

# Smartphone-based Activity Recognition and Multi-sensor Fusion based Indoor Positioning System

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## 1. Introduction

## 2. System Framework

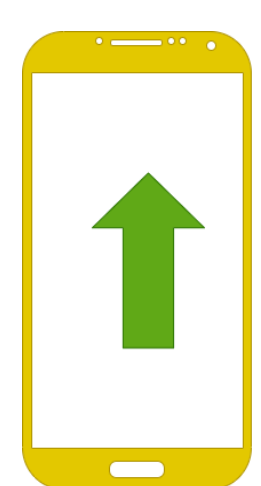
### Motivation

over 80% of time people spent indoors

without mature ubiquitous Global Positioning Systems and maps for indoors

Applications: smart health, smart home, asset management..

Multi sensors, e.g., built-in to smartphone enabling accurately positioning



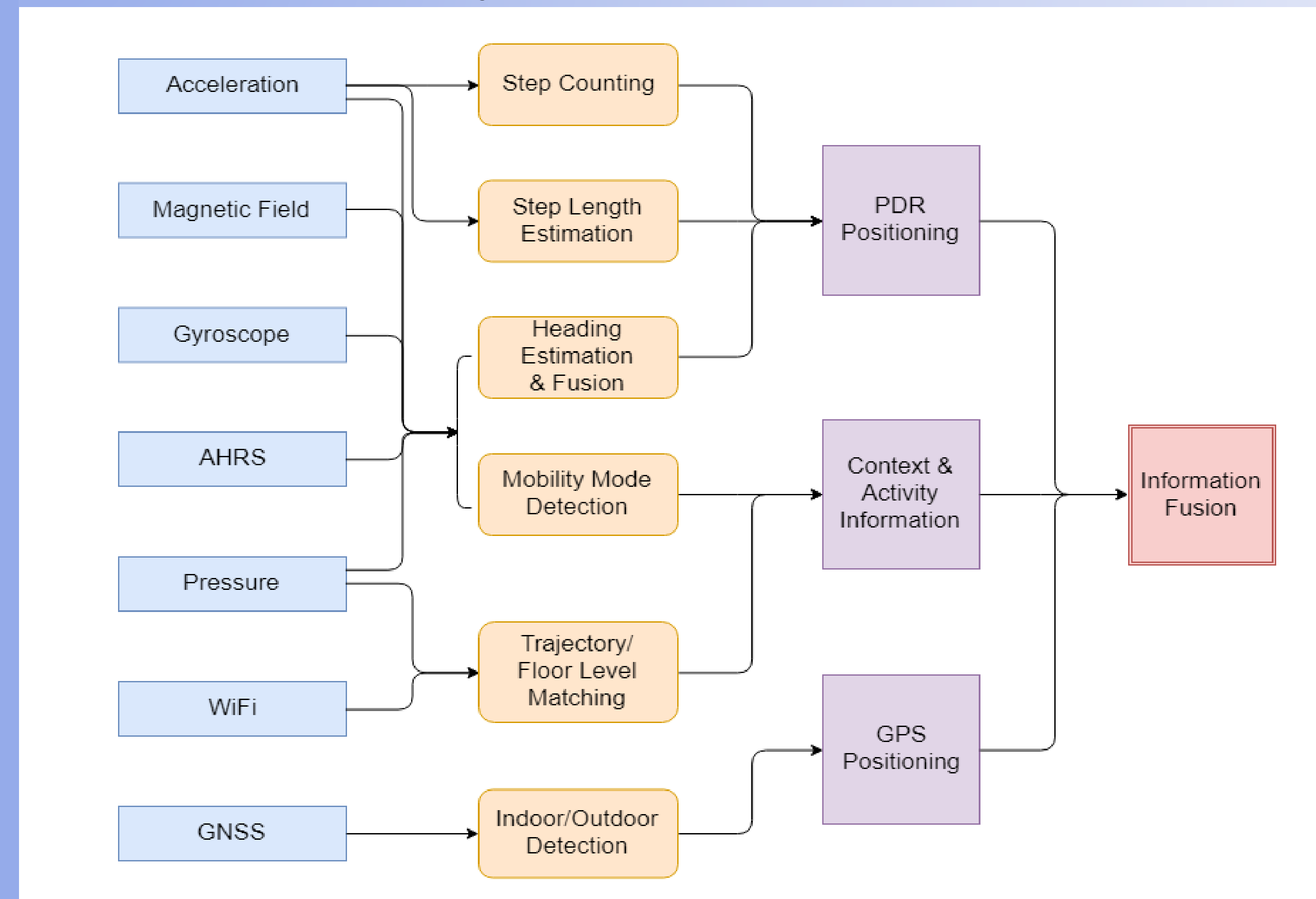
Where am I?

Which building?

Which floor?

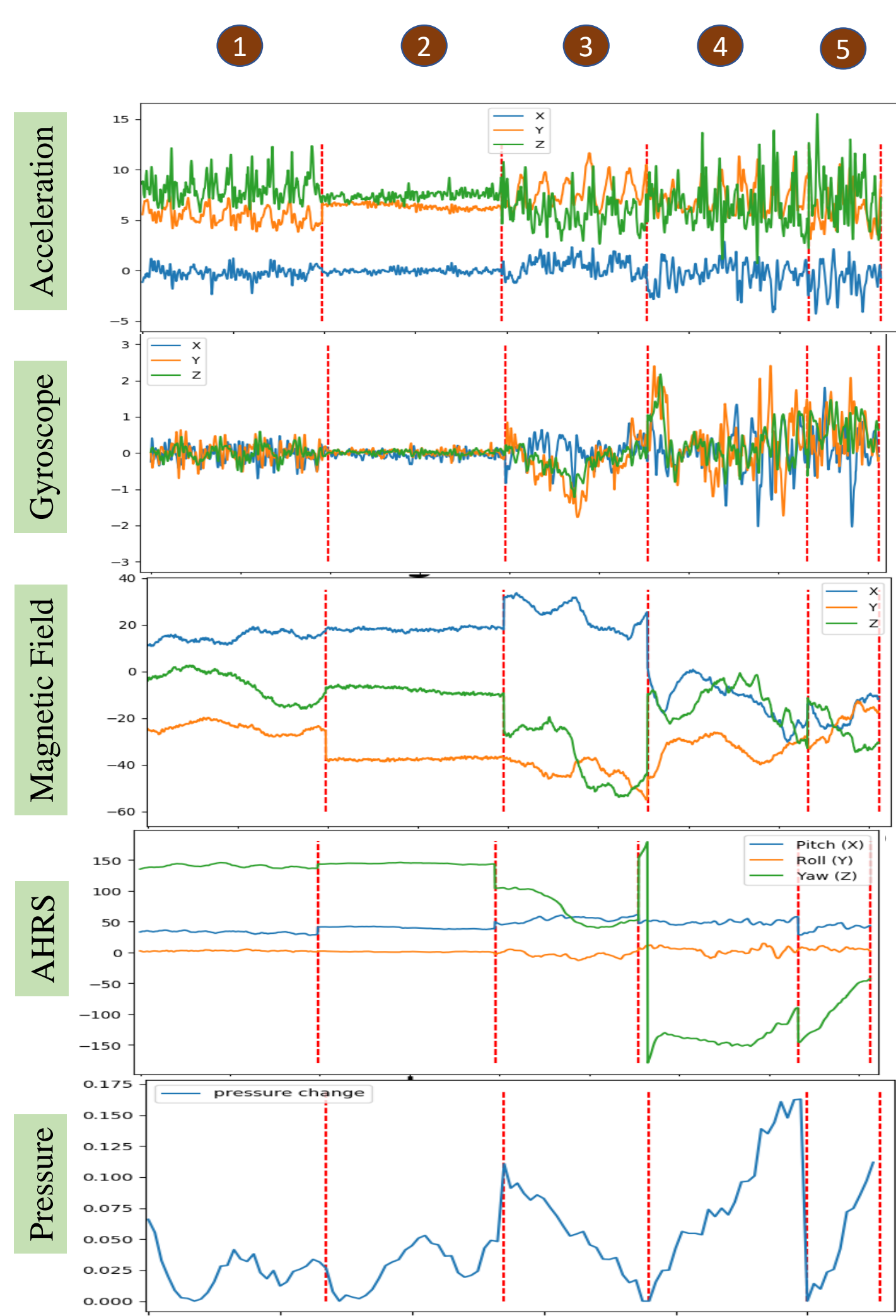
Which path?

...



## 3. Mobility Mode Detection

## 4. Trajectory/Floor ID Determination

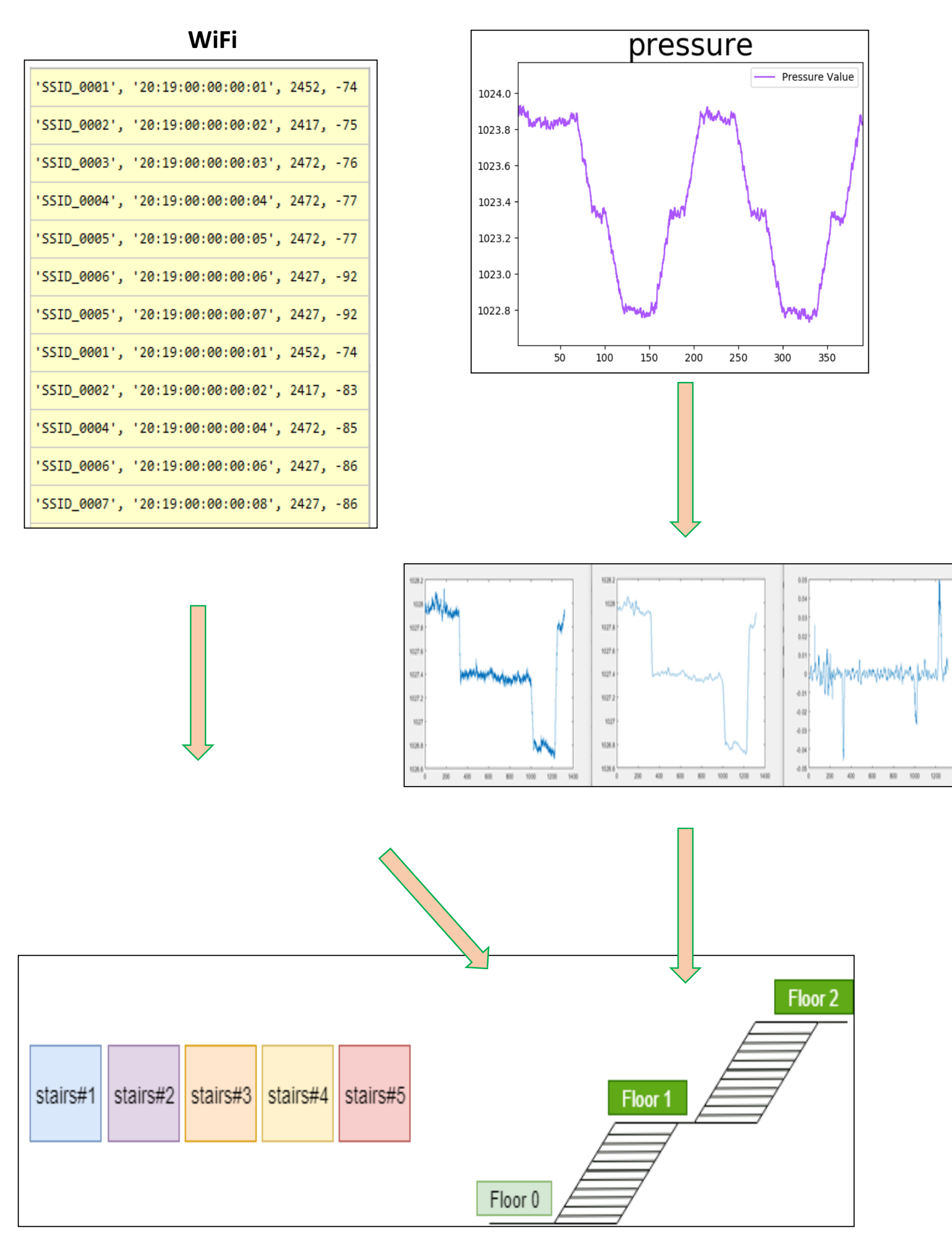


Different modes of mobility can be detected using machine learning or deep learning algorithms using multi-sensor data.

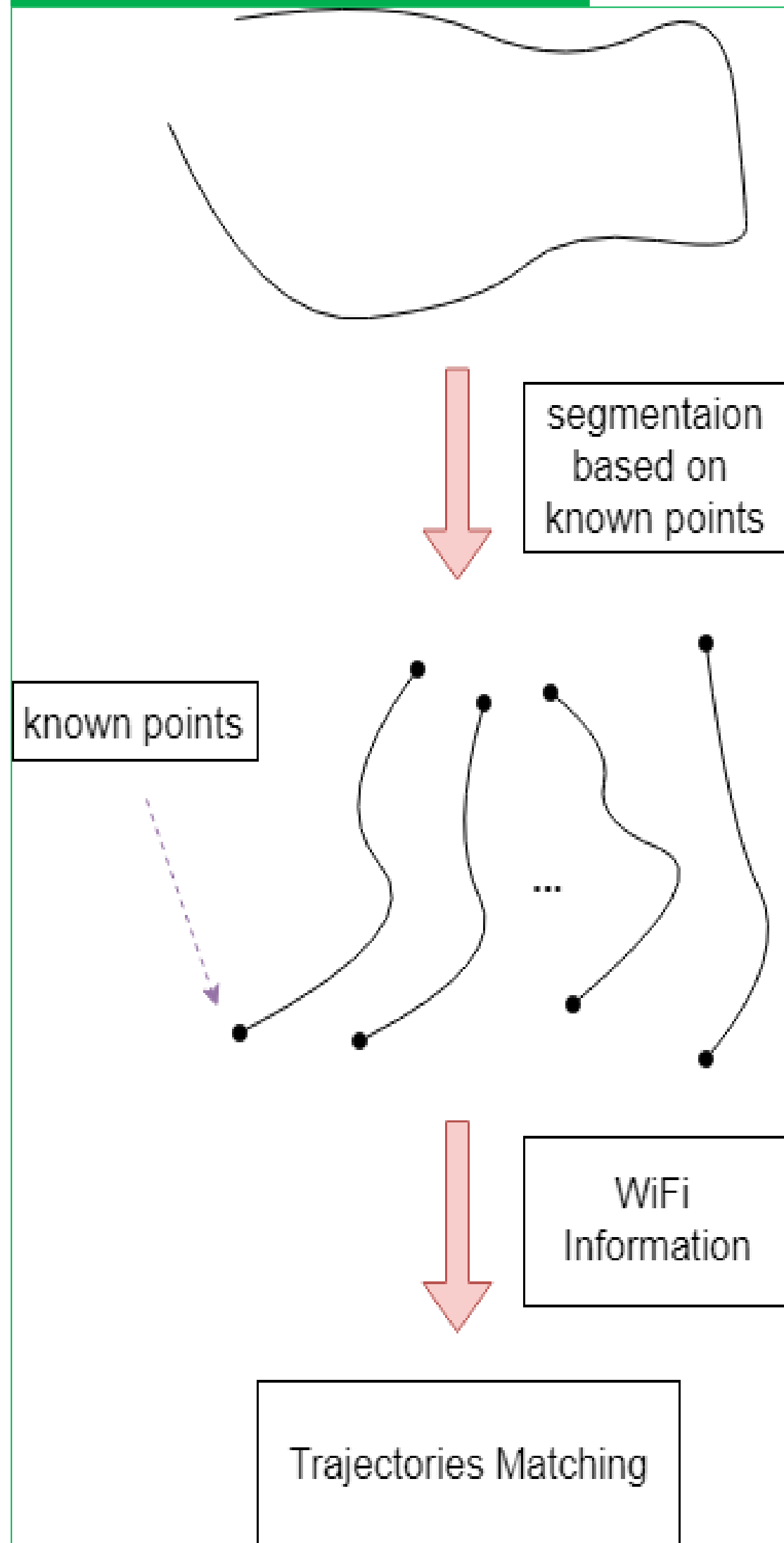
Supervised Learning Algorithms

- 1 walking
- 2 stationary
- 3 ascending (stairs)
- 4 descending
- 5 turning

### Floor Determination



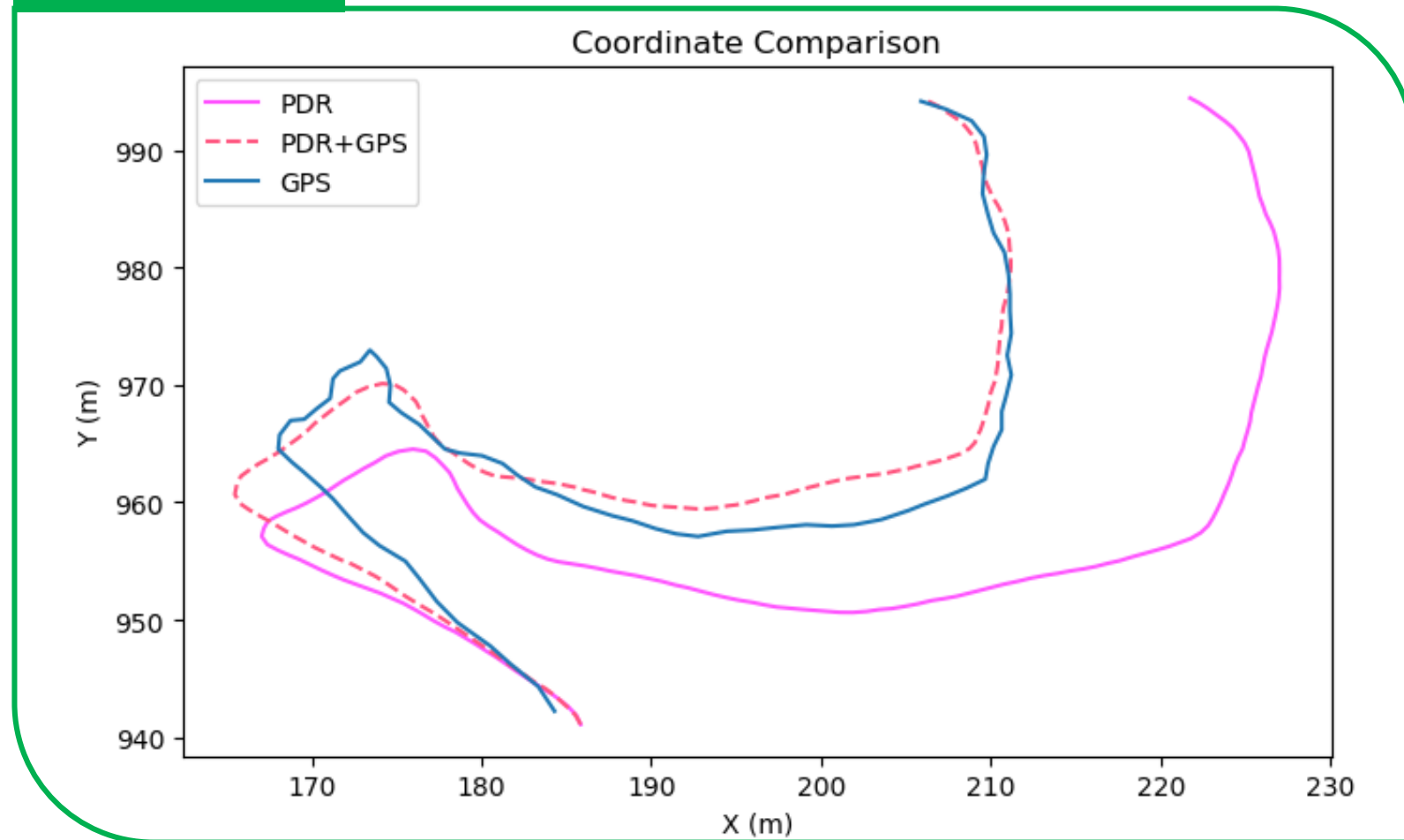
### Trajectory Segmentation and Matching



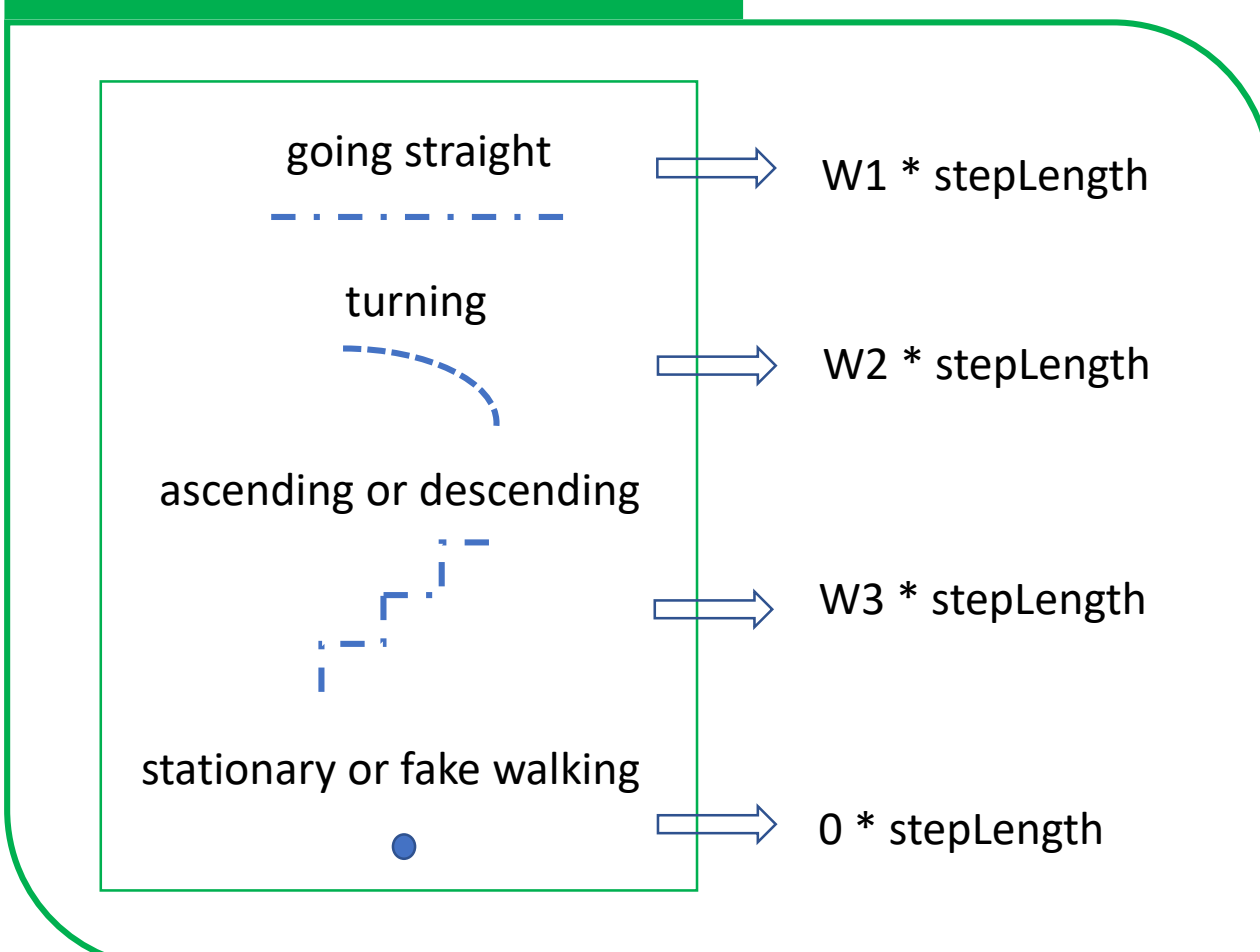
## 5. Information Fusion

## 6. Positioning Result Display

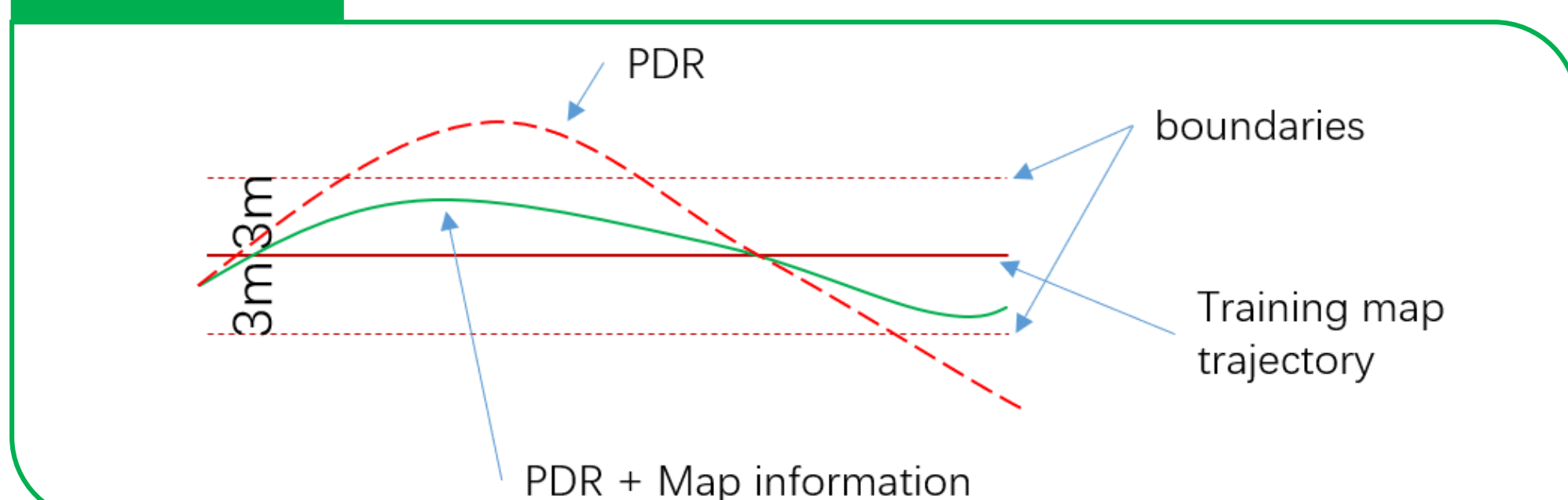
### PDR+GPS



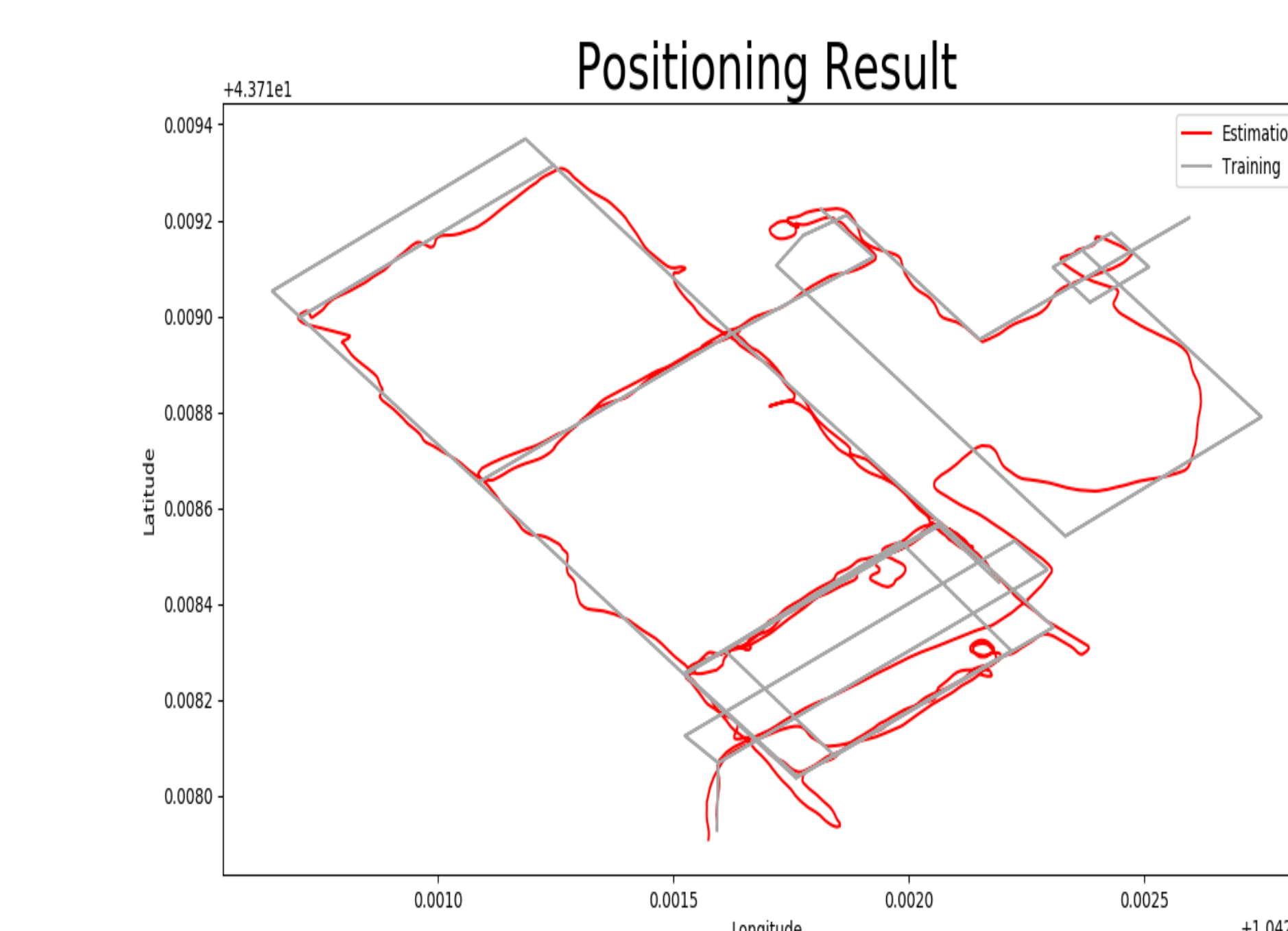
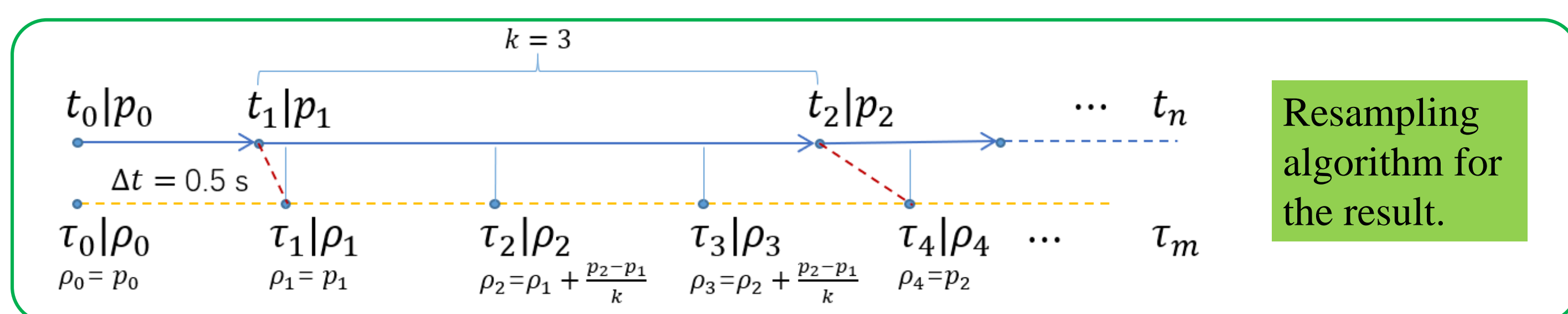
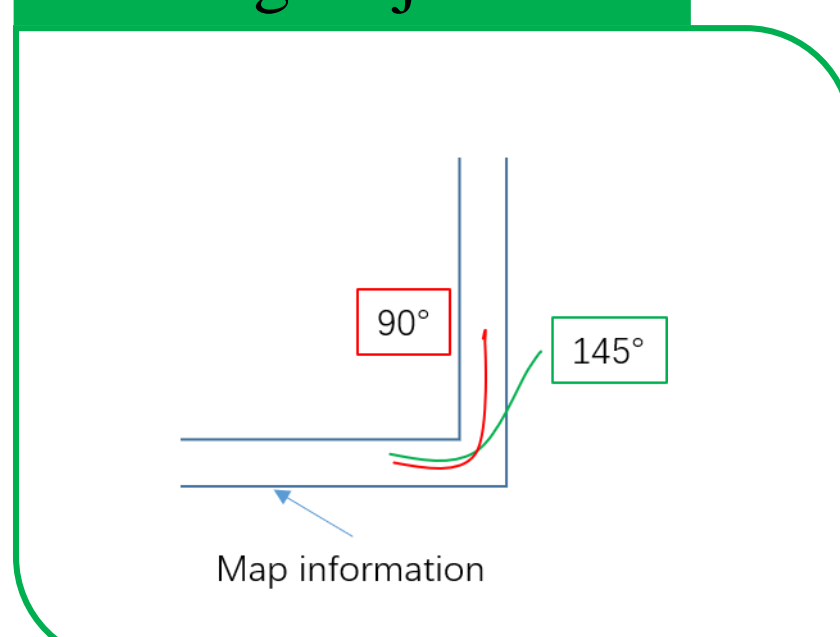
### Motion Mode + PDR



### PDR+MAP



### Heading Adjustment



The red line is the estimated result based on our system. The gray line is the training trajectories extracted from training map.