

Indoor Positioning in Smartphones by Adopting Inertial Sensors and Radio Information

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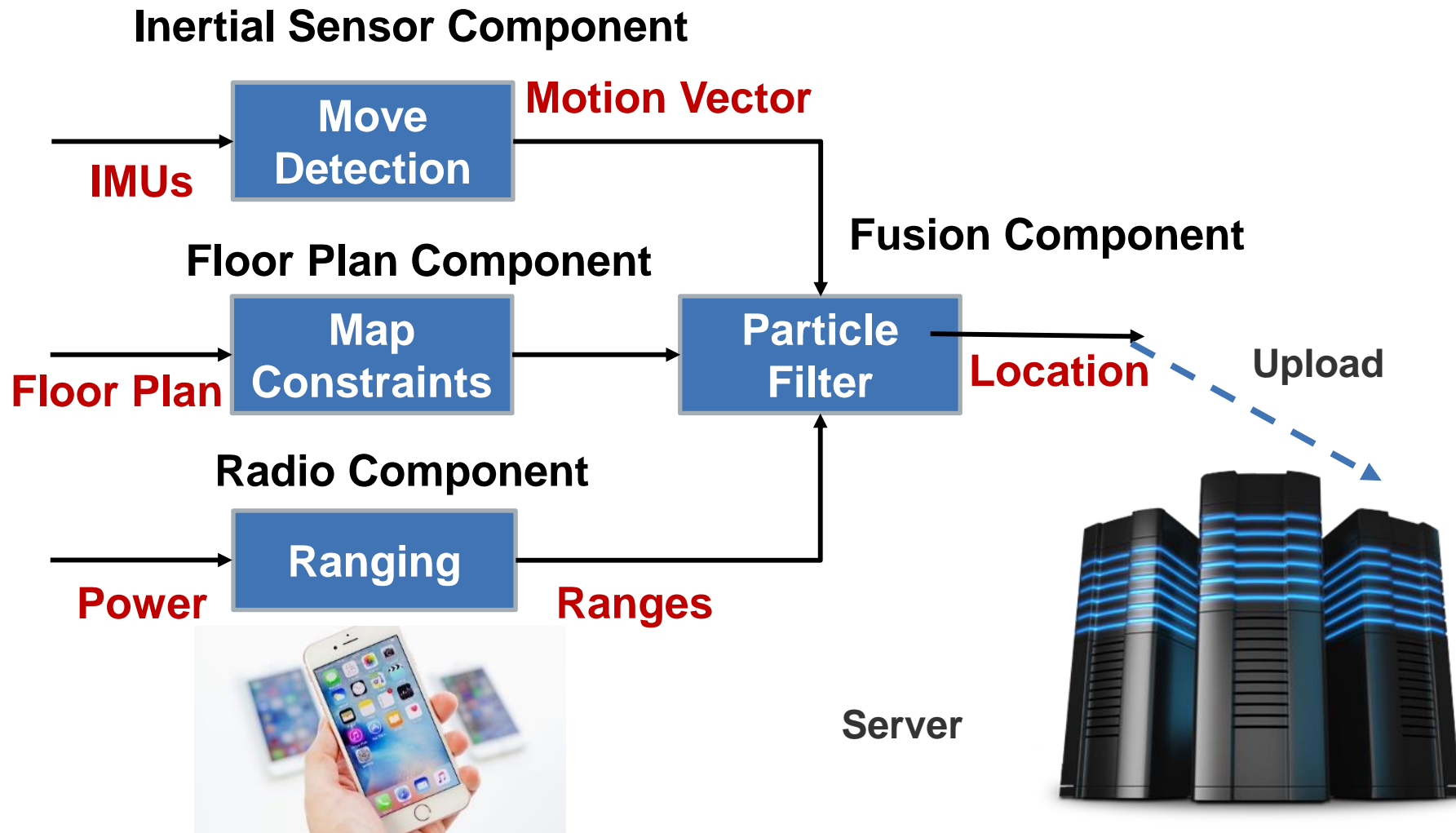
Outline

- > Motivation.
- > Proposed Tracking Algorithm
 - > Radio Information Component.
 - > Inertial Sensor Component.
 - > Floor Plan Information Component.
 - > Data Fusion Component.
- > Implementation of the Tracking Algorithm in Smartphone
- > Preliminary Experiment and Results
- > Conclusions and possible future work.

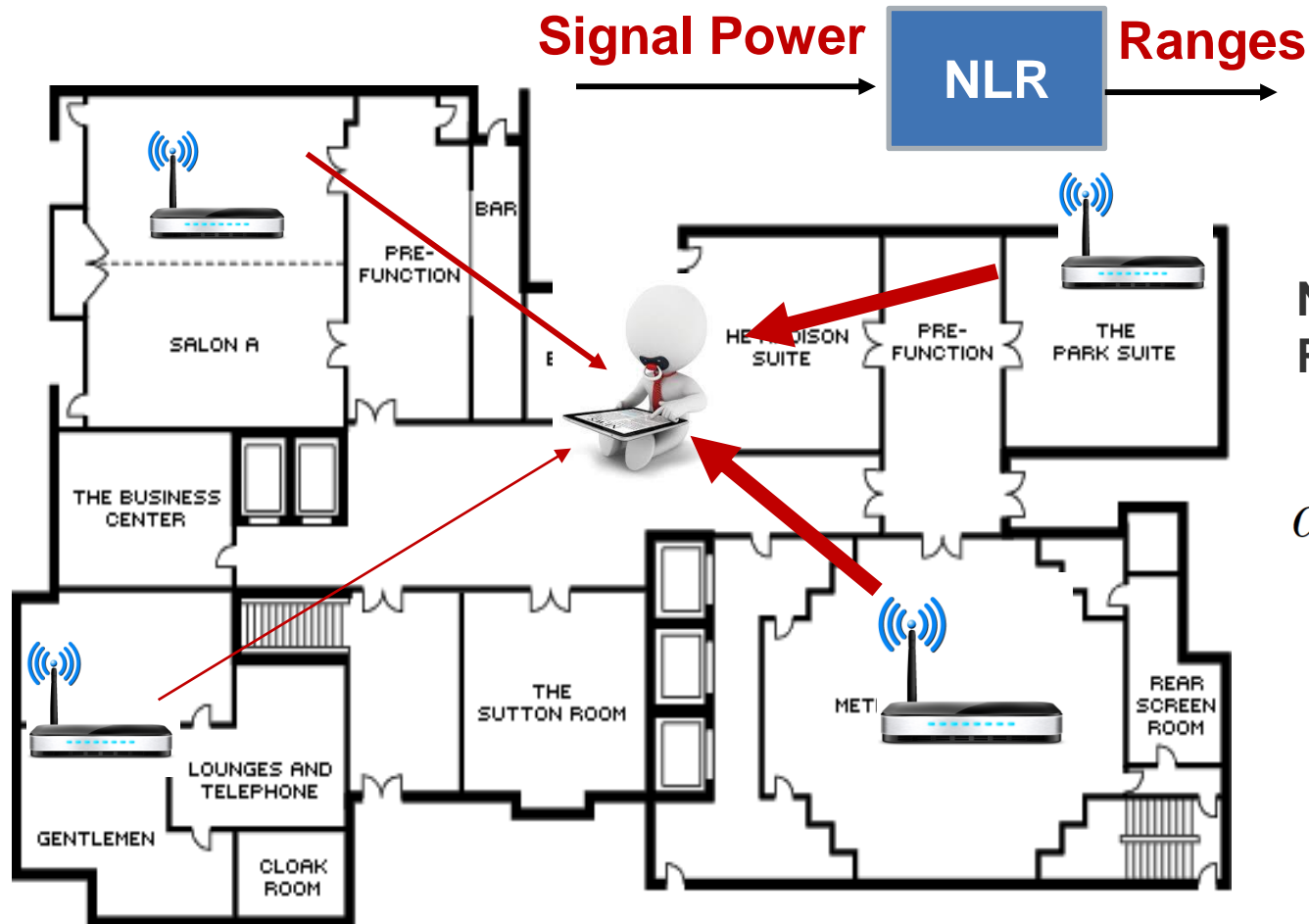
Motivation

- > Locating in indoor environments has become a key issue for emerging location based application.
 - > Mobile phones important interface user-environment.
- > Crucial for pervasive mobile applications:
 - > Tracking of medical equipment
 - > Store navigation
 - > Parking lots
 - > Tracking in disaster areas
- > No easy and accurate solution nowadays.
- > No accepted standards do yet exist.
- > SwissSenseSynergy project

Indoor Tracking System



Radio Information Component



Non-Linear
Regression Model [1]

$$\hat{d}_i = \alpha_i \cdot e^{\beta_i \cdot \text{RSS}_i}$$

Inertial Sensor Component

Accelerometer:

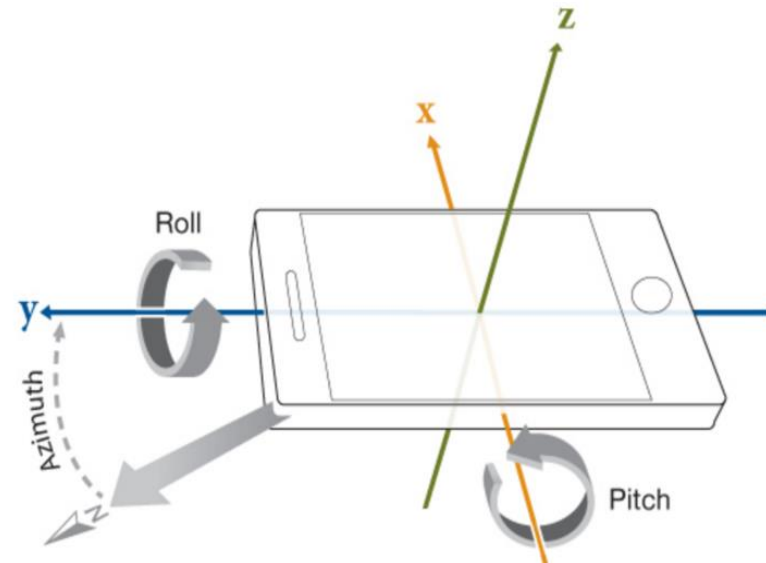
- Linear acceleration.

Gyroscope

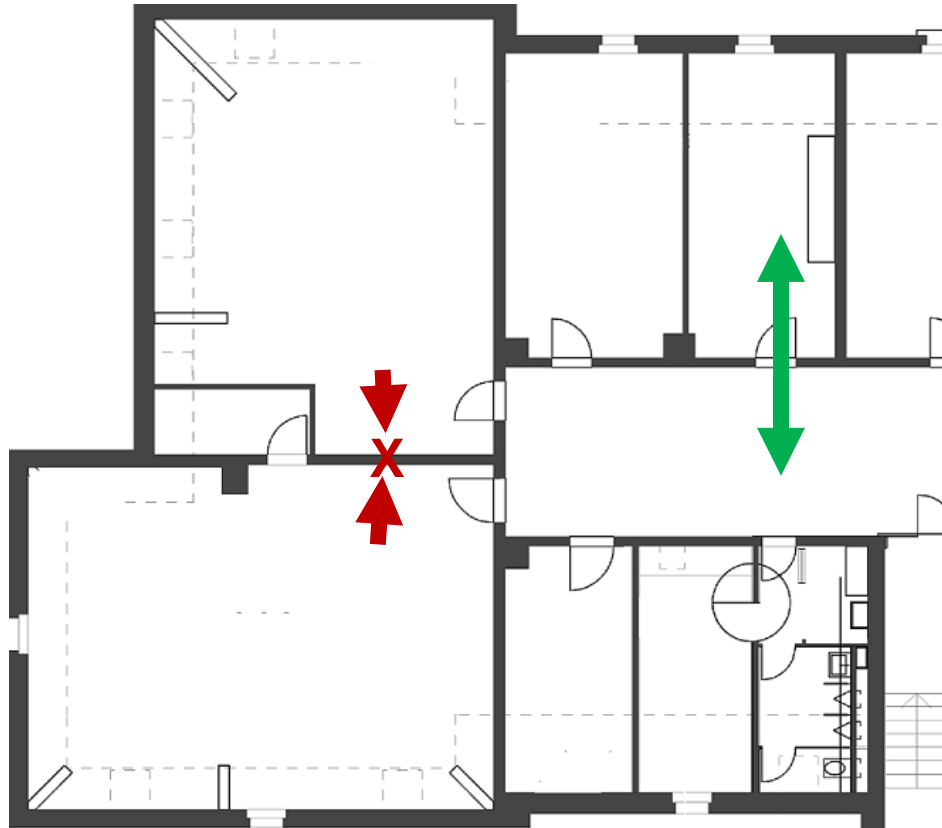
- Angular rotation velocity

Magnetometer

- Azimuth value

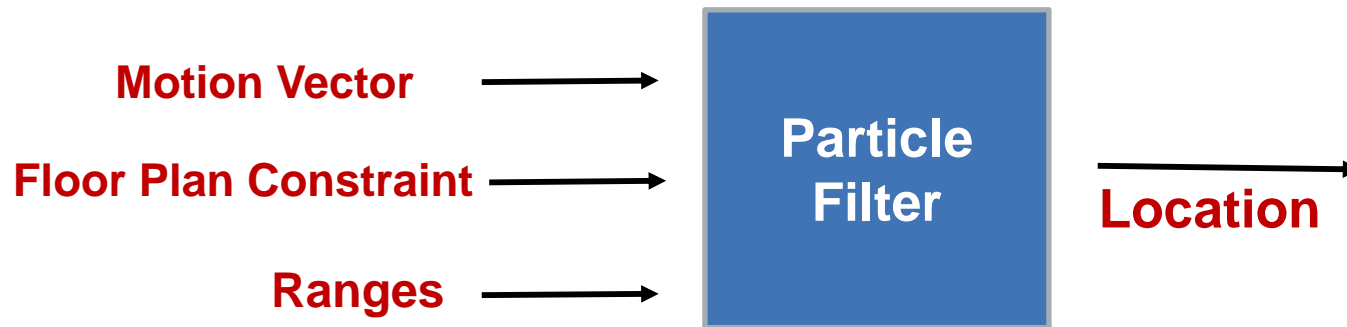


Floor Plan Component



Define “allowed” zones

Data Fusion Component

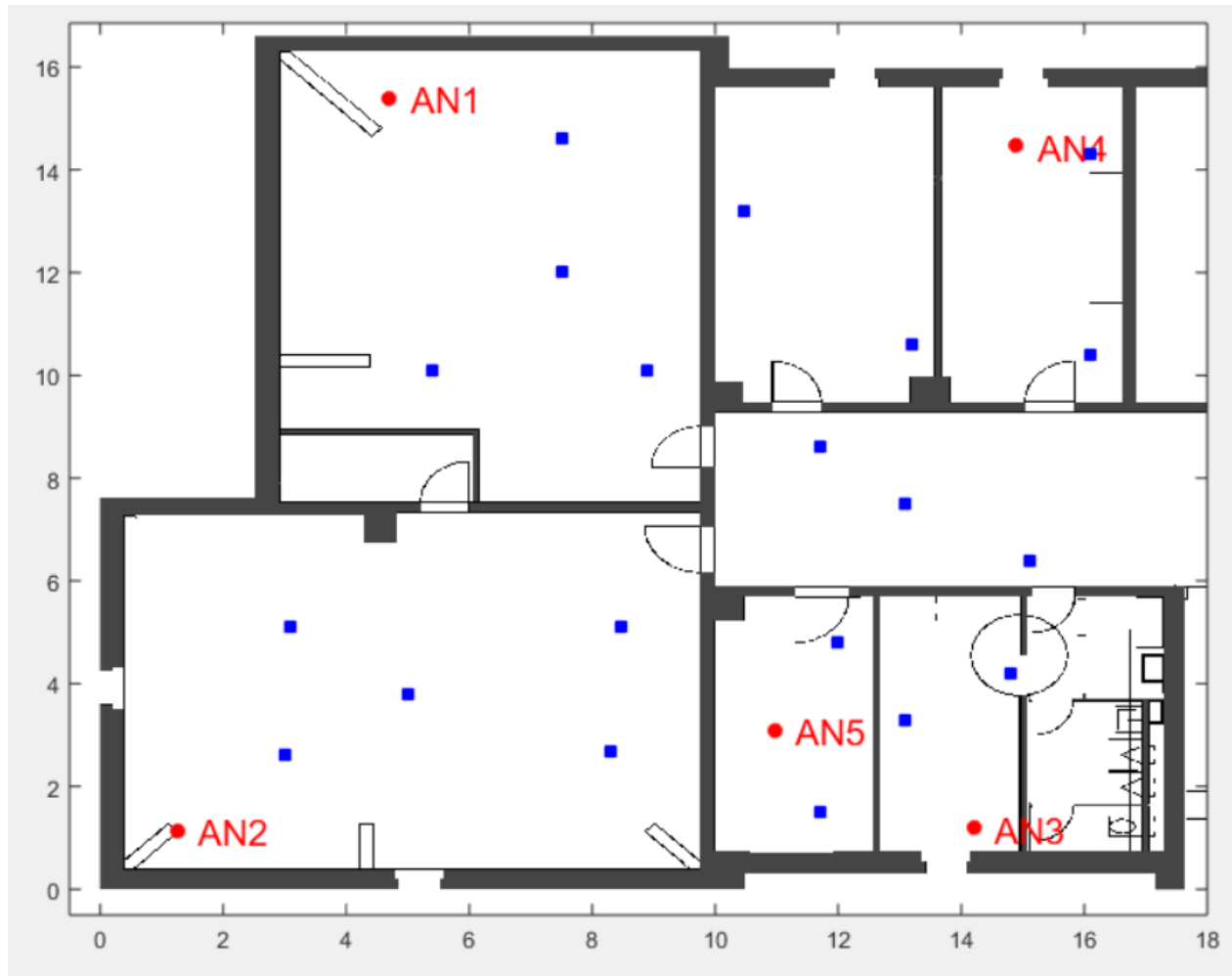


Bayesian Filter

- Model of how state changes in time.
- Model of what observations you should see.
- Represents a PDF as a set of samples (particles).
- Belief of the current state given all the observation so far.

Implementation

Ranging I

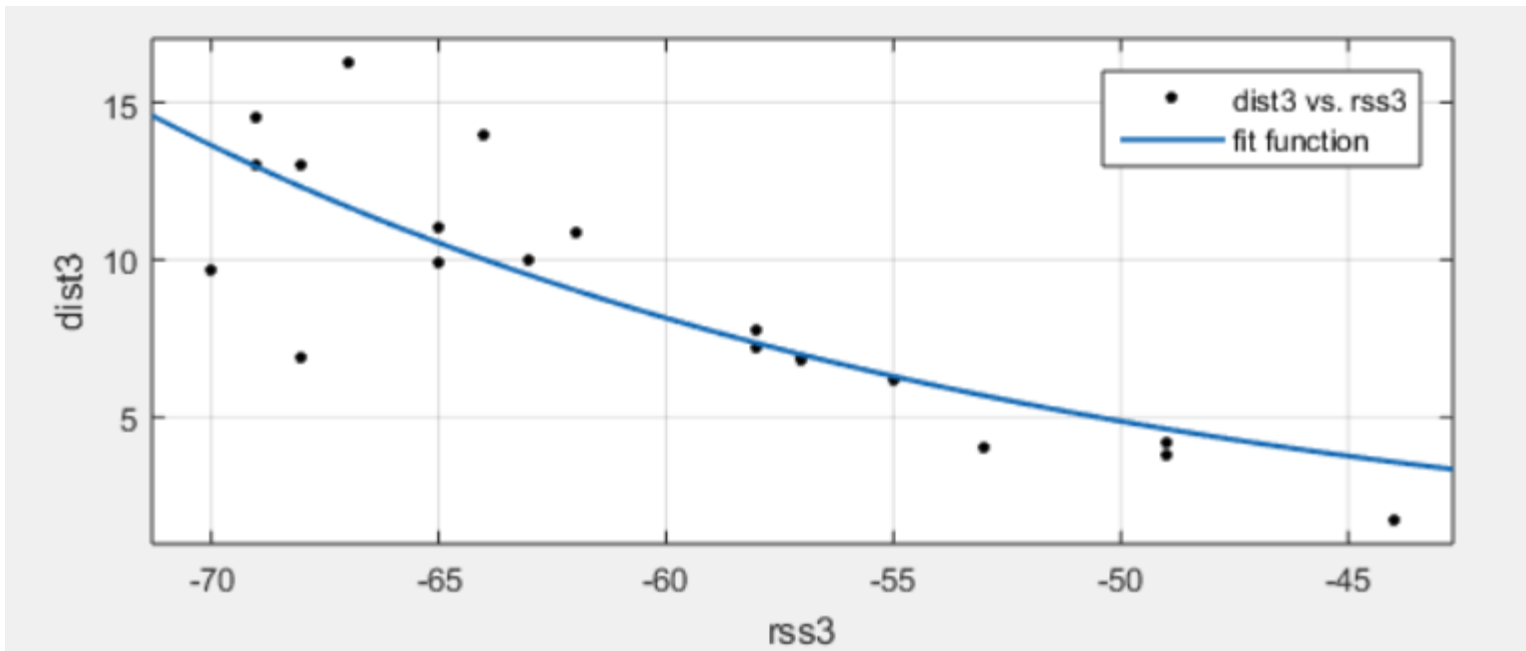


Implementation

Ranging II

Non-Linear Regression Model

$$\hat{d}_i = \alpha_i \cdot e^{\beta_i \cdot \text{RSS}_i}$$

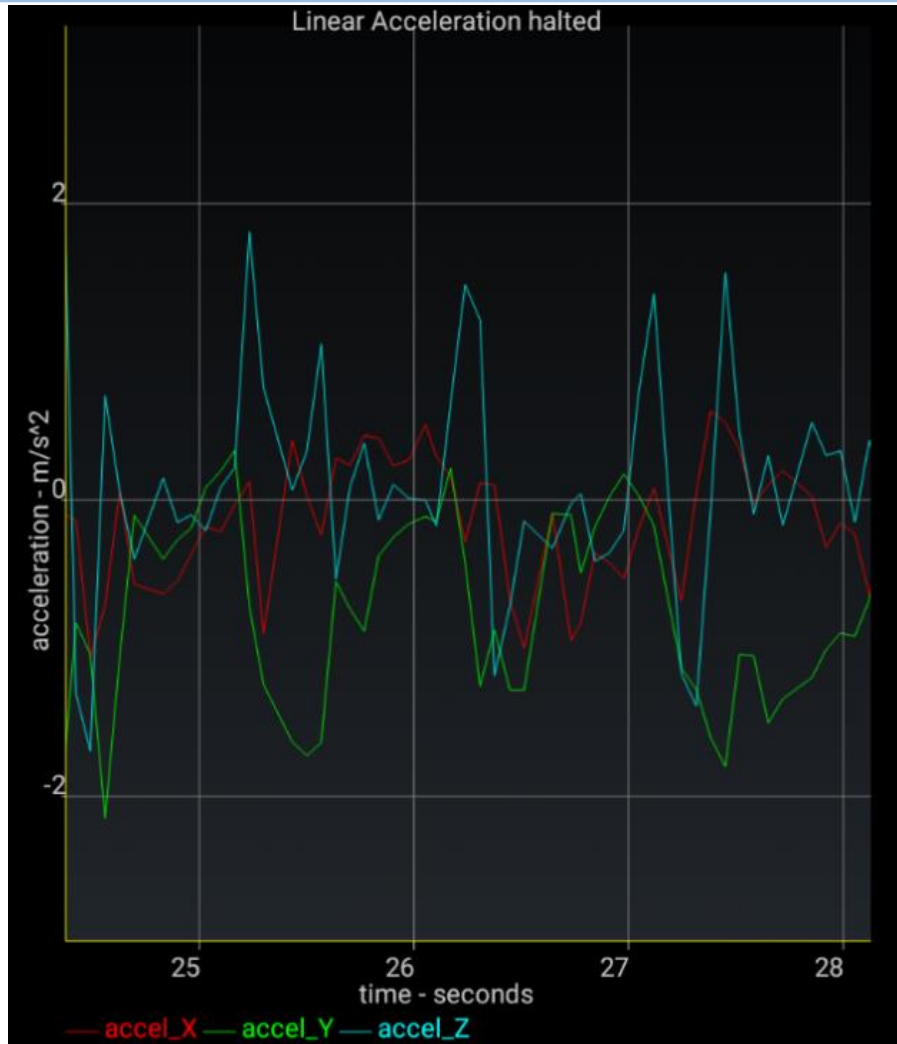


$$\alpha_3 = 0.37$$

$$\beta_3 = -0.05$$

Implementation

Inertial Measurement Unit I



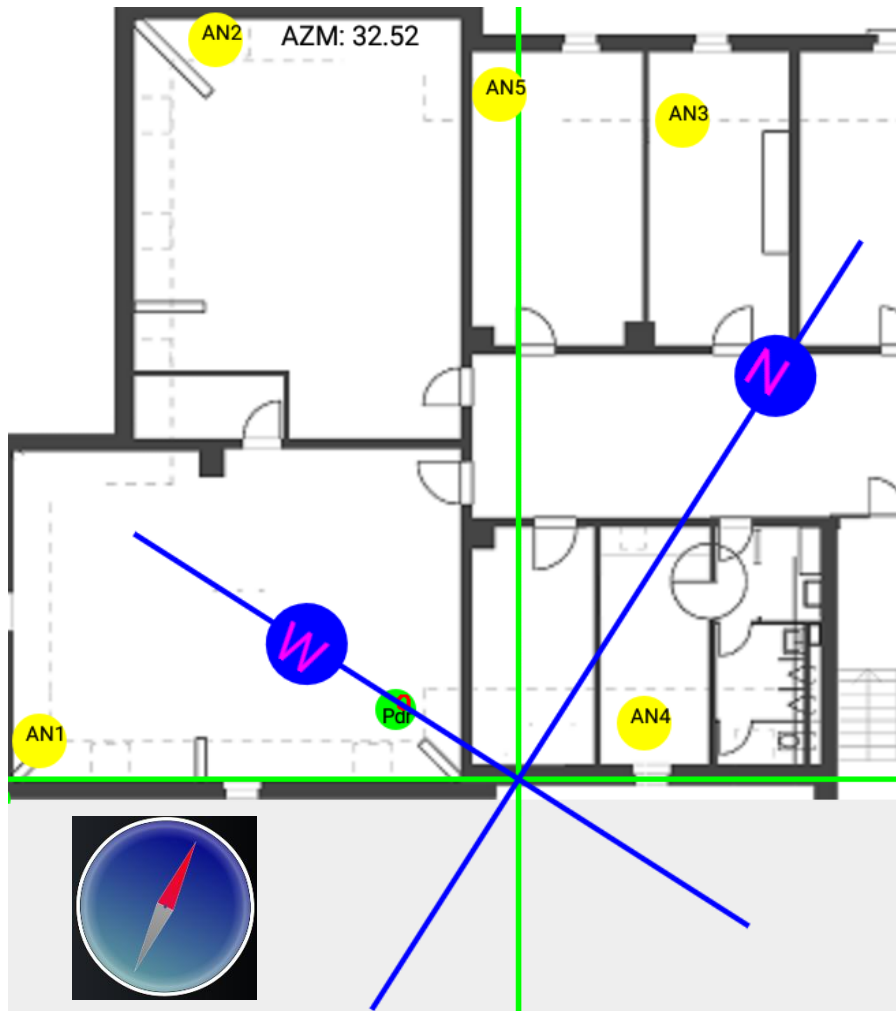
Accelerometer

Step Recognition

$$\begin{aligned} \hat{a}_{z,t} &> \text{Threshold} \ \&\& \\ \hat{a}_{z,t-1} &< \hat{a}_{z,t} \ \&\& \\ \hat{a}_{x,t} &< \hat{a}_{z,t} \ \&\& \\ \hat{a}_{y,t} &< \hat{a}_{z,t} \end{aligned}$$

Implementation

Inertial Measurement Unit II



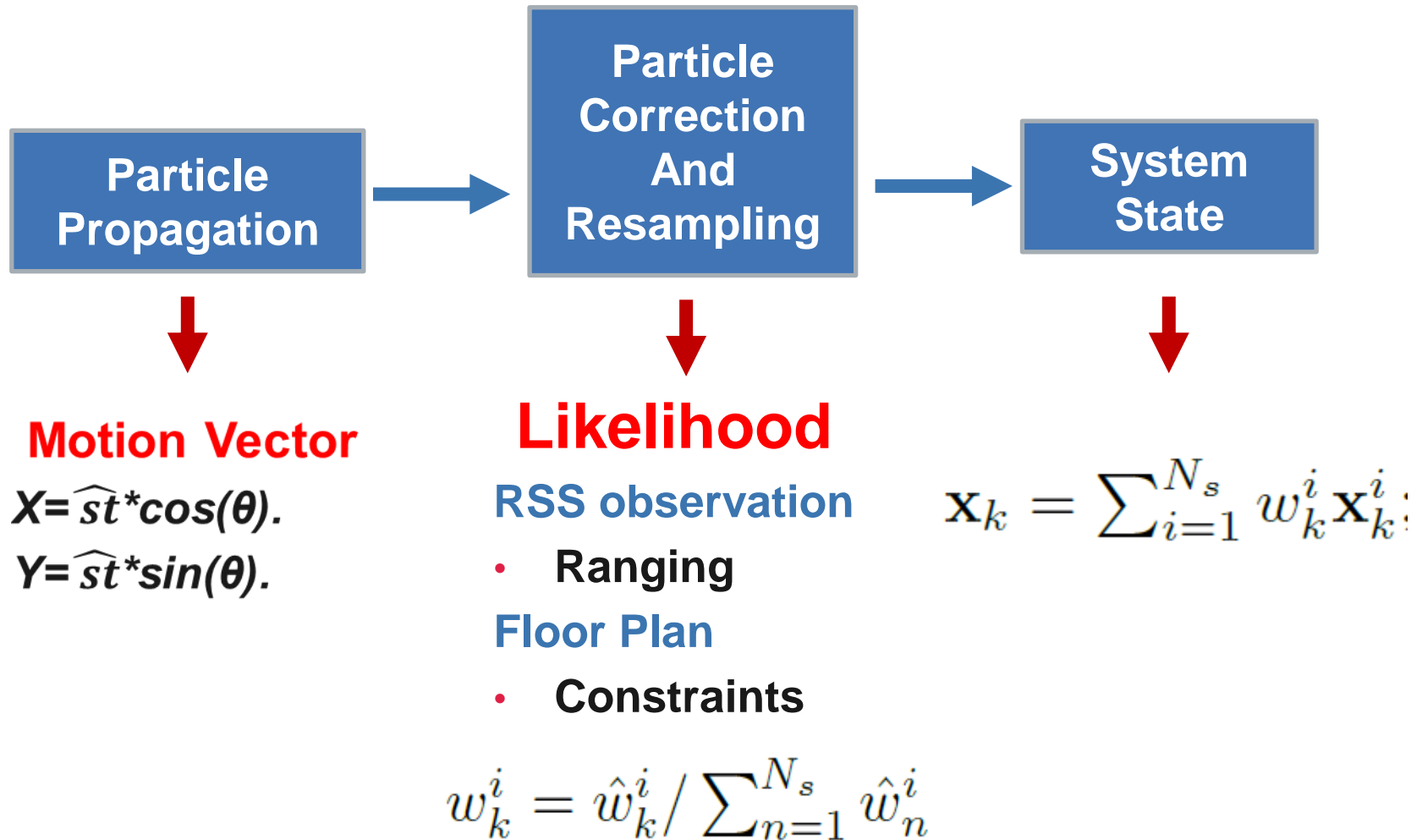
**Magnetometer,
Accelerometer,
Gyroscope**

Heading Orientation

OffsetX: Inclination X axis Magnetic North
Azimuth: Magnetic North and Y axis

$\theta = (\text{OffsetX} - \text{Azimuth}).$
 $st = \text{stride length}.$

Implementation Particle Filter

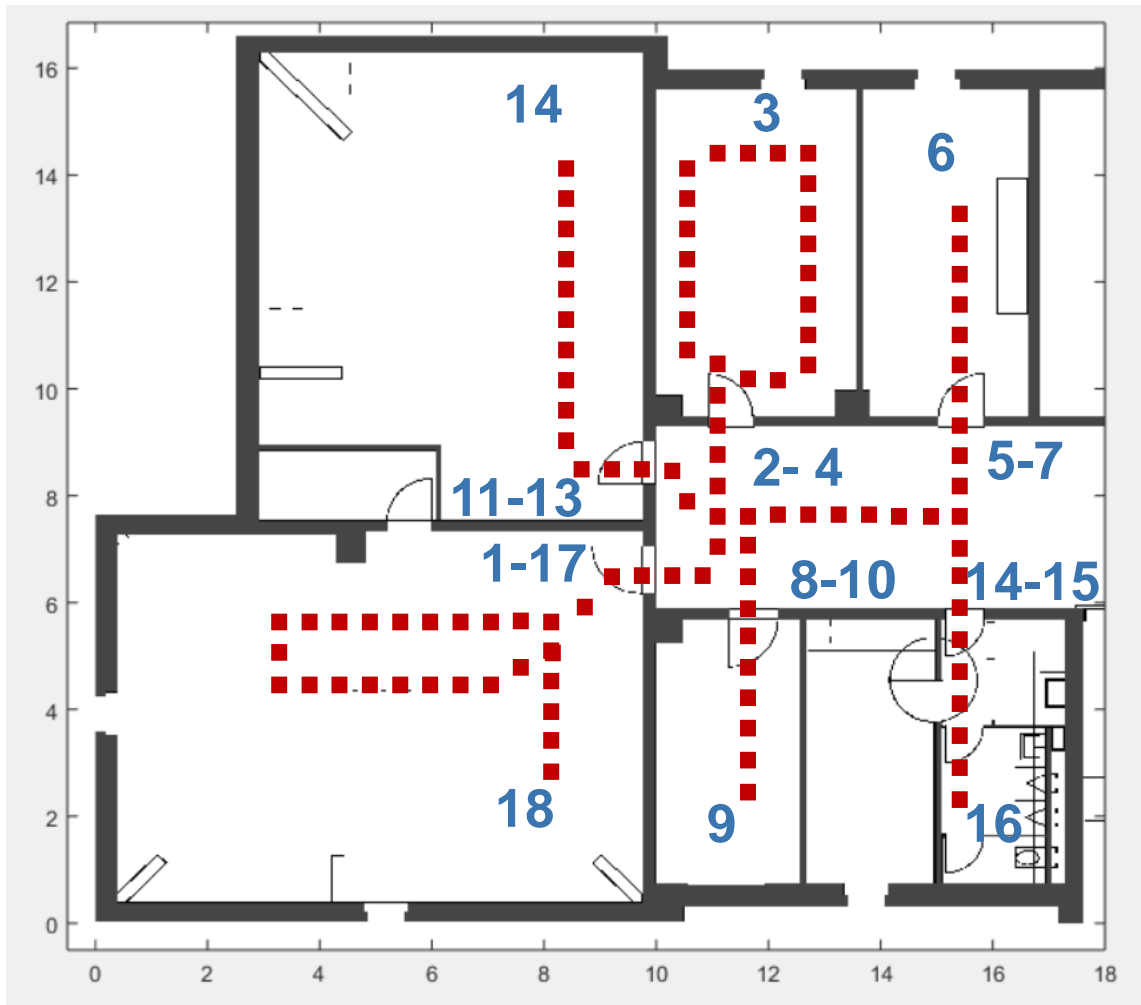


Implementation

Technical Challenges

- **Resources: Android solution.**
- **Sampling Rate.**
 - IMU 14Hz
 - WiFi sensor 3Hz.
- **Delay Uploading Position Information.**
 - Stop scanning process.
 - Cellular network (Future work).

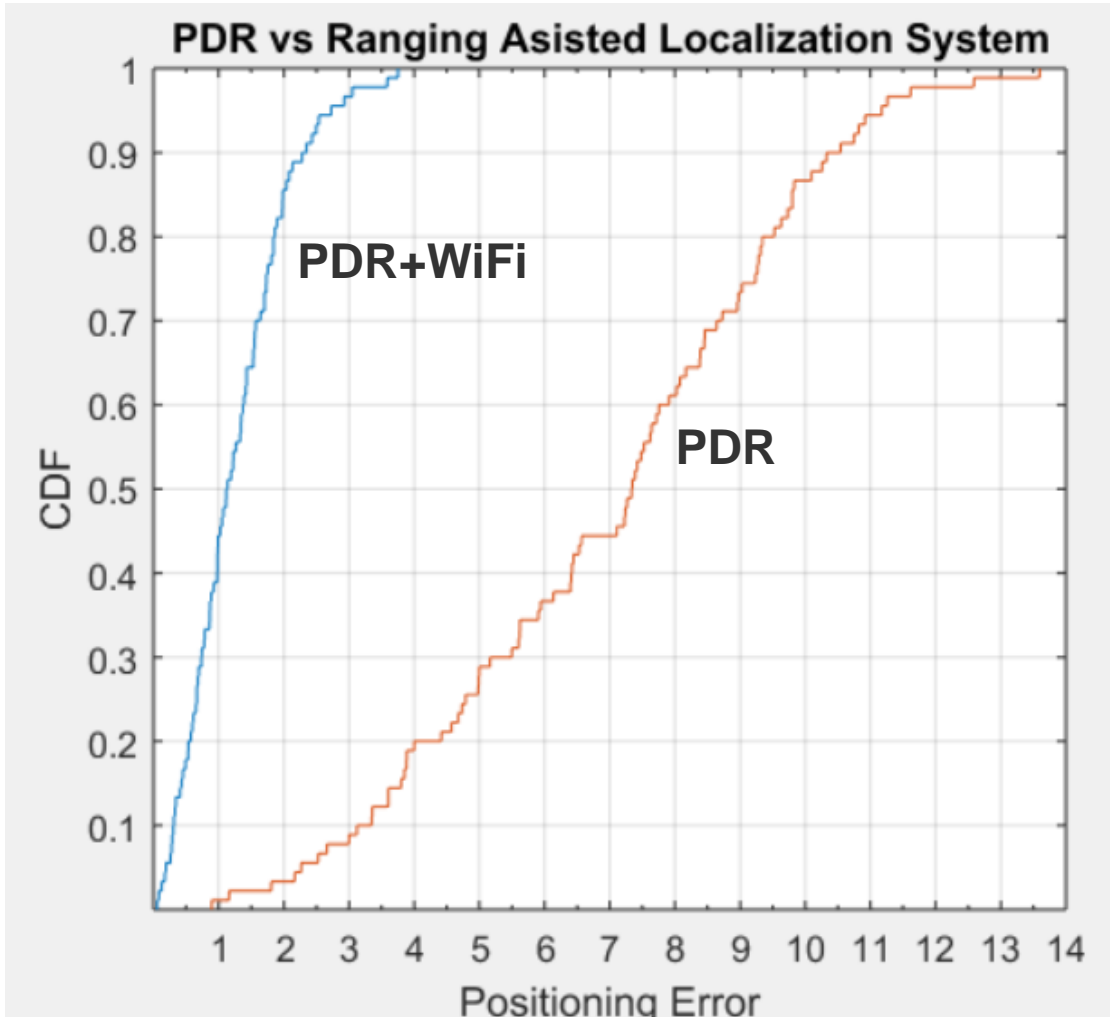
Experiment



EXPERIMENT

- 5 Trajectories
- 18 Check Points each
- 90 Check Points

Preliminary Results



Results

- 90 points.
- Mean Error:** 1.25m.
- Std:** 0.79 m.

Conclusions

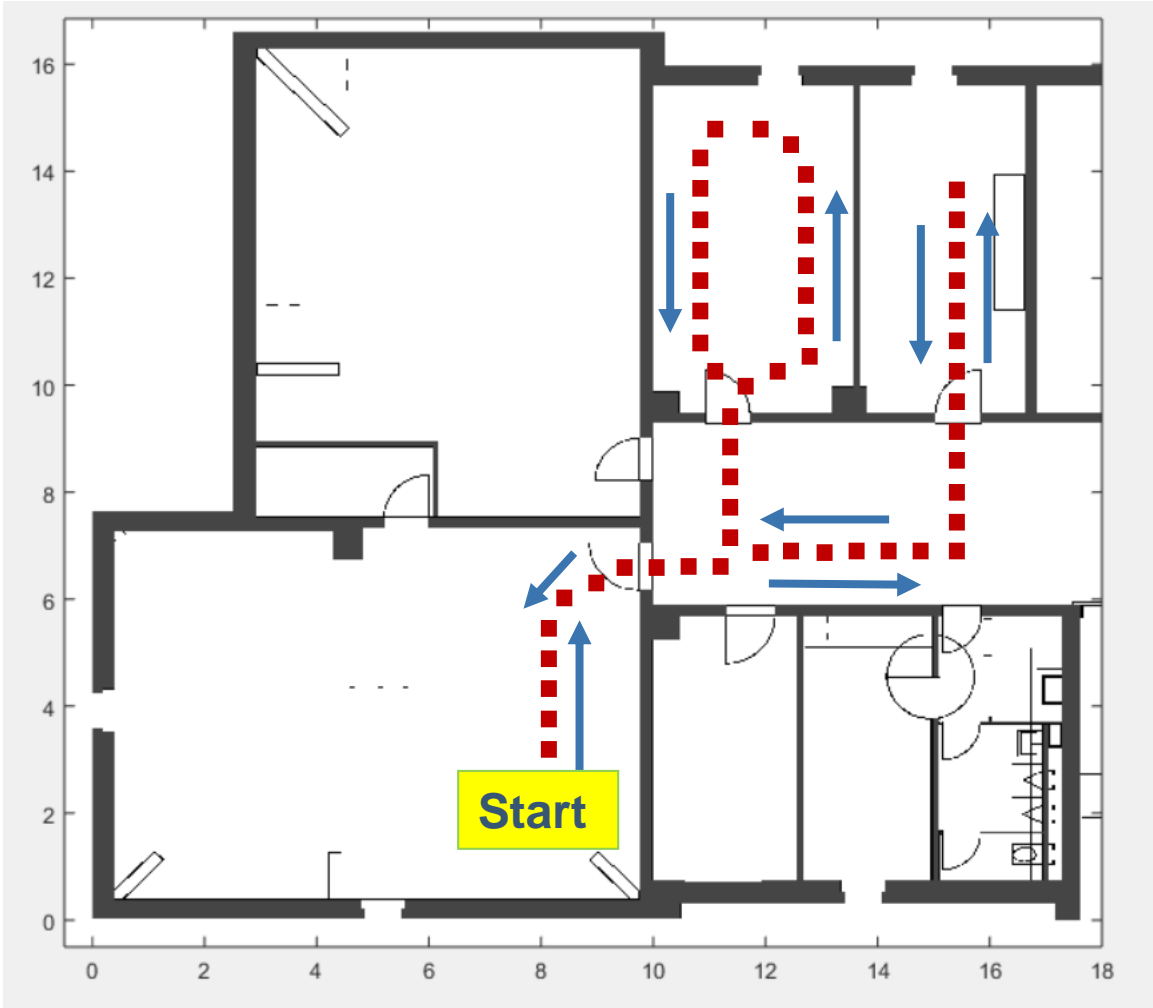
- > Tested complex scenario. Room entrance prone to error.
- > Proposed Ranging-PF assisted approach higher accuracy, more stable than PDR.
 - > 50% accuracy achieve around 1m. (PDR: 7.5m.)
 - > Outperforms PDR by 86%.
 - > 90% accuracy achieve around 2m. (PDR: 11m.)
 - > Outperforms PDR by 81%.
- > Use RSSI information to recalibrate PDR system and deal with accumulative errors.

Future Work

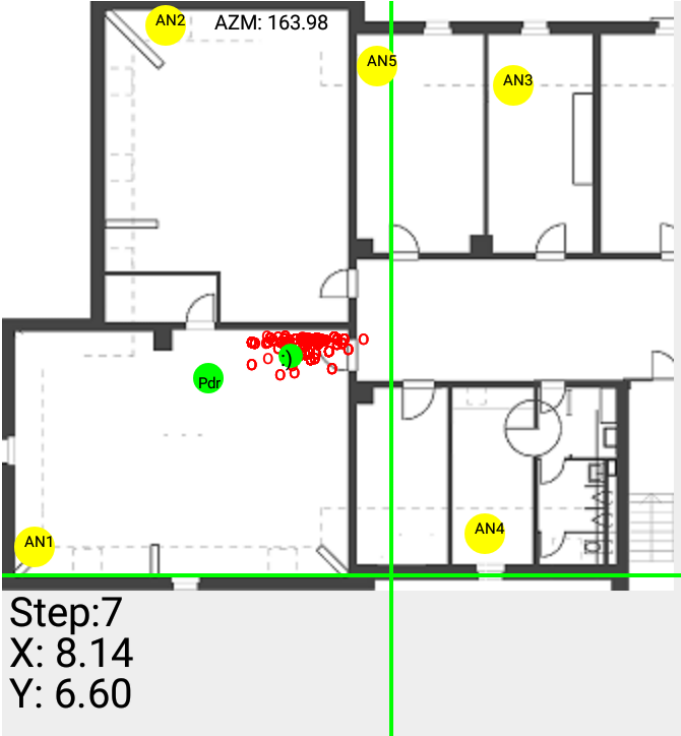
- > Design more experiments.
- > Publish results.
- > Try different technologies to upload information to the server.
- > Implement solution in the server side.
 - > Share computation server-phone. (more particles)
- > Test multiple user performance.
 - > Using cloud.
- > Include room recognition.

DEMO

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Android Application



Questions