# spam-email-checking

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### #Machine Learning Project on Spam Email Checking

##Importing all necessary librares

```
import pandas as pd
import numpy as np
import nltk
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import MultinomialNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import confusion_matrix, accuracy_score,
classification_report
import seaborn as sns
import matplotlib.pyplot as plt
```

#### • Download stopwords

```
[14]: nltk.download('stopwords')
      nltk.download('punkt')
     [nltk_data] Downloading package stopwords to /root/nltk_data...
                   Package stopwords is already up-to-date!
     [nltk_data]
     [nltk_data] Downloading package punkt to /root/nltk_data...
                   Unzipping tokenizers/punkt.zip.
     [nltk_data]
[14]: True
     ###Uploading dataset
[19]: def clean_data(file_path):
          with open(file_path, 'r', encoding='latin-1') as file:
              lines = file.readlines()
          data = []
          for line in lines[1:]: # Skip header
              parts = line.split('\t')
```

```
if len(parts) == 2:
            data.append(parts)
        else:
            # Handle cases where there are more than 2 parts due to extra tabs
            type_part = parts[0]
            message_part = ' '.join(parts[1:])
            data.append([type_part, message_part.strip()])
    # Convert to DataFrame
    df = pd.DataFrame(data, columns=['type', 'message'])
    return df
# 1. Collect Data Set
df = clean_data('spam.csv')
# Verify and drop any rows with missing values
df.dropna(subset=['type', 'message'], inplace=True)
df
```

```
[19]:
           type
                                                              message
            ham Go until jurong point, crazy.. Available only ...
      1
            ham
                                    Ok lar... Joking wif u oni...\n
      2
           spam Free entry in 2 a wkly comp to win FA Cup fina...
                 U dun say so early hor... U c already then say...
      3
            ham
      4
            ham
                 Nah I don't think he goes to usf, he lives aro...
      651
            ham
                                Anything lor. Juz both of us lor.\n
      652
            ham Get me out of this dump heap. My mom decided t...
                 Ok lor... Sony ericsson salesman... I ask shuh...
      653
            ham
      654
                                               Ard 6 like dat lor.\n
            ham
      655
            ham Why don't you wait 'til at least wednesday to ...
      [656 rows x 2 columns]
```

#### ####Pre-processing

- Convert labels to binary values: spam=1, ham=0
- Text cleaning and preprocessing

```
[22]: df['label'] = df['type'].map({'spam': 1, 'ham': 0})
      # Text cleaning and preprocessing
      def preprocess_text(text):
          text = text.lower()
          text = nltk.word_tokenize(text)
          text = [word for word in text if word.isalnum()]
          text = [word for word in text if word not in stopwords.words('english')]
          stemmer = PorterStemmer()
```

```
text = [stemmer.stem(word) for word in text]
          return ' '.join(text)
      df['message'] = df['message'].apply(preprocess_text)
      df['message']
[22]: 0
             go jurong point crazi avail bugi n great world...
                                          ok lar joke wif u oni
      1
      2
             free entri 2 wkli comp win fa cup final tkt 21...
      3
                           u dun say earli hor u c alreadi say
      4
                          nah think goe usf live around though
      651
                                           anyth lor juz us lor
                         get dump heap mom decid come low bore
      652
      653
             ok lor soni ericsson salesman ask shuhui say g...
      654
                                             ard 6 like dat lor
      655
                                   wait least wednesday see get
      Name: message, Length: 656, dtype: object
     ####Feature Selection
[23]: vectorizer = CountVectorizer()
      X = vectorizer.fit_transform(df['message']).toarray()
      y = df['label']
     ####Apply Spam Filter Algorithms * Train & Test Data * Naive Bayes Classifier * Decision
     Tree Classifier (J48 equivalent in sklearn)
[24]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,__
       →random_state=42)
[25]: # Naive Bayes Classifier
      nb_model = MultinomialNB()
      nb model.fit(X train, y train)
      y_pred_nb = nb_model.predict(X_test)
      # Decision Tree Classifier (J48 equivalent in sklearn)
      dt_model = DecisionTreeClassifier()
      dt_model.fit(X_train, y_train)
      y_pred_dt = dt_model.predict(X_test)
     #####Confusion Matrix
       1. Evaluate Naive Bayes
       2. Evaluate Decision Tree
[30]: def evaluate_model(y_test, y_pred, model_name):
          cm = confusion_matrix(y_test, y_pred)
          accuracy = accuracy_score(y_test, y_pred)
```

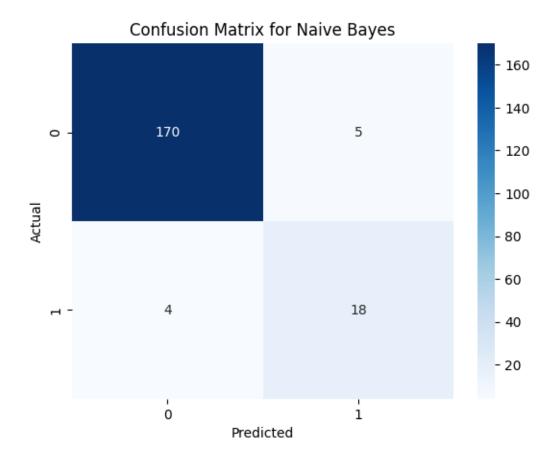
```
print(f"{model_name} Accuracy: {accuracy:.2f}")
print(classification_report(y_test, y_pred))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
plt.title(f'Confusion Matrix for {model_name}')
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.show()
```

```
[28]: # Evaluate Naive Bayes
evaluate_model(y_test, y_pred_nb, 'Naive Bayes')

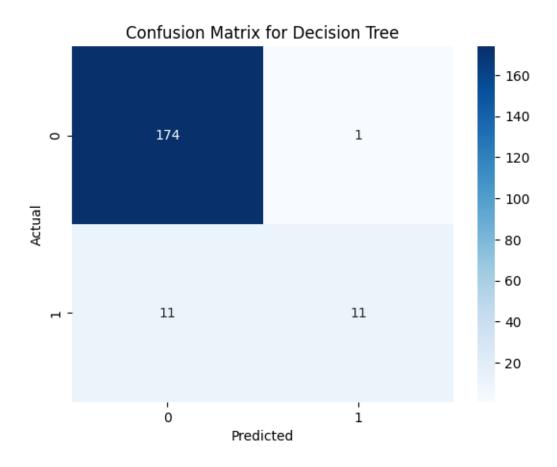
# Evaluate Decision Tree
evaluate_model(y_test, y_pred_dt, 'Decision Tree')
```

Naive Bayes Accuracy: 0.95

	precision	recall	f1-score	support
0	0.98	0.97	0.97	175
1	0.78	0.82	0.80	22
accuracy			0.95	197
macro avg	0.88	0.89	0.89	197
weighted avg	0.96	0.95	0.95	197



Decision Tre	e Accuracy:	0.94		
	precision	recall	f1-score	support
0	0.94	0.99	0.97	175
1	0.92	0.50	0.65	22
accuracy			0.94	197
macro avg	0.93	0.75	0.81	197
weighted avg	0.94	0.94	0.93	197



## #####Comparing Accuracy

```
[34]: print(f"Naive Bayes Accuracy: {accuracy_score(y_test, y_pred_nb):.2f}")
print(f"Decision Tree Accuracy: {accuracy_score(y_test, y_pred_dt):.2f}")
```

Naive Bayes Accuracy: 0.95 Decision Tree Accuracy: 0.94