

# Human Gait Activity Recognition

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# **1. Abstract**

As we know Activity Recognition is useful in many fields as pervasive healthcare and surveillance, The recognition of human gait can be useful to identify the characteristics of the places or physical space such as walking, walking downstairs ...etc.

In this Project a method of recognizing gait activities using acceleration data obtained from a smartphone, the acceleration signal were segmented based on the strides (Peaks and Valley ) then Feature vector of the segmented signals was extracted, Which was used to train five Classifier Using:

- Support Vector Machine
- Decision Tree
- Random Forest
- Naive Bayes
- K-Nearest neighbors

## **Our Dataset:**

It was collected from data consist of five Activities [ stand, walking, stair up, stair down, sit]

Where each activity has its own accelerometer data and time stamps

## 2. Method

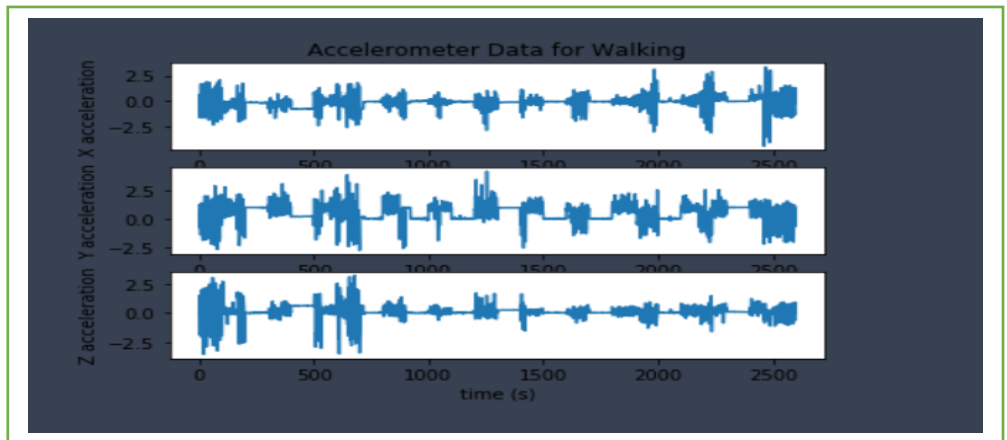
Our Method is Classified into 3 Phases

### a. Phase 1: Preprocessing

#### i. Load Data

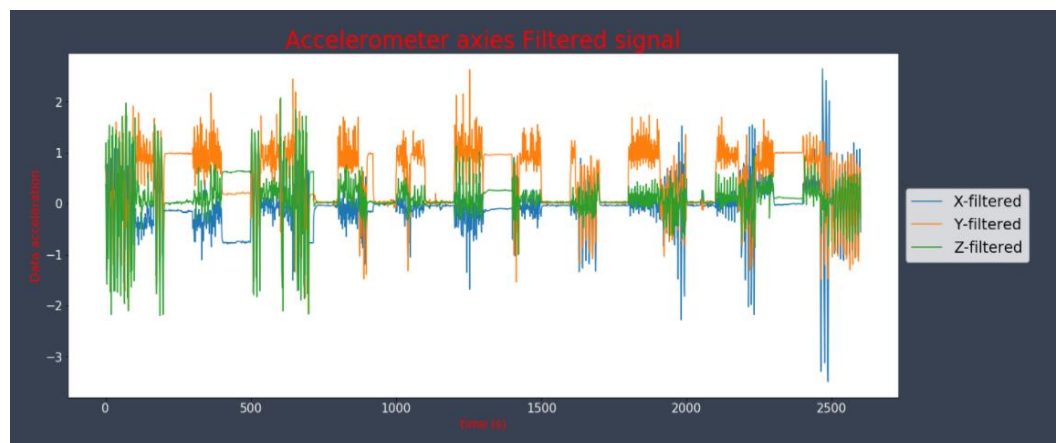
First, we load our data, then we find that it consists of 5 columns “ time Elapsed, x, y, z, and label “ and it consists of 2600 row

We plot Axes of Accelerometer Data:



#### ii. Applying Low\_pass Filter

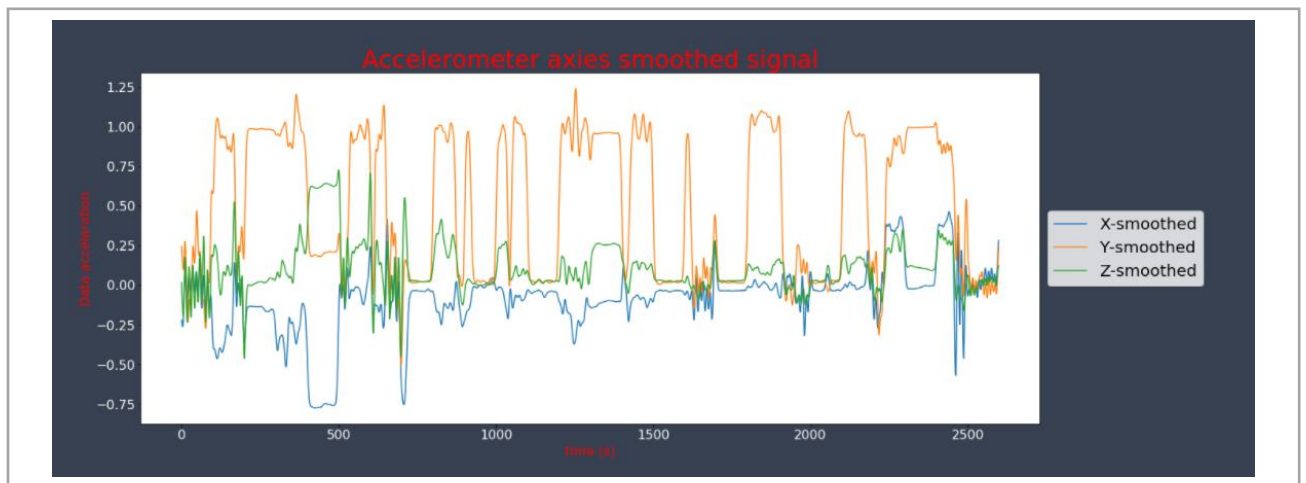
A low-pass filter is applied to accX, accY, accZ, we use the coefficient a which filter the highest frequencies with a **value of 0.5** the plot after filtration was :



### iii. Applying smoothing

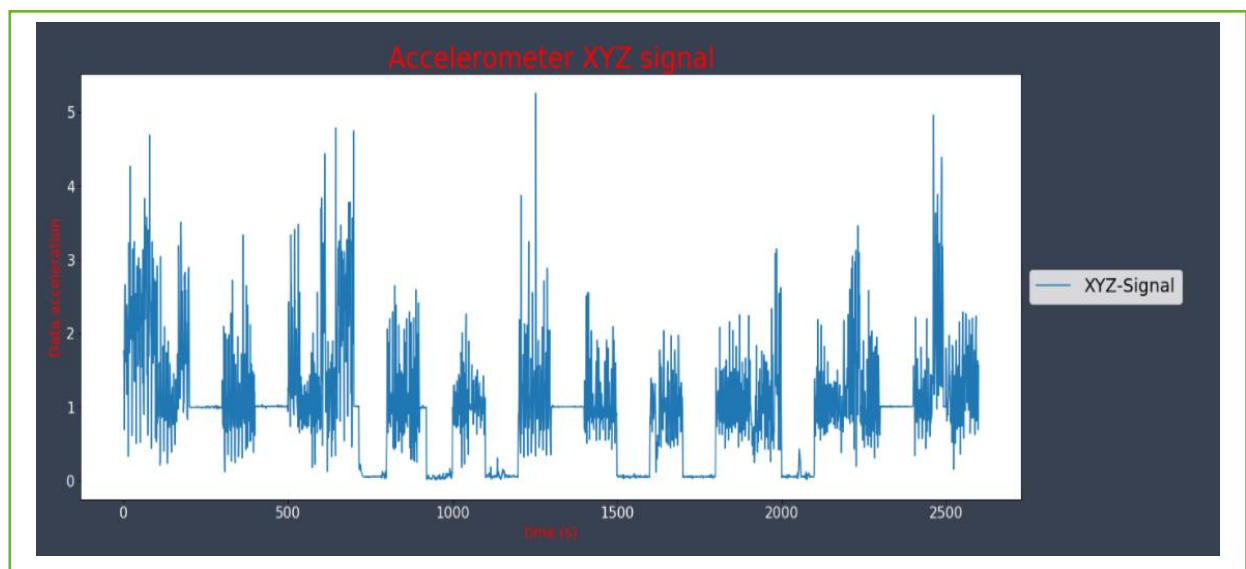
By using gaussian filter1d from SciPy library with sigma value = 4 we smoothed the

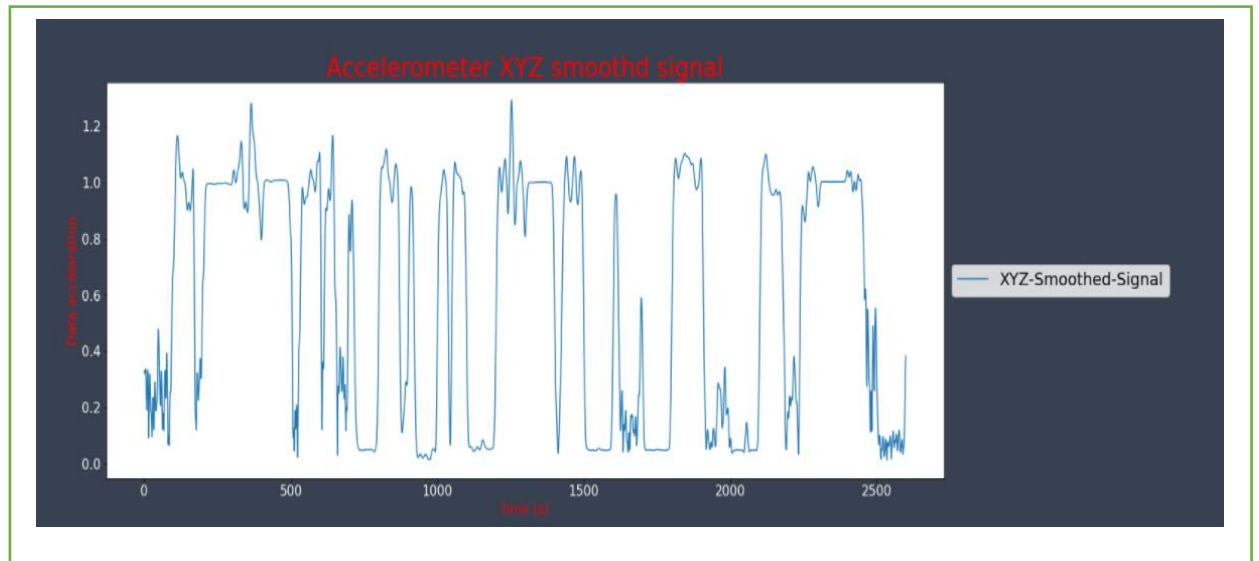
accelerometer axes data and the plot :



### iv. Applying XYZ

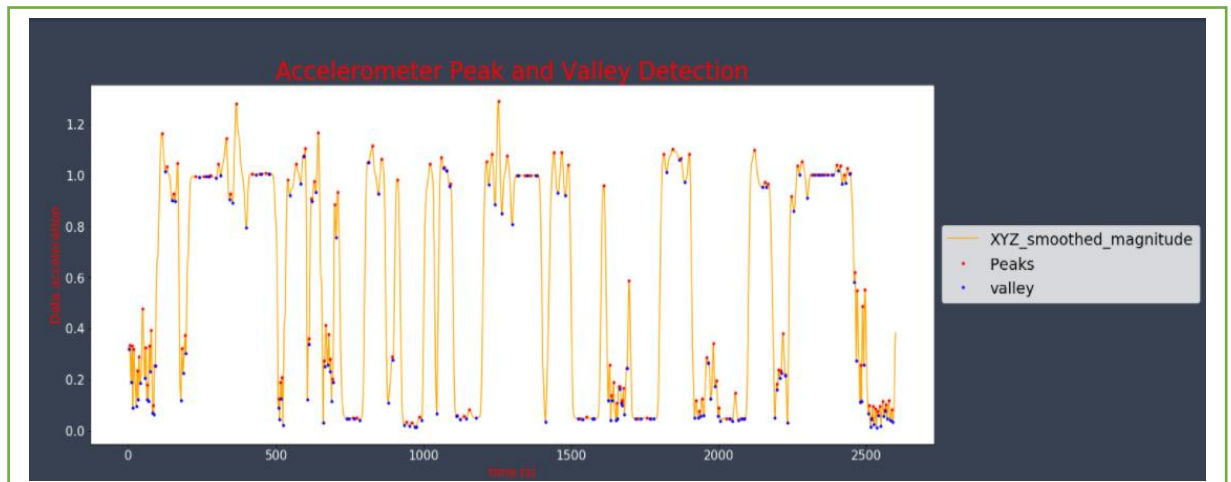
Now we try to find the magnitude of the three axes and we make this on smoothed and filtered x, y, z acceleration data by applying  $\sqrt{x^2 + y^2 + z^2}$   
Plot for Data after applying:





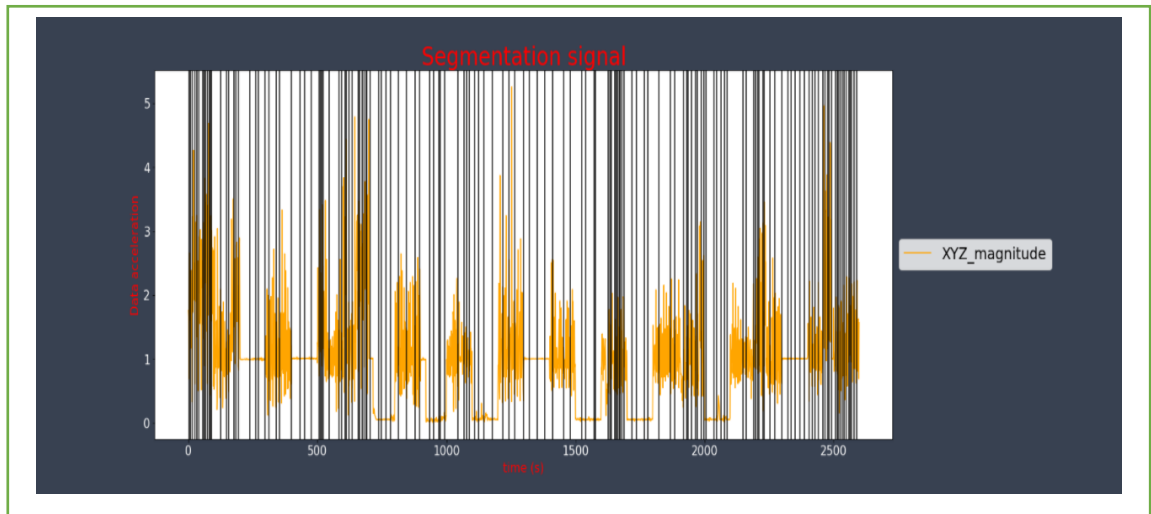
## v. Detection of Peaks and Valleys

We use `find_peaks` function from SciPy library we use it also for detecting valleys by multiplying XYZ\_smoothed values with -1 and the peaks



## vi. Segmentation

Applying segmentation depends on valleys of dataset XYZ\_Smoothed then we apply this segmentation on XYZ\_magnitude:



## b. Phase 2: Feature extraction

### i. Peak Height

To find the height of the Peak we started with detecting the maximum point and the minimum point in the peak and then we subtracted the minimum from the maximum.

### ii. Mean

To find the mean of the wave we detected the period of the wave and used all the points in the wave to calculate the mean.

### iii. Peak width

To find the width of the peak we used time steps by subtracting the time of the end of the wave from the beginning of the wave.

#### **iv. Standard Division**

To calculate the standard deviation we did the same as in the mean calculation phase.

#### **v. Data Frame for feature**

After we extract features we put this feature in the data frame (df\_feature) then we labeled each features row according to the actual label for example: as we know them from time 0 to 99 there is an activity done which is walking, the first 12 valley points are from 0 to 90 so they are labeled as Walking and so on.  
We found feature data consists of 151 rows and 5 columns.

### **c. Phase 3: Classification**

#### **i. Train Test Split**

We split the Feature data set to X And Y where X is (acc\_mean, acc\_Std, peak\_height, peak\_width) Y labels.

Then we use function train test split from the scikit-learn library and split data with 0.1 for test and random state = 49.

#### **ii. Preprocessing for Features data**

we use here preprocessing label encoder to convert Y to label and Standard Scaler to make feature scaling to X data.

### iii. Algorithms Used

- Support Vector machine:

We use Kernel 'rbf' , random state =0 and fit with train data and predict with test

- K-Nearest Neighbors:

We use n\_neighbor =3 metric= 'minkowski'  
p=3

- Random forest:

n\_estimators = 17, criterion = 'entropy',  
random\_state = 0

- Decision Tree: max\_depth = 5

- Naive Bayes:

## 3. Results and conclusions:

To Calculate accuracy we use the accuracy score function from metrics in sklearn

Algorithm	Accuracy
SVC	0.6875
Decision Tree	0.6875
Random forest	0.625
KNN	0.5625
Naive Bayes	0.5625



#### **4. Challenges:**

We Faced a lot of challenges as the paper of Recognition of Gait Activities Using Acceleration Data from A Smartphone and A Wearable Device. Has many functions with no identification for it like Power( ), Forward Direction, segmenting, power for feature extraction, and we couldn't produce the classification result as TPR and TNR because it's not mentioned in the paper. Also dataset we didn't find a perfect dataset, So we started collecting the dataset manually from different accelerometers read for activity and it was so small.