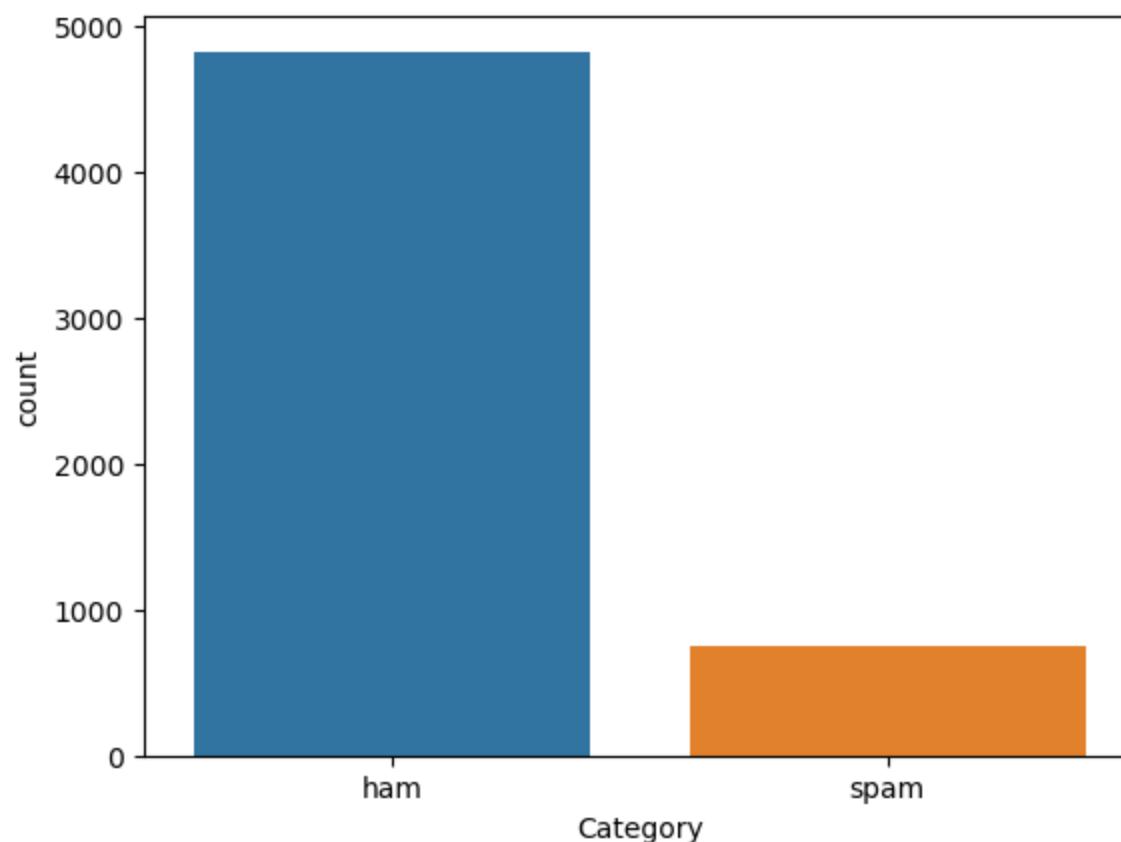
```
In [6]: # Downloading Stop Words  
nltk.download('stopwords')
```

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

```
Out[6]: True
```

```
In [7]: sns.countplot(x='Category', data=spam_text)
```

```
Out[7]: <Axes: xlabel='Category', ylabel='count'>
```



## Implementing Stemming and removing stop words in the messages

```
In [8]: # Calling portstemmer for the purpose of stemming the words in messages  
ps = PorterStemmer()
```

```
In [9]: corpus = []  
for i in range(0, len(spam_text)):  
    messages = re.sub('[^a-zA-Z]', ' ', spam_text['Message'][i])  
    messages = messages.lower()  
    messages = messages.split()  
    messages = [ps.stem(word) for word in messages if not word in stopwords.words('english')]  
    messages = ' '.join(messages)  
    corpus.append(messages)
```

# Applying BagOfWords (BOW) Method

```
In [10]: # Creating a bag of words model  
cv = CountVectorizer(max_features=2500)  
x = cv.fit_transform(corpus).toarray()  
  
In [11]: # Converting the categorical values into dummy variables  
y = pd.get_dummies(spam_text['Category'])  
y = y.iloc[:,1].values  
  
In [12]: # splitting the data for traing and testing  
X_train, X_test, y_train, y_test = train_test_split(x,y,test_size=0.25,random_state=0)
```

## Navie Bayes

```
In [13]: #Creating navie bayes model  
text_analysis_model = MultinomialNB().fit(X_train,y_train)  
  
In [14]: # prediction  
y_pred = text_analysis_model.predict(X_test)  
  
In [15]: # getting accuracy score for test data and predicted data  
score = accuracy_score(y_test,y_pred)  
print(score)  
print(cr(y_test,y_pred))  
  
0.9856424982053122  


|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0            | 0.99      | 0.99   | 0.99     | 1208    |
| 1            | 0.95      | 0.94   | 0.95     | 185     |
| accuracy     |           |        | 0.99     | 1393    |
| macro avg    | 0.97      | 0.97   | 0.97     | 1393    |
| weighted avg | 0.99      | 0.99   | 0.99     | 1393    |


```

## Implementation of Lemmatization and removing stop words in the messages

```
In [16]: lemmatizer=WordNetLemmatizer()  
  
In [17]: corpus = []  
for i in range(0, len(spam_text)):  
    messages = re.sub('[^a-zA-Z]', ' ', spam_text['Message'][i])  
    messages = messages.lower()  
    messages = messages.split()  
  
    messages = [lemmatizer.lemmatize(word) for word in messages if not word in stopwords]  
    messages = ' '.join(messages)  
    corpus.append(messages)
```

```
In [18]: # Creating Bag of Words model
cv = CountVectorizer(max_features=2500)
x = cv.fit_transform(corpus).toarray()

In [19]: # Converting the categorical values into dummy variables
y = pd.get_dummies(spam_text['Category'])
y = y.iloc[:,1].values

In [20]: # splitting the data for traing and testing
X_train, X_test, y_train, y_test = train_test_split(x,y,test_size=0.25,random_state=0)
```

## Naive Bayes

```
In [21]: text_analysis_model2 = MultinomialNB().fit(X_train,y_train)

In [22]: y_prediction = text_analysis_model2.predict(X_test)

In [23]: # getting accuracy score for test data and predicted data
score = accuracy_score(y_test,y_prediction)
print(score)
print(cr(y_test,y_prediction))

0.9834888729361091
      precision    recall   f1-score   support
          0         0.99     0.99     0.99     1208
          1         0.95     0.92     0.94      185
          accuracy           0.98     1393
          macro avg       0.97     0.96     0.96     1393
          weighted avg     0.98     0.98     0.98     1393
```

## Implementing Tf-Idf for Lemmatization

```
In [24]: # Creating a TFIDF model
from sklearn.feature_extraction.text import TfidfVectorizer
tv = TfidfVectorizer(max_features=2500)
X = tv.fit_transform(corpus).toarray()

In [25]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.20, random_state=0)
```

## Naive bayes for Tf-Idf

```
In [26]: text_analysis_model2 = MultinomialNB().fit(X_train,y_train)

In [27]: y_predtf = text_analysis_model2.predict(X_test)

In [28]: score=accuracy_score(y_test,y_predtf)
print(score)
print(cr(y_test,y_predtf))
```

0.97847533632287

	precision	recall	f1-score	support
0	0.98	1.00	0.99	955
1	0.99	0.86	0.92	160
accuracy			0.98	1115
macro avg	0.98	0.93	0.95	1115
weighted avg	0.98	0.98	0.98	1115

## Results of Count Vectorizer (CV)

Accuracy

Stemming Naive-Bayes - 98.56% ~ (99%)

Lemmatization Naive-Bayes - 98.34% ~ (98%)

## Results of TF-IDF

Lemmatization Naive-Bayes - 97.84% ~ (98%)

In [ ]: