



# PREDICTING HUMAN ACTIVITIES USING SMARTPHONES

(Data Science Capstone)

# PRESENTATION FLOW

- Explore the research question
- Discuss the Dataset Chosen
- Identify the Purpose of the Study
- Exploratory analysis of Data
- Classification Models Explored
- Concluding Remarks

# WHAT DOES YOUR SMARTPHONE KNOW ABOUT YOU?



# ABOUT DATASET

The Human Activity Recognition database was built from the recordings of 30 study participants performing activities of daily living (ADL) while carrying a waist-mounted smartphone with embedded inertial sensors.

## **Attribute information :**

- Triaxial acceleration from the accelerometer (total acceleration) and the estimated body acceleration.
- Triaxial Angular velocity from the gyroscope.
- A 561-feature vector with time and frequency domain variables.
- Its activity label.
- An identifier of the subject who carried out the experiment.

# THE PURPOSE OF THE STUDY

*To classify activities into one of the six activities performed.*

Tracking Human Activities :

- ❖ WALKING
- ❖ WALKING\_UPSTAIRS
- ❖ WALKING\_DOWNSTAIRS
- ❖ SITTING
- ❖ STANDING
- ❖ LAYING

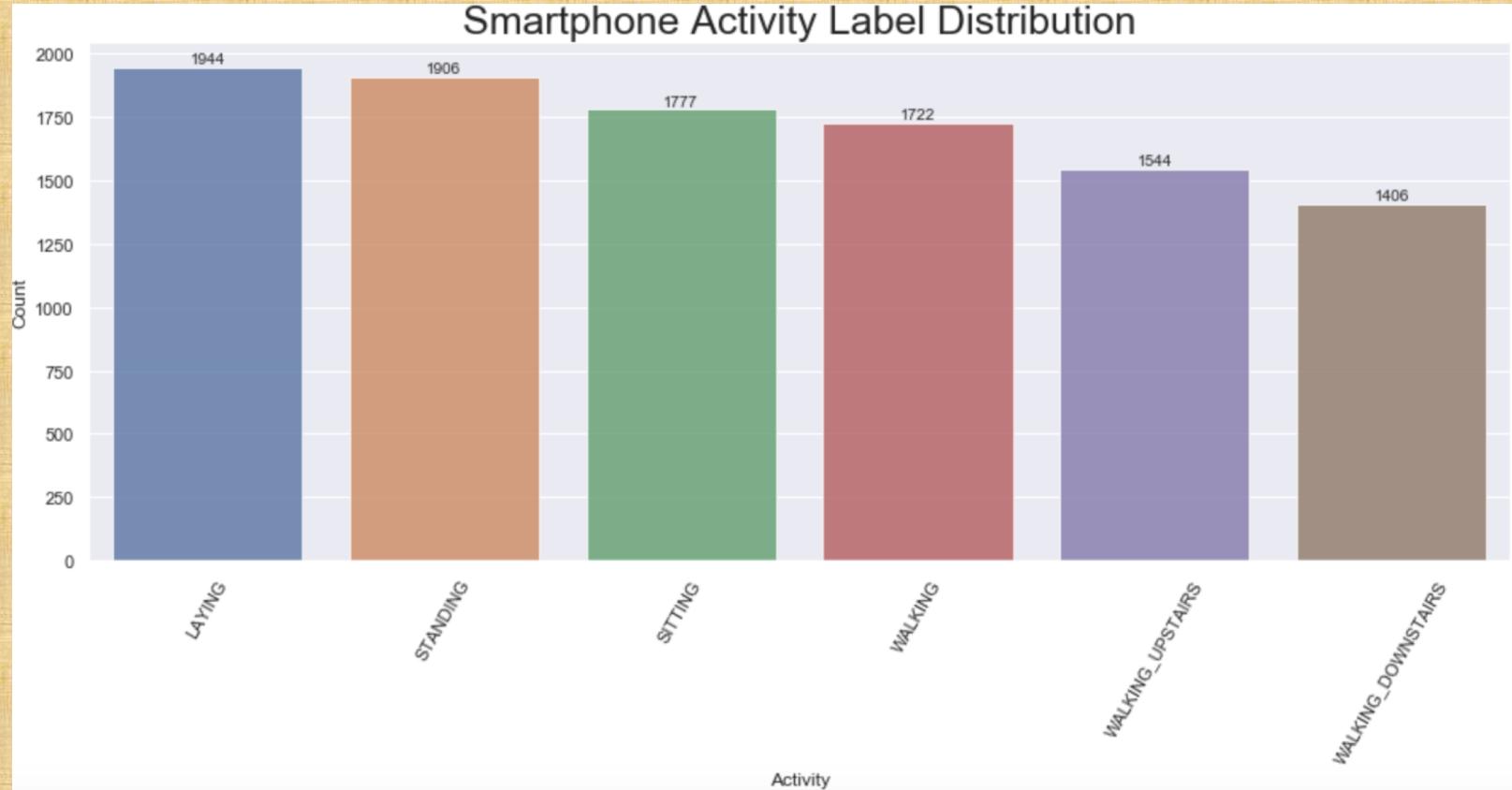
# EXPLORATORY ANALYSIS OF DATA

- Mainly there are 'acceleration' and 'gyroscope' features. A few 'gravity' features as well.
- Total 563 features
- Except 'Activity' and 'subject' features there is only numerical data. Fortunately there are no missing values.
- The features seem to have a main name and some information on how they have been computed attached

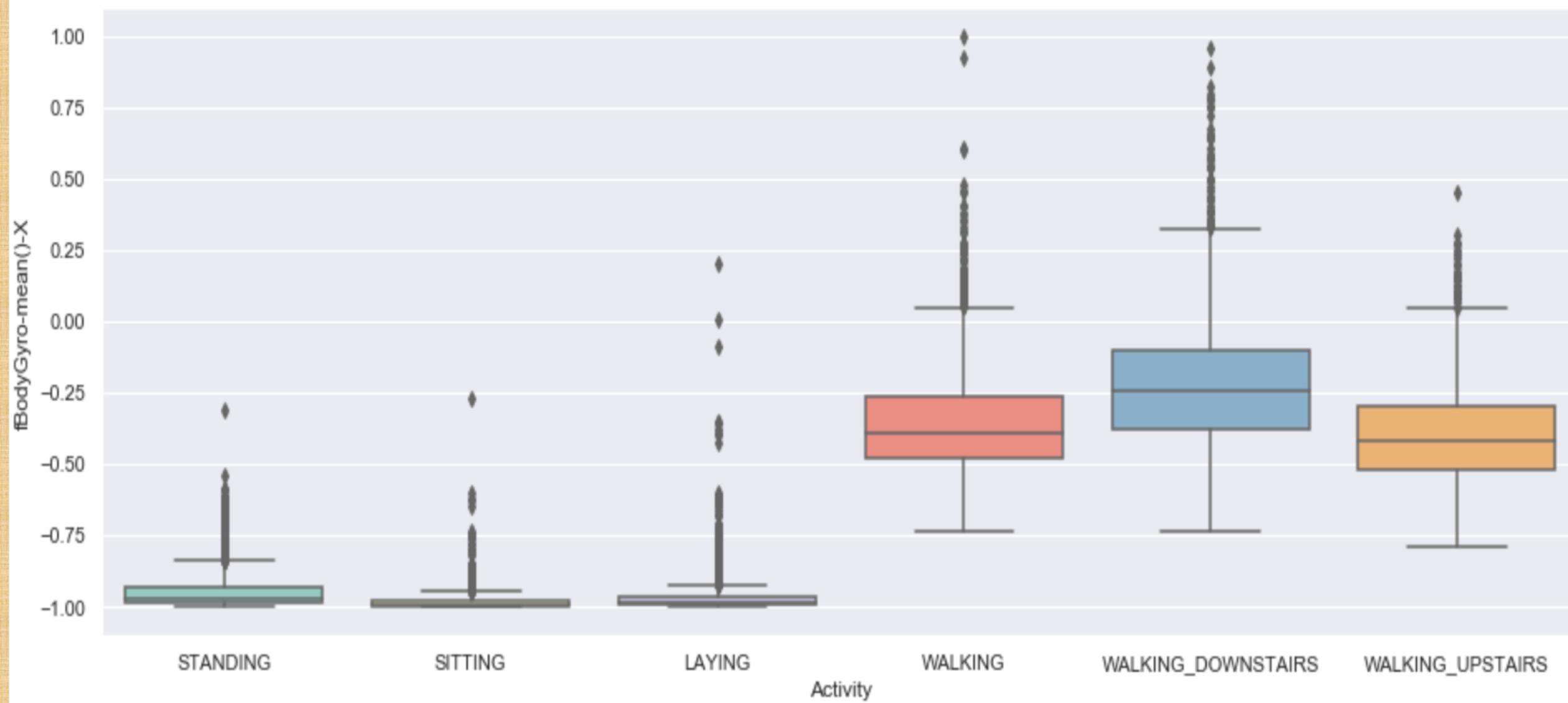
**Data:** <https://www.kaggle.com/morrisb/what-does-your-smartphone-know-about-you/data>

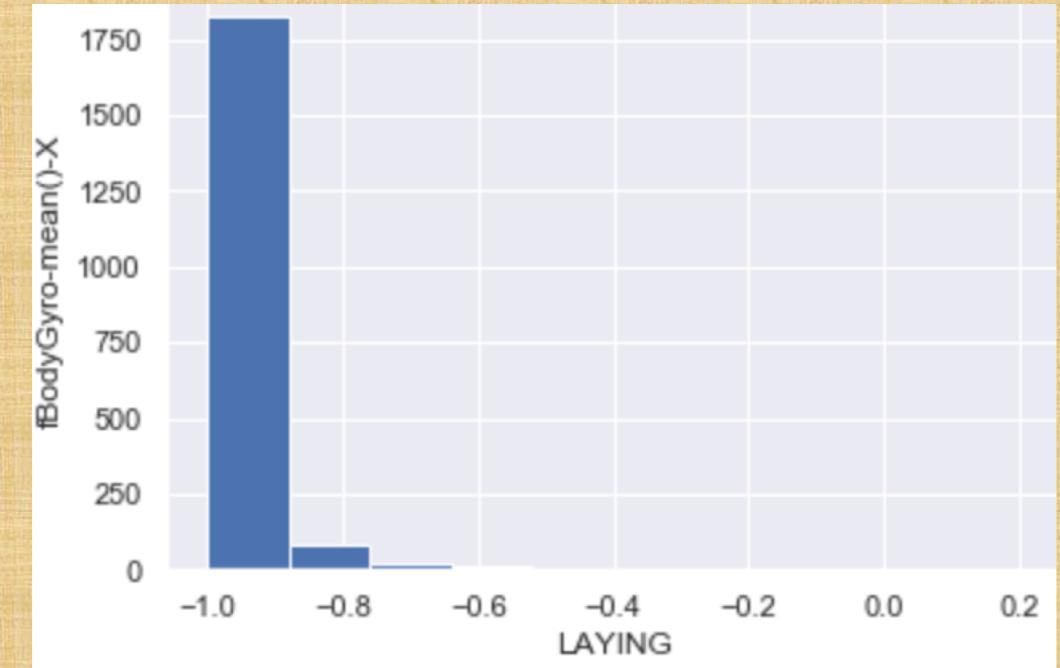
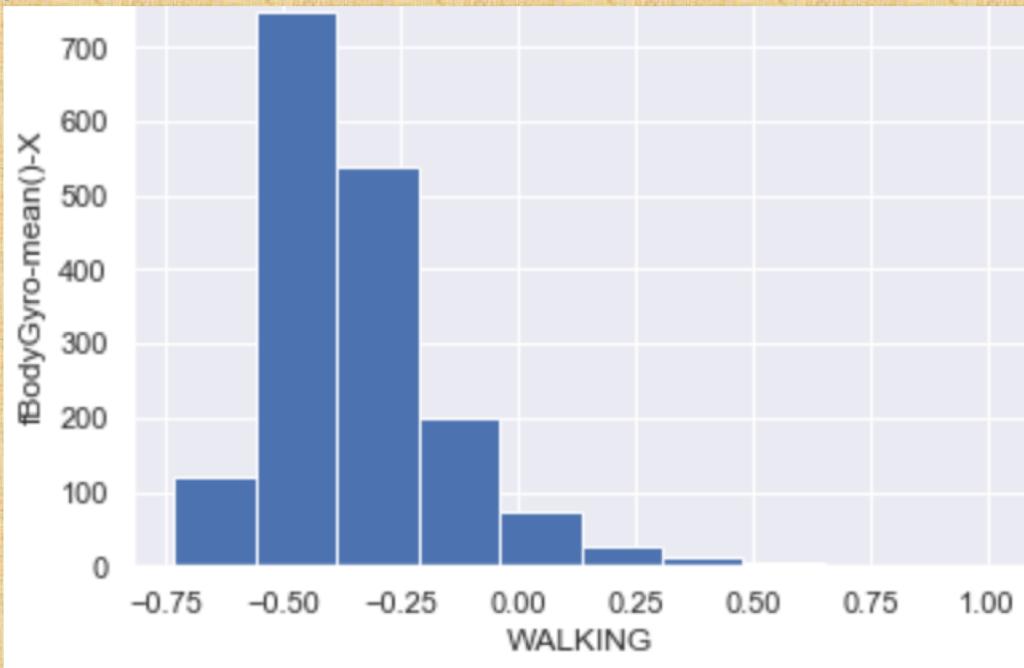
|                      | count |
|----------------------|-------|
| fBodyAcc             | 79    |
| fBodyGyro            | 79    |
| fBodyAccJerk         | 79    |
| tGravityAcc          | 40    |
| tBodyAcc             | 40    |
| tBodyGyroJerk        | 40    |
| tBodyGyro            | 40    |
| tBodyAccJerk         | 40    |
| tBodyAccMag          | 13    |
| tGravityAccMag       | 13    |
| tBodyAccJerkMag      | 13    |
| tBodyGyroMag         | 13    |
| tBodyGyroJerkMag     | 13    |
| fBodyAccMag          | 13    |
| fBodyBodyAccJerkMag  | 13    |
| fBodyBodyGyroMag     | 13    |
| fBodyBodyGyroJerkMag | 13    |
| angle                | 7     |
| subject              | 1     |
| Activity             | 1     |

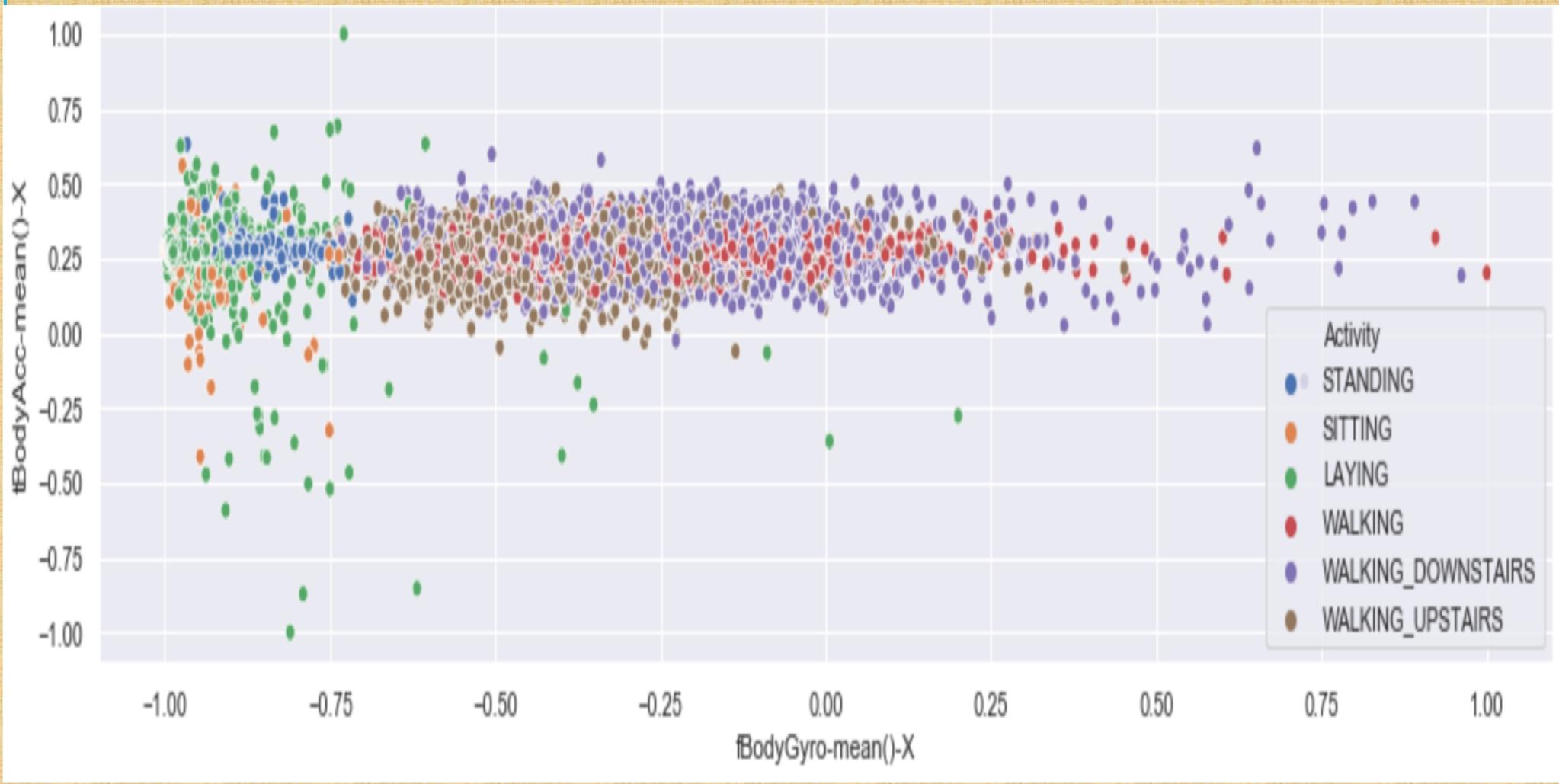
# DATA ANALYSIS

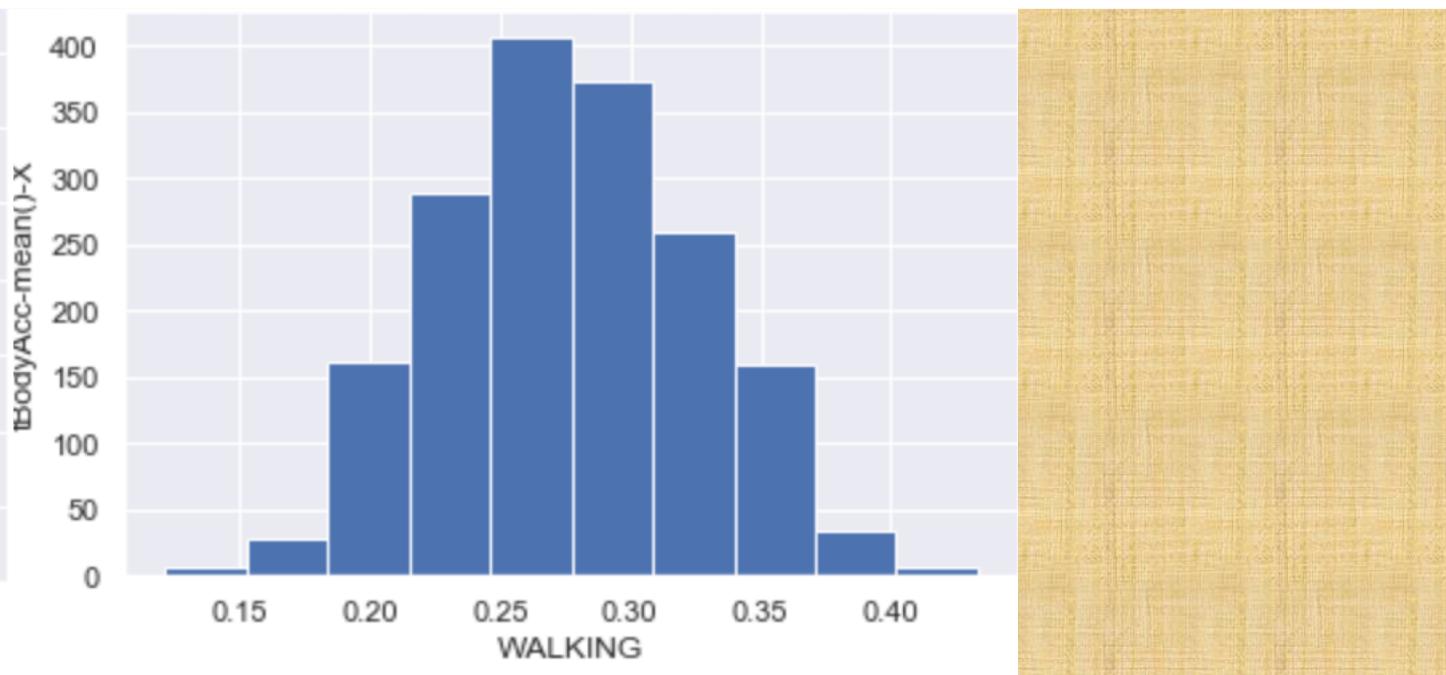
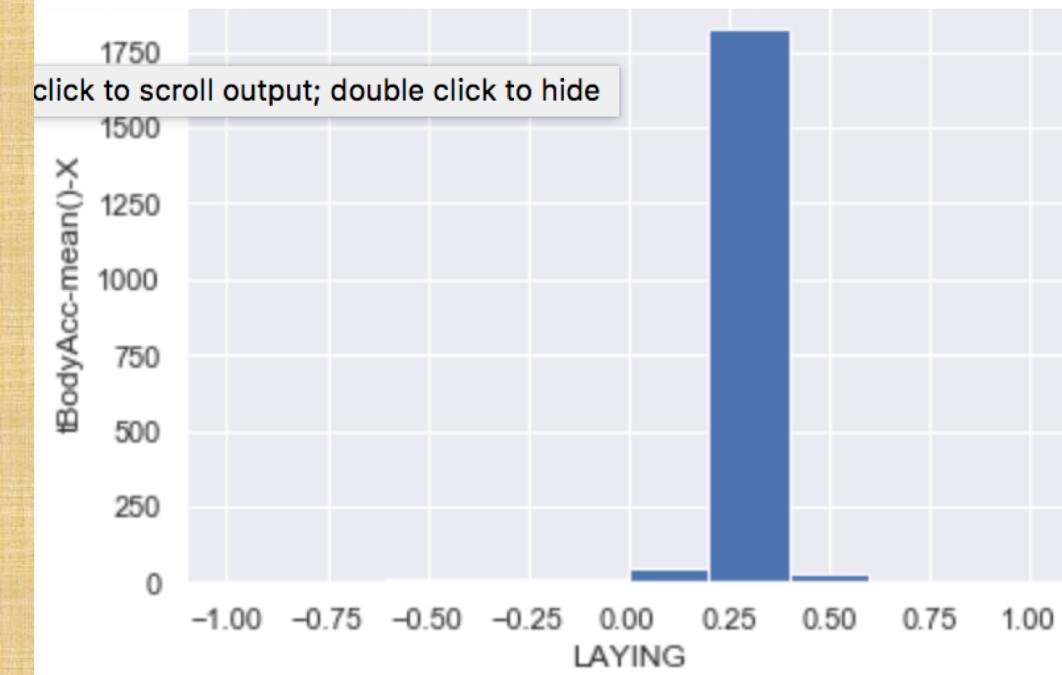


Disregarding the possibility of flawed data, the participants seem to **walk roughly 10% faster downwards**.









# CLASSIFICATION MODELS

➤ Before PCA

|                            | Score    |
|----------------------------|----------|
| Model                      |          |
| GradientBoostingClassifier | 0.984466 |
| LogisticRegression         | 0.983689 |
| XGBClassifier              | 0.983301 |
| RandomForestClassifier     | 0.969709 |
| KNeighborsClassifier       | 0.961942 |
| SVC                        | 0.958835 |
| DecisionTreeClassifier     | 0.931650 |
| GaussianNB                 | 0.730874 |

➤ After PCA

|                            | Score    |
|----------------------------|----------|
| Model                      |          |
| LogisticRegression         | 0.953398 |
| KNeighborsClassifier       | 0.946019 |
| GradientBoostingClassifier | 0.940583 |
| XGBClassifier              | 0.927767 |
| SVC                        | 0.919223 |
| RandomForestClassifier     | 0.894369 |
| DecisionTreeClassifier     | 0.838835 |
| GaussianNB                 | 0.833398 |

## CONCLUSION AFTER PCA

- Data analysis shows that PCA is not helping to improve model accuracy as there must be some loss of information so we are not considering PCA for improving model performance
- The best performing model is Logistic Regression with improved performance after tuning and accuracy of 98%
- Vanilla gradient boosting without tuning performed the best, with tuning it didn't help much
- Logistic Regression is way faster than Gradient boosting

# GRADIENT BOOSTING CLASSIFIER

|  | precision | recall | f1-score | support |
|--|-----------|--------|----------|---------|
| click to scroll output; double click to hide |           |        |          |         |
| LAYING                                       | 1.00      | 1.00   | 1.00     | 502     |
| SITTING                                      | 0.95      | 0.95   | 0.95     | 443     |
| STANDING                                     | 0.96      | 0.95   | 0.95     | 457     |
| WALKING                                      | 0.99      | 0.99   | 0.99     | 446     |
| WALKING_UPSTAIRS                             | 0.99      | 0.97   | 0.98     | 343     |
| WALKING_DOWNSTAIRS                           | 0.97      | 0.99   | 0.98     | 384     |
| avg / total                                  | 0.98      | 0.98   | 0.98     | 2575    |

|                    | LAYING | SITTING | STANDING | WALKING | WALKING_UPSTAIRS | WALKING_DOWNSTAIRS |
|--------------------|--------|---------|----------|---------|------------------|--------------------|
| LAYING             | 502    | 0       | 0        | 0       | 0                | 0                  |
| SITTING            | 0      | 422     | 20       | 0       | 0                | 1                  |
| STANDING           | 0      | 21      | 434      | 0       | 0                | 2                  |
| WALKING            | 0      | 0       | 0        | 442     | 2                | 2                  |
| WALKING_UPSTAIRS   | 0      | 0       | 0        | 4       | 334              | 5                  |
| WALKING_DOWNSTAIRS | 0      | 0       | 0        | 1       | 3                | 380                |

# LOGISTIC REGRESSION

|                    | precision | recall | f1-score | support |
|--------------------|-----------|--------|----------|---------|
| LAYING             | 1.00      | 1.00   | 1.00     | 502     |
| SITTING            | 0.95      | 0.95   | 0.95     | 443     |
| STANDING           | 0.95      | 0.96   | 0.96     | 457     |
| WALKING            | 1.00      | 1.00   | 1.00     | 446     |
| WALKING_UPSTAIRS   | 1.00      | 1.00   | 1.00     | 343     |
| WALKING_DOWNSTAIRS | 1.00      | 1.00   | 1.00     | 384     |
| avg / total        | 0.98      | 0.98   | 0.98     | 2575    |

|                    | LAYING | SITTING | STANDING | WALKING | WALKING_UPSTAIRS | WALKING_DOWNSTAIRS |
|--------------------|--------|---------|----------|---------|------------------|--------------------|
| LAYING             | 502    | 0       | 0        | 0       | 0                | 0                  |
| SITTING            | 0      | 422     | 21       | 0       | 0                | 0                  |
| STANDING           | 0      | 20      | 437      | 0       | 0                | 0                  |
| WALKING            | 0      | 0       | 0        | 446     | 0                | 0                  |
| WALKING_UPSTAIRS   | 0      | 0       | 0        | 0       | 343              | 0                  |
| WALKING_DOWNSTAIRS | 0      | 0       | 0        | 1       | 0                | 383                |

# CONCLUDING REMARKS

- The smartphone has enough data to determine what its user is doing (**98%**: 6 activities)
- Logistic Regression is giving more accurate and improved performance compared to other classification models also faster compared to gradient boost
- This data can be used by many healthcare and fitness apps which can provide more insights
- As the dataset only provide a small amount of data the variance of the results can be high

Thank  
you!!