

Overview

Context

The SMS Spam Collection is a set of SMS tagged messages that have been collected for SMS Spam research. It contains one set of SMS messages in English of 5,574 messages, tagged according to being ham (legitimate) or spam.

Content

The files contain one message per line. Each line is composed by two columns: v1 contains the label (ham or spam) and v2 contains the raw text.

```
import pandas as pd
import numpy as np
import nltk
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split # Import from model_selection instead of cross_validation
from sklearn.naive_bayes import GaussianNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report

from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
import re

import chardet
import requests
import pandas as pd
import io

# Fetch the content from the raw CSV URL
url = "https://raw.githubusercontent.com/PragadishTRS/SMS_Spam_Collection_Dataset/main/SMS%20Spam%20Collection%20Dataset.csv" # Use raw
response = requests.get(url)

# Detect the encoding
result = chardet.detect(response.content)

# Decode the content using the detected encoding
text = response.content.decode(result['encoding'])

# Read the CSV data into a pandas DataFrame
df = pd.read_csv(io.StringIO(text)) # Use StringIO to treat the decoded text as a file-like object

df = df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], axis=1)
df.head()
```

| | 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... |

Next steps:

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```
# Replace ham with 0 and spam with 1
df = df.replace(['ham', 'spam'], [0, 1])
```

```
df.head()
```

| | v1 | v2 |
|---|----|---|
| 0 | 0 | Go until jurong point, crazy.. Available only ... |
| 1 | 0 | Ok lar... Joking wif u oni... |
| 2 | 1 | Free entry in 2 a wkly comp to win FA Cup fina... |
| 3 | 0 | U dun say so early hor... U c already then say... |
| 4 | 0 | Nah I don't think he goes to usf. he lives aro... |

Next steps:

[Generate code with df](#)[View recommended plots](#)[New interactive sheet](#)

```
# Count the number of words in each Text
df['Count']=0
for i in np.arange(0,len(df.v2)):
    df.loc[i,'Count'] = len(df.loc[i,'v2'])
df.head()
```

| | v1 | v2 | Count |
|---|----|---|-------|
| 0 | 0 | Go until jurong point, crazy.. Available only ... | 111 |
| 1 | 0 | Ok lar... Joking wif u oni... | 29 |
| 2 | 1 | Free entry in 2 a wkly comp to win FA Cup fina... | 155 |
| 3 | 0 | U dun say so early hor... U c already then say... | 49 |
| 4 | 0 | Nah I don't think he goes to usf. he lives aro... | 61 |

Next steps:

[Generate code with df](#)[View recommended plots](#)[New interactive sheet](#)

```
# Total ham(0) and spam(1) messages
df['v1'].value_counts()
```

| | count |
|----|-------|
| v1 | |
| 0 | 4825 |
| 1 | 747 |

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5572 entries, 0 to 5571
Data columns (total 3 columns):
#   Column  Non-Null Count  Dtype
---  ---
0    v1      5572 non-null    int64
1    v2      5572 non-null    object
2   Count  5572 non-null    int64
dtypes: int64(2), object(1)
memory usage: 130.7+ KB
```

```
corpus = []
ps = PorterStemmer()
# Original Messages
```

```
print (df['v2'][0])
print (df['v2'][1])
```

```
Go until jurong point, crazy.. Available only in bugis n great world la e buffet... Cine there got amore wat...
Ok lar... Joking wif u oni...
```

```
import nltk
nltk.download('stopwords')
```

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Unzipping corpora/stopwords.zip.
True
```

```
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
```

```
# Download stopwords if not already downloaded
nltk.download('stopwords')

corpus = []
ps = PorterStemmer()

for i in range(0, 5572):

    # Applying Regular Expression

    '''
    Replace email addresses with 'emailaddr'
    Replace URLs with 'httpaddr'
    Replace money symbols with 'moneysymb'
    Replace phone numbers with 'phonenumbr'
    Replace numbers with 'numbr'
    '''

    msg = df['v2'][i]
    msg = re.sub('\b[\w\.-]+?@\w+\.\w{2,4}\b', 'emailaddr', df['v2'][i])
    msg = re.sub('(http[s]?|https?|https?)|(\w+.[A-Za-z]{2,4}S*)', 'httpaddr', df['v2'][i])
    msg = re.sub('£|\$', 'moneysymb', df['v2'][i])
    msg = re.sub('\b(\+|\d{1,2})\s)?\d?[\-\.]\d{3}\)?\s.-]? \d{3}[\s.-]? \d{4}\b', 'phonenumbr', df['v2'][i])
    msg = re.sub('\d+(\.\d+)?', 'numbr', df['v2'][i])

    ''' Remove all punctuations '''
    msg = re.sub('[^\w\d\s]', ' ', df['v2'][i])

    if i<2:
        print("\t\t\t\t MESSAGE ", i)

    if i<2:
        print("\n After Regular Expression - Message ", i, " : ", msg)


    # Each word to lower case
    msg = msg.lower()
    if i<2:
        print("\n Lower case Message ", i, " : ", msg)

    # Splitting words to Tokenize
    msg = msg.split()
    if i<2:
        print("\n After Splitting - Message ", i, " : ", msg)

    # Stemming with PorterStemmer handling Stop Words
    msg = [ps.stem(word) for word in msg if not word in set(stopwords.words('english'))]
    if i<2:
        print("\n After Stemming - Message ", i, " : ", msg)

    # preparing Messages with Remaining Tokens
    msg = ' '.join(msg)
    if i<2:
        print("\n Final Prepared - Message ", i, " : ", msg, "\n\n")

    # Preparing WordVector Corpus
    corpus.append(msg)
```

 [nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
MESSAGE 0

After Regular Expression - Message 0 : Go until jurong point crazy Available only in bugis n great world la e buffet Cine
Lower case Message 0 : go until jurong point crazy available only in bugis n great world la e buffet cine there got amore
After Splitting - Message 0 : ['go', 'until', 'jurong', 'point', 'crazy', 'available', 'only', 'in', 'bugis', 'n', 'great', 'wor
After Stemming - Message 0 : ['go', 'jurong', 'point', 'crazi', 'avail', 'bugi', 'n', 'great', 'world', 'la', 'e', 'buffet', 'ci
Final Prepared - Message 0 : go jurong point crazi avail bugi n great world la e buffet cine got amor wat

MESSAGE 1

After Regular Expression - Message 1 : Ok lar Joking wif u oni
Lower case Message 1 : ok lar joking wif u oni
After Splitting - Message 1 : ['ok', 'lar', 'joking', 'wif', 'u', 'oni']
After Stemming - Message 1 : ['ok', 'lar', 'joke', 'wif', 'u', 'oni']
Final Prepared - Message 1 : ok lar joke wif u oni

```
cv = CountVectorizer()
x = cv.fit_transform(corpus).toarray()
```

```
y = df['v1']
print (y.value_counts())
```

```
print(y[0])
print(y[1])
```

```
↗ v1
0    4825
1     747
Name: count, dtype: int64
0
0
```

```
le = LabelEncoder()
y = le.fit_transform(y)
```

```
print(y[0])
print(y[1])
```

```
↗ 0
0
```

```
# Splitting to Training and Testing DATA
xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size= 0.20, random_state = 0)
```

```
bayes_classifier = GaussianNB()
bayes_classifier.fit(xtrain, ytrain)
```

```
↗ GaussianNB
GaussianNB()
```

```
# Predicting
y_pred = bayes_classifier.predict(xtest)
```

```
# Evaluating
cm = confusion_matrix(ytest, y_pred)
cm
```

```
↗ array([[824, 125],
        [ 19, 147]])
```

```
print ("Accuracy : %0.5f \n\n" % accuracy_score(ytest, bayes_classifier.predict(xtest)))
print (classification_report(ytest, bayes_classifier.predict(xtest)))
```

```
↗ Accuracy : 0.87085
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.98 | 0.87 | 0.92 | 949 |
| 1 | 0.54 | 0.89 | 0.67 | 166 |
| accuracy | | | 0.87 | 1115 |
| macro avg | 0.76 | 0.88 | 0.80 | 1115 |
| weighted avg | 0.91 | 0.87 | 0.88 | 1115 |

```
# Applying Decision Tree
dt = DecisionTreeClassifier(random_state=50)
dt.fit(xtrain, ytrain)
```

```
↗ DecisionTreeClassifier
DecisionTreeClassifier(random_state=50)
```

```
# Predicting
y_pred_dt = dt.predict(xtest)
```

```
# Evaluating
cm = confusion_matrix(ytest, y_pred_dt)
```

```
print(cm)
```

```
[[944  5]
 [ 27 139]]
```

```
print ("Accuracy : %0.5f \n\n" % accuracy_score(ytest, dt.predict(xtest)))
print (classification_report(ytest, dt.predict(xtest)))
```

```
Accuracy : 0.97130
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.97 | 0.99 | 0.98 | 949 |
| 1 | 0.97 | 0.84 | 0.90 | 166 |
| accuracy | | | 0.97 | 1115 |
| macro avg | 0.97 | 0.92 | 0.94 | 1115 |
| weighted avg | 0.97 | 0.97 | 0.97 | 1115 |

✓ Final Accuracy

1) Decision Tree : 96.861%

2) Guassian NB : 87.085%

Thanks for having a look!!!