Process_data_From_Blockchain

June 1, 2020

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In [1]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.impute import SimpleImputer
        from sklearn.naive_bayes import GaussianNB, MultinomialNB
        from sklearn.ensemble import RandomForestClassifier
        from sklearn import svm
        from sklearn.preprocessing import StandardScaler
        from sklearn.metrics import confusion_matrix, classification_report, accuracy_score, f
            precision_score, roc_curve, auc, matthews_corrcoef
        from sklearn.model_selection import train_test_split
        from sklearn.linear_model import LogisticRegression
In [2]: file_read = open("data_retrieved_from_blockchain.txt","r")
        input_data = file_read.readlines()
        # input_data[0]
In [3]: for i in range(len(input_data)):
            input_data[i] = (input_data[i].split(":")[1].split('"')[1]).split(',')
In [4]: input_data = np.array(input_data)
In [5]: # input_data[0].split('"')
In [6]: total_labels = input_data[:,-1].astype(np.float).astype(np.int)
        total_data = input_data[:,:-1].astype(np.float)
In [7]: # total_labels
In [8]: # total_labels = total_labels.tolist()
        for i in range(len(total_labels)):
            if(total_labels[i] <= 3):</pre>
                total_labels[i] = 1
        # total_labels = np.array(total_labels)
In [9]: # total_labels
In [10]: # total data
```

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In [11]: from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(total_data, total_labels, test_size
In [12]: \# for i in range(len(y_train)):
               if(y_train[i] == 1):
                   y_train[i] = 4
         #
               elif(y_train[i] == 4):
                  y_train[i] = 1
               elif(y_train[i] == 5):
         #
                   y_train[i] = 4
In [13]: nbclf = GaussianNB()
         rfclf = RandomForestClassifier(n_estimators=100)
         svmclf = svm.SVC(kernel='linear') #Linear Kernel
In [14]: def classify_and_report(classifier, X_train, X_test, y_train, y_test):
             classifier.fit(X_train,y_train)
             y_predicted = classifier.predict(X_test)
             #print(confusion_matrix(Y_test, Y_predicted))
               print("Naive Bayes Classifier: \n")
             print("Multiclass classification: ")
             print('Accuracy:', accuracy_score(y_test, y_predicted))
             #print('F1 score:', f1_score(Y_test, Y_predicted, average='macro'))
             #print('F1 score:', f1_score(Y_test, Y_predicted, average='weighted'))
             print('F1 score:', f1_score(y_test, y_predicted,average='macro'))
             print('Recall:', recall_score(y_test, y_predicted, average='macro'))
             print('Precision:', precision_score(y_test, y_predicted,average='macro'))
             print("Matthews Correlation Coefficient: ",matthews_corrcoef(y_test, y_predicted)
             print('Classification report:', classification_report(y_test, y_predicted))
In [15]: print ("Naive Bayes")
         classify_and_report(nbclf,X_train, X_test, y_train, y_test)
Naive Bayes
Multiclass classification:
Accuracy: 0.8
F1 score: 0.541666666666666
Recall: 0.625
Precision: 0.5833333333333333
Matthews Correlation Coefficient: 0.6968731476445666
Classification report:
                                     precision
                                                  recall f1-score
                                                                      support
                   1.00
                             1.00
                                       1.00
           1
                                                    6
           4
                   0.33
                             1.00
                                       0.50
                                                    1
           5
                   0.00
                             0.00
                                       0.00
                                                    1
           6
                   1.00
                             0.50
                                       0.67
                                                    2
                                       0.80
                                                    10
    accuracy
                   0.58
                             0.62
                                       0.54
                                                    10
  macro avg
```

weighted avg 0.83 0.80 0.78 10

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/home/ashish/anaconda3/lib/python3.7/site-packages/sklearn/metrics/classification.py:1437: Under 'precision', 'predicted', average, warn_for)

Random Forest

Multiclass classification:

Accuracy: 0.9

F1 score: 0.83333333333333333

Recall: 0.875 Precision: 0.875

Matthews Correlation Coefficient: 0.8448275862068966

Classification report: precision recall f1-score support 1 1.00 1.00 1.00 6 4 0.50 1.00 0.67 1 5 1.00 1.00 1.00 1 6 1.00 0.50 0.67 2 0.90 10 accuracy 0.83 10 macro avg 0.88 0.88 weighted avg 0.95 0.90 0.90 10

SVM

Multiclass classification:

Accuracy: 0.7 F1 score: 0.375 Recall: 0.5

Matthews Correlation Coefficient: 0.5360562674188974

Classification report: precision recall f1-score support

 1
 1.00
 1.00
 1.00
 6

 4
 0.00
 0.00
 0.00
 1

5	0.33	1.00	0.50	1
6	0.00	0.00	0.00	2
accuracy			0.70	10
macro avg	0.33	0.50	0.38	10
weighted avg	0.63	0.70	0.65	10

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