15train15testLogisticRegression

December 13, 2017

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In [1]: import pandas as pd
        from sklearn.linear_model import LogisticRegression
        from sklearn.metrics import confusion_matrix
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import roc_auc_score
        import numpy as np
        from pandas import DataFrame, Series
        from matplotlib.colors import ListedColormap
        from random import sample
        import matplotlib.pyplot as plt
In [4]: # Data of 15 people for training & testing the model, splitting the train-test set
        df_features = pd.read_csv("H:/mastersProject/activity_analyzer/LogisticRegression/Data
        df_features_3people = pd.read_csv("H:/mastersProject/activity_analyzer/LogisticRegress
        frames = [df_features, df_features_3people]
        df_15 = pd.concat(frames)
        #Drop duplicates
        df_unique = df_15.drop_duplicates(subset=['User', 'Timestamp'])
        df_unique.head()
        df_unique.describe()
        print("Shape of training and testing data", df_unique.shape)
        X_data = df_unique.values[:, 2:45]
        y_data = df_unique.values[:, 45]
        usersList = set(df_unique.values[:,0])
        print(len(usersList)+2) # Userid is for 3 people hence
Shape of training and testing data (821, 46)
15
In [6]: # Splitting the training and testing set by 33%
       X_train, X_test, y_train, y_test = train_test_split(X_data, y_data, test_size = 0.33, :
        # Fitting the logistic regression model
        clf = LogisticRegression(C=0.01, random_state=1)
```

clf.fit(X_train, y_train)

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Out[6]: LogisticRegression(C=0.01, class_weight=None, dual=False, fit_intercept=True,
                 intercept_scaling=1, max_iter=100, multi_class='ovr', n_jobs=1,
                 penalty='12', random_state=1, solver='liblinear', tol=0.0001,
                 verbose=0, warm_start=False)
In [9]: predict = clf.predict(X_test)
       logisticRegScore = clf.score(X_test, y_test)
       plt.figure(1, figsize=(4, 3))
       plt.clf()
       print("Logistic regression Score")
       print(logisticRegScore*100)
       print("Coefficients of the features")
       print(clf.coef_)
       print(X_train.shape)
        # Convert all the values to float
       float_array=np.array(X_train,dtype=np.float32)
        feature_importance = np.std(float_array, 0)*np.absolute(clf.coef_)
       np_column_list = np.array(df_unique.columns.tolist())
        column_names = np_column_list[2:45,]
        # featureimp_list = feature_importance.split(" ")
        # print("List= ", featureimp list)
       print("Column Names=", column_names)
       print("Feature importance=", feature_importance)
       print(np.sort(feature_importance))
        # TimeDiff-X
        # StdDev-x
        # TimeDiffPeaks-y
Logistic regression Score
95.2029520295
Coefficients of the features
[[ -2.27658727e-03 -4.33109059e-03 -8.19361746e-03 -1.12167114e-03
   8.39744110e-03 1.59354401e-02 7.70941170e-03 -6.50587633e-03
  -9.84746560e-03 -4.17763020e-03 -3.97150608e-03 -7.38401821e-03
  -9.37709399e-03 -6.46962045e-03 1.55653758e-04 -2.87614307e-03
   3.75548377e-03 6.85732060e-03 7.38420808e-03 4.26326533e-03
  -3.47273162e-03 -3.52646235e-03 -8.43356319e-05 1.52866626e-03
   3.17276648e-04 -5.96757471e-04 6.35076775e-05 1.31835280e-03
  -5.78737659e-04 -7.59632202e-04 -6.43644303e-02 5.89593726e-02
   7.33384679e-02 -3.54361750e-01 -2.49804221e-01 -3.33732817e-01
   4.96195335e-02 1.26346378e-01 -9.51147655e-02 -4.10671927e-01
  -2.96865368e-01 -3.98479809e-01 1.17495901e-02]]
(550, 43)
```

```
Column Names= ['Bin1,x' 'Bin2,x' 'Bin3,x' 'Bin4,x' 'Bin5,x' 'Bin6,x' 'Bin7,x' 'Bin8,x'
 'Bin9,x' 'Bin10,x' 'Bin1,y' 'Bin2,y' 'Bin3,y' 'Bin4,y' 'Bin5,y' 'Bin6,y'
 'Bin7,y' 'Bin8,y' 'Bin9,y' 'Bin10,y' 'Bin1,z' 'Bin2,z' 'Bin3,z' 'Bin4,z'
 'Bin5,z' 'Bin6,z' 'Bin7,z' 'Bin8,z' 'Bin9,z' 'Bin10,z' 'TimeDiffPeaks-x'
 'TimeDiffPeaks-y' 'TimeDiffPeaks-z' 'AvgAbsDiff-x' 'AvgAbsDiff-y'
 'AvgAbsDiff-z' 'AvgAcc-x' 'AvgAcc-y' 'AvgAcc-z' 'StdDev-x' 'StdDev-y'
 'StdDev-z' 'AvgResAcc']
Feature importance= [[ 4.51210672e-05
                                         2.13806574e-04
                                                          5.73472398e-04
                                                                           7.16694391e-05
   5.27913677e-04
                   9.82845945e-04
                                      4.95280130e-04
                                                       4.45199729e-04
   5.17130669e-04
                    1.10893246e-04
                                      7.46570933e-05
                                                       2.53391396e-04
   4.24106620e-04
                     3.34478990e-04
                                      7.73687539e-06
                                                       1.35693186e-04
    1.89577957e-04
                     4.09273708e-04
                                      4.35308498e-04
                                                       1.39048616e-04
   7.55109139e-05
                     2.03519943e-04
                                      6.03784730e-06
                                                       8.37054101e-05
    1.81514225e-05
                     3.14207686e-05
                                      2.72818876e-06
                                                       4.87373135e-05
    1.64332315e-05
                     1.18137449e-05
                                      8.30391359e-01
                                                       9.21619905e-01
    1.00655259e+00
                     8.56465079e-01
                                      5.37591537e-01
                                                       7.00916065e-01
    1.24014775e-01
                     2.20914156e-01
                                      1.20557788e-01
                                                       1.24561416e+00
    7.86127595e-01
                     1.06455267e+00
                                      4.79028656e-03]]
[[ 2.72818876e-06
                     6.03784730e-06
                                      7.73687539e-06 1.18137449e-05
    1.64332315e-05
                     1.81514225e-05
                                      3.14207686e-05
                                                       4.51210672e-05
    4.87373135e-05
                     7.16694391e-05
                                      7.46570933e-05
                                                       7.55109139e-05
   8.37054101e-05
                     1.10893246e-04
                                      1.35693186e-04
                                                       1.39048616e-04
    1.89577957e-04
                     2.03519943e-04
                                      2.13806574e-04
                                                       2.53391396e-04
   3.34478990e-04
                     4.09273708e-04
                                      4.24106620e-04
                                                       4.35308498e-04
   4.45199729e-04
                     4.95280130e-04
                                      5.17130669e-04
                                                       5.27913677e-04
   5.73472398e-04
                     9.82845945e-04
                                      4.79028656e-03
                                                       1.20557788e-01
    1.24014775e-01
                                      5.37591537e-01
                                                       7.00916065e-01
                     2.20914156e-01
   7.86127595e-01
                     8.30391359e-01
                                      8.56465079e-01
                                                       9.21619905e-01
    1.00655259e+00
                     1.06455267e+00
                                      1.24561416e+00]]
In [10]: from sklearn.metrics import confusion_matrix
         cm = confusion matrix(y test, predict, labels=["walking", "running"])
        print(cm)
        6]
[[143
 [ 7 115]]
In [15]: #Area under ROC
        from sklearn.metrics import roc_curve
        from sklearn.metrics import auc
        from sklearn.preprocessing import LabelEncoder
         import matplotlib.pyplot as plt
         # # Encode the labels for ROC plot
         def encode_label(y_test):
```

```
y_test_binary = []
    for y in y_test:
        if y == "walking":
            y_test_binary.append(1)
        else:
            y_test_binary.append(0)
    return y_test_binary
y_test_binary = encode_label(y_test)
y_predict_binary = encode_label(predict)
# Compute fpr, tpr, thresholds and roc auc
# fpr, tpr, thresholds = roc_curve(y_test_binary, probas_[:, 1])
fpr, tpr, thresholds = roc_curve(y_test_binary, y_predict_binary)
roc_auc = auc(fpr, tpr)
print(roc_auc)
# Plot ROC curve
plt.plot(fpr, tpr, label='AUC = %0.2f' % roc_auc, color="green")
plt.plot([0, 1], [0, 1], 'k--', color="salmon") # random predictions curve, 50% accu
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.0])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('Area under ROC curve')
plt.legend(loc="lower right")
# plt.savefig('H:/mastersProject/activity_analyzer/LogisticRegression/roc_lr', dpi=20
plt.show()
```

0.951177247222

