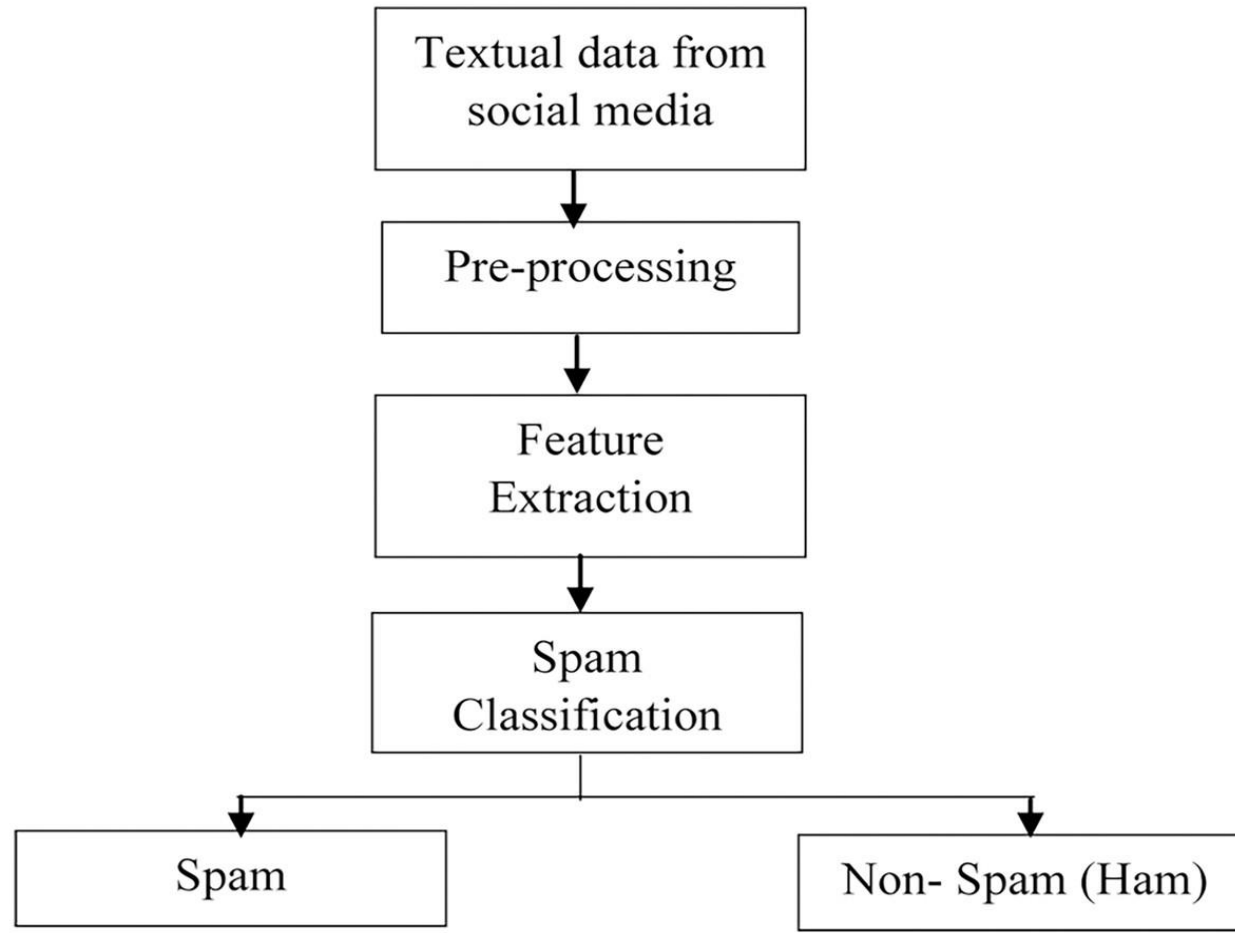
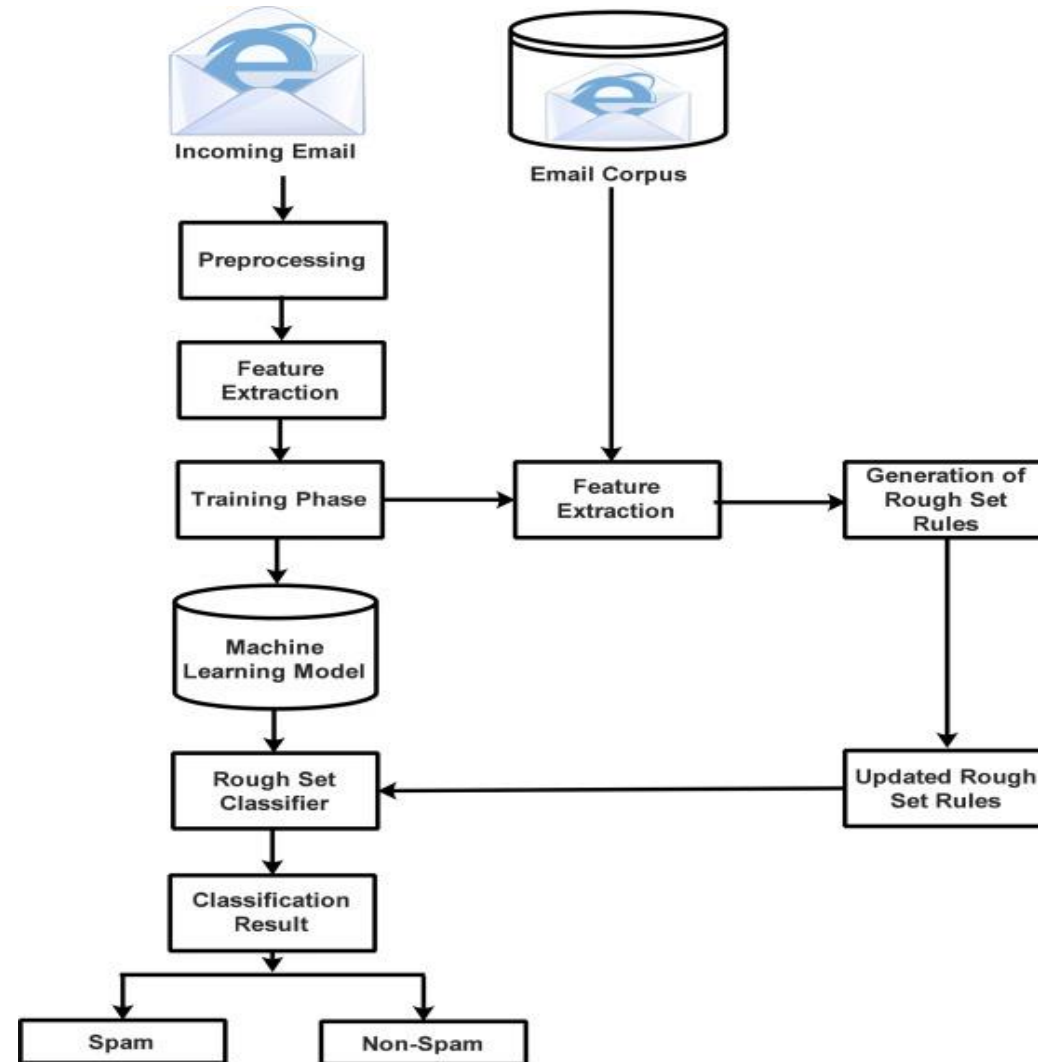


Real Time Text Classification for Spam Detection

Flow Chart



Architecture Diagram



Coding

```
In [2]: # import modules

# load the data
import pandas as pd

# splitting the data into training and testing
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import CountVectorizer

# build the naive bayes multinomial model
from sklearn.naive_bayes import MultinomialNB

# to see the accuracy
from sklearn.metrics import accuracy_score, classification_report
```

```
In [3]: # import the dataset
data = pd.read_table('SMSSpamCollection', sep='\t', names=['label', 'message'])
```

```
In [4]: data['label'] = data.label.map({'ham': 0, 'spam': 1})
```

```
In [5]: # split the data into training & testing
X_train, X_test, Y_train, Y_test = train_test_split(data['message'], data['label'])
print('total number of data in the set: {}'.format(data.shape))
print('total number of data in the training: {}'.format(X_train.shape))
print('total number of data in the testing: {}'.format(X_test.shape))

total number of data in the set: (5572, 2)
total number of data in the training: (4179,)
total number of data in the testing: (1393,)
```

```
In [6]: count_vector = CountVectorizer()
training_data = count_vector.fit_transform(X_train)
testing_data = count_vector.transform(X_test)
```

```
In [7]: # build the NB multinomial model
NB_M = MultinomialNB()
model = NB_M.fit(training_data, Y_train)
Prediction = model.predict(testing_data)
```

```
In [7]: ► # build the NB multinomial model
        NB_M = MultinomialNB()
        model = NB_M.fit(training_data, Y_train)
        Prediction = model.predict(testing_data)
```

```
In [8]: ► # predict the score & report
        Accuracy = accuracy_score(Y_test, Prediction)
        Report = classification_report(Y_test, Prediction)
        print('Accuracy score: ', Accuracy)
        print('Classification report: ', Report)
```

Output

Accuracy score: 0.9770279971284996

Classification report:

			precision	recall	f1-score	support
	0	0.98	0.99	0.99		1202
	1	0.95	0.88	0.91		191
	accuracy		0.98			1393
	macro avg	0.97	0.94	0.95		1393
	weighted avg	0.98	0.98	0.98		1393