EMAIL SPAM DETECTION WITH MACHINE LEARNING

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1 EMAIL SPAM DETECTION WITH MACHINE LEARNING

Dataset Information

The "spam" concept is diverse: advertisements for products/web sites, make money fast schemes, chain letters, pornography...

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 5,572 messages, tagged according being ham (legitimate) or spam.

Attributes

Email Messages Label (spam/ham)

```
[1]: #Import modules
     import pandas as pd
     import numpy as np
     import nltk
     import re
     from nltk.corpus import stopwords
[2]: df = pd.read_csv("C:/Users/MyPc/Downloads/task4/spam.csv",__
      ⇔encoding=('ISO-8859-1'), low_memory =False)
[3]: df.head()
[3]:
          v1
                                                                 v2 Unnamed: 2 \
     0
         ham
              Go until jurong point, crazy.. Available only ...
                                                                         NaN
     1
                                    Ok lar... Joking wif u oni...
                                                                       NaN
         ham
     2
        spam
             Free entry in 2 a wkly comp to win FA Cup fina...
                                                                         NaN
         ham U dun say so early hor... U c already then say...
     3
                                                                       NaN
              Nah I don't think he goes to usf, he lives aro...
                                                                         NaN
       Unnamed: 3 Unnamed: 4
     0
              {\tt NaN}
                          NaN
     1
              NaN
                          NaN
     2
              NaN
                          NaN
     3
              NaN
                          NaN
     4
              NaN
                          NaN
```

```
[4]: # get necessary columns for processing
      df = df[['v2', 'v1']]
      # df.rename(columns={'v2': 'messages', 'v1': 'label'}, inplace=True)
      df = df.rename(columns={'v2': 'messages', 'v1': 'label'})
      df.head()
 [4]:
                                                   messages label
      O Go until jurong point, crazy.. Available only ...
                             Ok lar... Joking wif u oni...
      2 Free entry in 2 a wkly comp to win FA Cup fina...
      3 U dun say so early hor... U c already then say...
      4 Nah I don't think he goes to usf, he lives aro...
                                                            ham
     2 Preprocessing the dataset
 [5]: # check for null values
      df.isnull().sum()
 [5]: messages
      label
      dtype: int64
 [6]: df.shape
 [6]: (5572, 2)
 [7]: df.columns
 [7]: Index(['messages', 'label'], dtype='object')
 [8]: df.drop_duplicates(inplace=True)
      print(df.shape)
     (5169, 2)
 [9]: # download the stopwords package
      nltk.download("stopwords")
     [nltk_data] Downloading package stopwords to
                     C:\Users\MyPc\AppData\Roaming\nltk_data...
     [nltk_data]
                   Package stopwords is already up-to-date!
     [nltk_data]
 [9]: True
[10]: STOPWORDS = set(stopwords.words('english'))
      def clean_text(text):
          # convert to lowercase
```

```
text = text.lower()
# remove special characters
text = re.sub(r'[^0-9a-zA-Z]', ' ', text)
# remove extra spaces
text = re.sub(r'\s+', ' ', text)
# remove stopwords
text = " ".join(word for word in text.split() if word not in STOPWORDS)
return text
```

```
[11]: # clean the messages
df['clean_text'] = df['messages'].apply(clean_text)
df.head()
```

```
[11]:
                                                   messages label \
      O Go until jurong point, crazy.. Available only ...
                             Ok lar... Joking wif u oni... ham
      1
      2 Free entry in 2 a wkly comp to win FA Cup fina...
      3 U dun say so early hor... U c already then say...
      4 Nah I don't think he goes to usf, he lives aro...
                                                 clean text
     O go jurong point crazy available bugis n great ...
                                   ok lar joking wif u oni
      2 free entry 2 wkly comp win fa cup final tkts 2...
      3
                       u dun say early hor u c already say
                    nah think goes usf lives around though
      4
```

3 Input Split

```
[12]: X = df['clean_text']
y = df['label']
```

4 Model Training

```
[13]: from sklearn.pipeline import Pipeline
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.metrics import classification_report
from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer,

TfidfTransformer

def classify(model, X, y):
    # train test split
    x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.25, u)
    random_state=42, shuffle=True, stratify=y)
    # model training
```

```
pipeline_model = Pipeline([('vect', CountVectorizer()),
                                    ('tfidf', TfidfTransformer()),
                                    ('clf', model)])
          pipeline_model.fit(x_train, y_train)
          print('Accuracy:', pipeline_model.score(x_test, y_test)*100)
            cv_score = cross_val_score(model, X, y, cv=5)
            print("CV Score:", np.mean(cv_score)*100)
          y_pred = pipeline_model.predict(x_test)
          print(classification_report(y_test, y_pred))
[14]: from sklearn.linear_model import LogisticRegression
      model = LogisticRegression()
      classify(model, X, y)
     Accuracy: 95.7463263727765
                   precision
                                recall f1-score
                                                    support
                        0.95
                                   1.00
                                             0.98
                                                       1130
              ham
             spam
                        0.99
                                  0.67
                                             0.80
                                                        163
                                             0.96
                                                       1293
         accuracy
        macro avg
                                             0.89
                                                       1293
                        0.97
                                  0.83
     weighted avg
                        0.96
                                  0.96
                                             0.95
                                                       1293
```

[15]: from sklearn.naive_bayes import MultinomialNB
model = MultinomialNB()
classify(model, X, y)

Accuracy: 96.13302397525135

	precision	recall	f1-score	support
ham	0.96	1.00	0.98	1130
spam	1.00	0.69	0.82	163
			0.00	1000
accuracy			0.96	1293
macro avg	0.98	0.85	0.90	1293
weighted avg	0.96	0.96	0.96	1293

```
[16]: from sklearn.svm import SVC
model = SVC(C=3)
classify(model, X, y)
```

Accuracy: 97.6798143851508

precision recall f1-score support

ham	0.97	1.00	0.99	1130
spam	0.99	0.82	0.90	163
accuracy			0.98	1293
macro avg	0.98	0.91	0.94	1293
weighted avg	0.98	0.98	0.98	1293

[17]: from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier()
classify(model, X, y)

Accuracy: 97.7571539056458

·	precision	recall	f1-score	support
ham	0.98	1.00	0.99	1130
spam	0.99	0.83	0.90	163
accuracy			0.98	1293
macro avg	0.98	0.91	0.95	1293
weighted avg	0.98	0.98	0.98	1293