

The comments/sections provided are your cues to perform the assignment. You don't need to limit yourself to the number of rows/cells provided. You can add additional rows in each section to add more lines of code.

Assignment: Classify Kinematic Data

If at any point in time you need help on solving this assignment, view our demo video to understand the different steps of the code.

Happy coding!

Classify Kinematic Data

DESCRIPTION

You are supposed to detect whether the person is running or walking based on the sensor data collected from iOS device. The dataset contains a single file which represents sensor data samples collected from

accelerometer and gyroscope from iPhone 5c in 10 seconds interval and ~5.4/second frequency.

Objective: Practice classification based on Naive Bayes algorithm. Identify the predictors that can be influential. **Actions to Perform:**

1. Load the kinematics dataset as measured on mobile sensors from the file "run_or_walk.csv." 2. List the columns in the dataset.

5. Generate a classification report using Scikit-learn. 6. Repeat the model once using only the acceleration values as predictors and then using only the gyro

3. Let the target variable "y" be the activity, and assign all the columns after it to "x." 4. Using Scikit-learn, fit a Gaussian Naive Bayes model and observe the accuracy.

values as predictors. 7. Comment on the difference in accuracy between both models.

import matplotlib.pyplot as plot

RangeIndex: 88588 entries, 0 to 88587 Data columns (total 11 columns):

%matplotlib inline

date

3

username

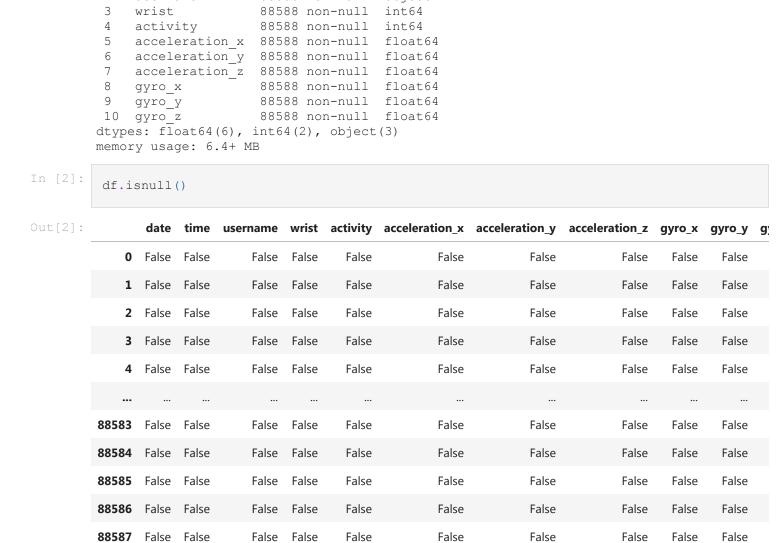
wrist

- Load the kinematics dataset as measured on mobile sensors from the file "run_or_walk.csv."
 - #import required libraries and dataset import pandas as pd

df = pd.read csv("run or walk.csv") df.info() <class 'pandas.core.frame.DataFrame'>

Dtype

88588 non-null object 88588 non-null object 88588 non-null object



#Split the dataset in to training and testing sets from sklearn.model selection import train test split X, y = df.iloc[:, 5:].values, df.iloc[:, 4].values

print(y_test[0:10])

[1 0 0 1 1 1 0 1 1 1]

classifier = GaussianNB()

#print the accuracy score

print(conf mat)

accuracy macro avq weighted avg

df.info()

0 date

1 time

gyro_y 10 gyro_z

[[8610 706] [63 8339]]

Gyro values as predictors

#print the accuracy

print(conf mat)

[[6528 4100] [2145 4945]]

classifier.fit(X train,y train)

y predict = classifier.predict(X test)

memory usage: 6.4+ MB

Acceleration values as predictors

[[8583 699] [90 8346]]

classifier.fit(X train,y train)

(70870, 6)

after it to "X"

In [4]:

In [6]:

dtype='object')

88588 rows × 11 columns

df.columns

List the columns in the dataset.

Out[3]: Index(['date', 'time', 'username', 'wrist', 'activity', 'acceleration_x', 'acceleration_y', 'acceleration_z', 'gyro_x', 'gyro_y', 'gyro_z'],

Let the target variable "y" be the activity, and assign all the columns

X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=

print(X_train.shape)

from sklearn.naive bayes import GaussianNB

from sklearn.metrics import accuracy score accuracy = accuracy score(y predict,y test)

from sklearn.metrics import confusion matrix conf_mat = confusion_matrix(y_predict,y_test)

#print the classification report

<class 'pandas.core.frame.DataFrame'> RangeIndex: 88588 entries, 0 to 88587 Data columns (total 11 columns):

Column Non-Null Count Dtype

6 acceleration_y 88588 non-null float64 acceleration_z 88588 non-null float64 8 gyro_x 88588 non-null float64

dtypes: $\overline{\text{float64}(6)}$, int64(2), object(3)

target names = ["Walk", "Run"]

y predict = classifier.predict(X test)

#Fit the training data into Gaussian Naive Bayes classifier

Using Scikit-learn, fit a Gaussian Naive Bayes model and observe the

print(accuracy) 0.9554690145614629 #print the confusion matrix

print(classification report(y test, y predict, target names=target names))

0.96 0.95

9045

precision recall f1-score support

0.96 0.96 0.96 17718 0.96 0.96 0.96 17718 0.96 0.96 0.96 17718

Repeat the model once using only the acceleration values as predictors and then using only the gyro values as predictors.

Generate a classification report using Scikit-learn

from sklearn.metrics import classification report

0.92 0.99 0.99 0.92

88588 non-null object 88588 non-null int64 88588 non-null int64 username 2 wrist 88588 non-null int64 activity 88588 non-null int64 acceleration_x 88588 non-null float64

88588 non-null object

88588 non-null object

88588 non-null float64 88588 non-null float64

#Repeat the model once using Acceleration values as predictors

from sklearn.model selection import train test split X, y = df.iloc[:, [5,6,7]].values, <math>df.iloc[:, 4].valuesX train, X test, y train, y test = train test split(X, y, test size=0.2, random state= classifier.fit(X_train,y_train) y predict = classifier.predict(X test) #print the accuracy print("Accuracy:",accuracy_score(y_predict,y_test)) Accuracy: 0.9565978101365843 In [14]: #print the confusion matrix conf mat = confusion matrix(y predict, y test) print(conf mat)

> #Repeat the model once using Gyro values as predictors X, y = df.iloc[:, [8,9,10]].values, df.iloc[:, 4].values

print("Accuracy:",accuracy score(y predict,y test))

conf mat = confusion_matrix(y_predict,y_test)

Accuracy: 0.6475335816683598 #print the confusion matrix

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=

Comment on the difference in accuracy between both models. The accelerometer provides changes in device's velocity along 3 axes, which is a crucial piece of information about how one moves his hand. And a gyroscope delivers the rate at which a device rotates

around a spatial axis which carries probably fewer insights for distinguishing running and walking activities.

Using our model, the accuracy score obtained using 'Acceleration value' as 'predictor' is 95%. And the accuracy score obtained using 'Gyro value' as 'predictor' is 64%. From the above results, we could cleary say accelerometer data provides better results when distinguishing 'run or walk' compared to Gyroscope data.