

A Real-time Indoor Tracking System by Fusing Inertial Sensor, Radio Signal and Floor Plan.

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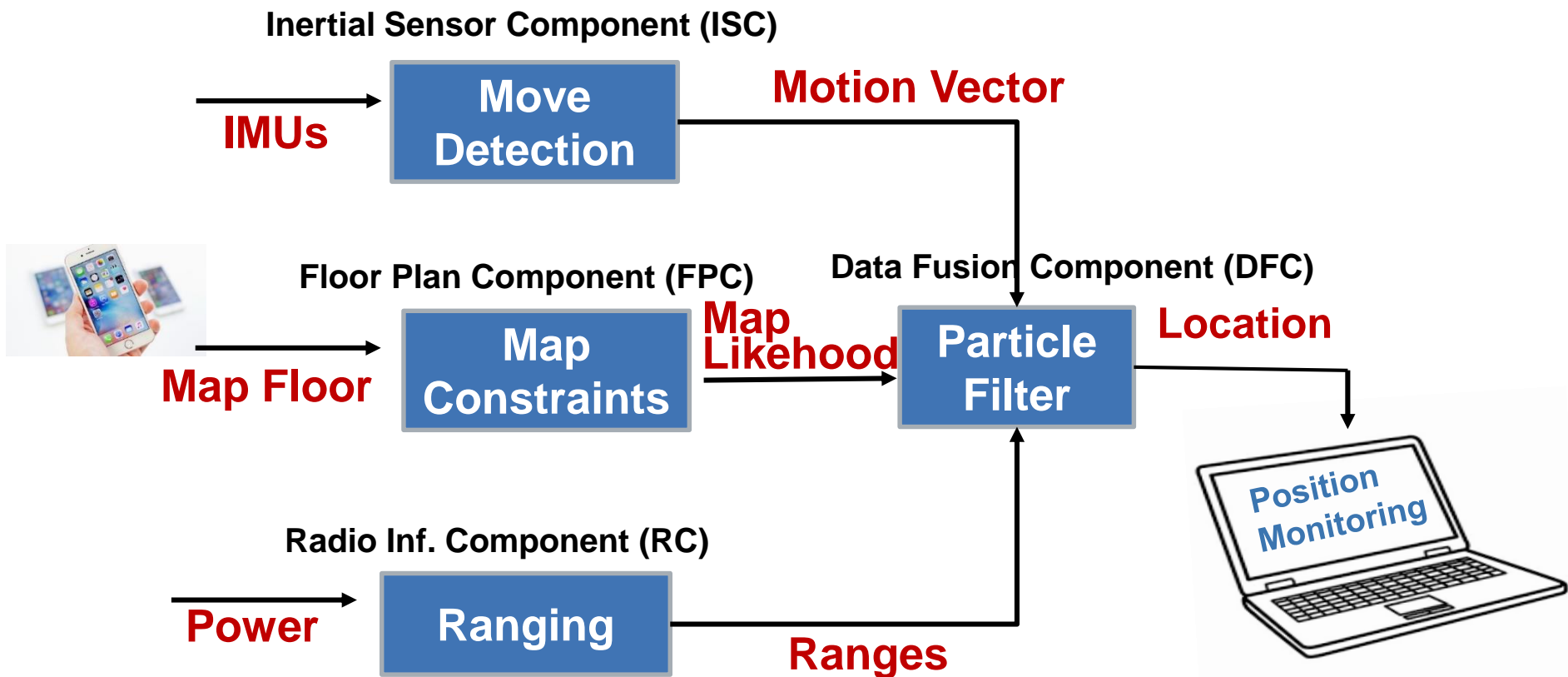
University of Bern

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Outline

- > Proposed Indoor Positioning System
 - > Inertial Sensor Component.
 - > Radio Information Component.
 - > Floor Plan Information Component.
 - > Data Fusion Component.
- > Implementation
 - > Inertial Measurement Unit (IMU) process.
 - > Ranging process.
 - > Particle Filter.
- > Experiments
- > Preliminary Results
- > Conclusions

Proposed Indoor Positioning System



Inertial Sensor Component

Accelerometer:

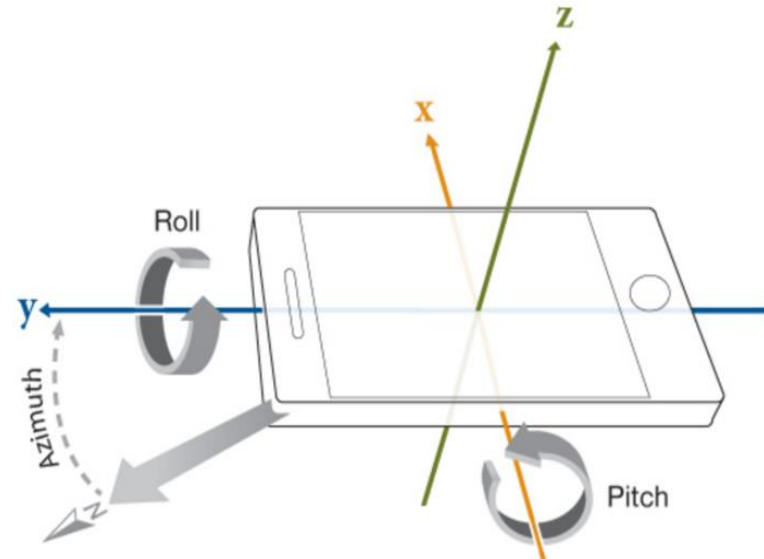
- Linear acceleration.

Gyroscope

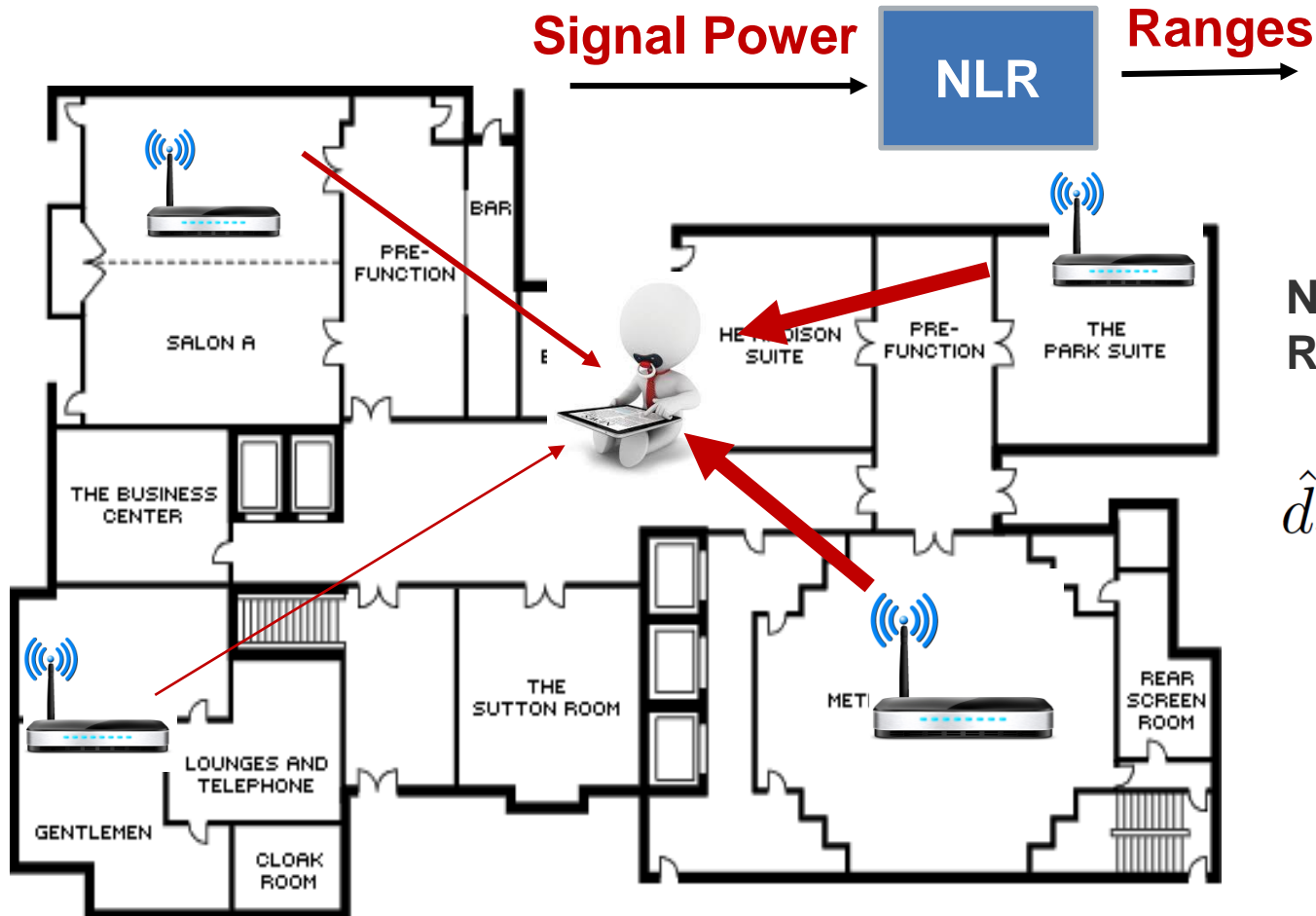
- Angular rotation velocity

Magnetometer

- Azimuth value



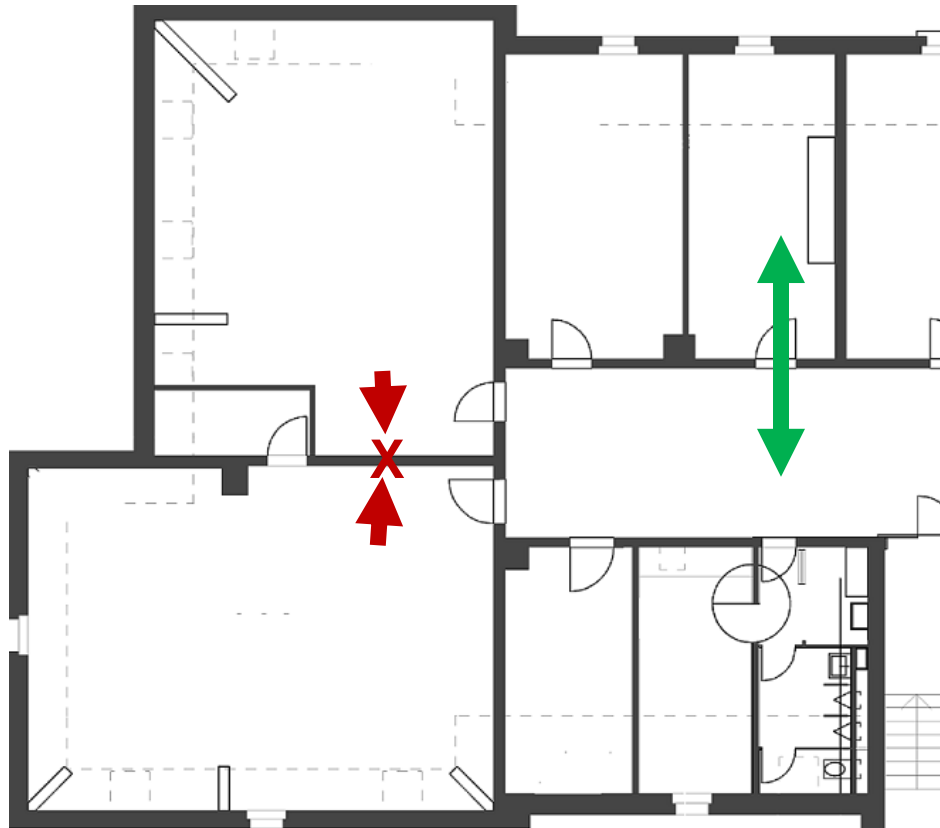
Radio Information Component



Non-Linear
Regression Model [1]

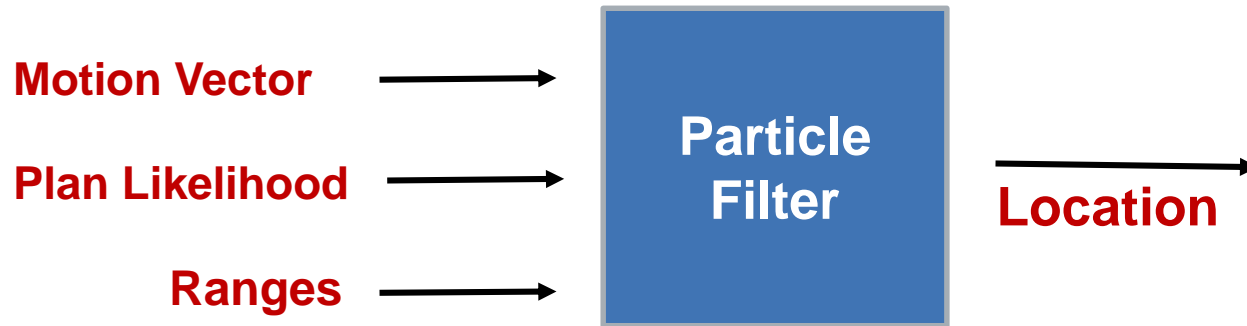
$$\hat{d}_i = \alpha_i \cdot e^{\beta_i \cdot \text{RSS}_i}$$

Floor Plan Information Component



Define “allowed” zones

Data Fusion Component

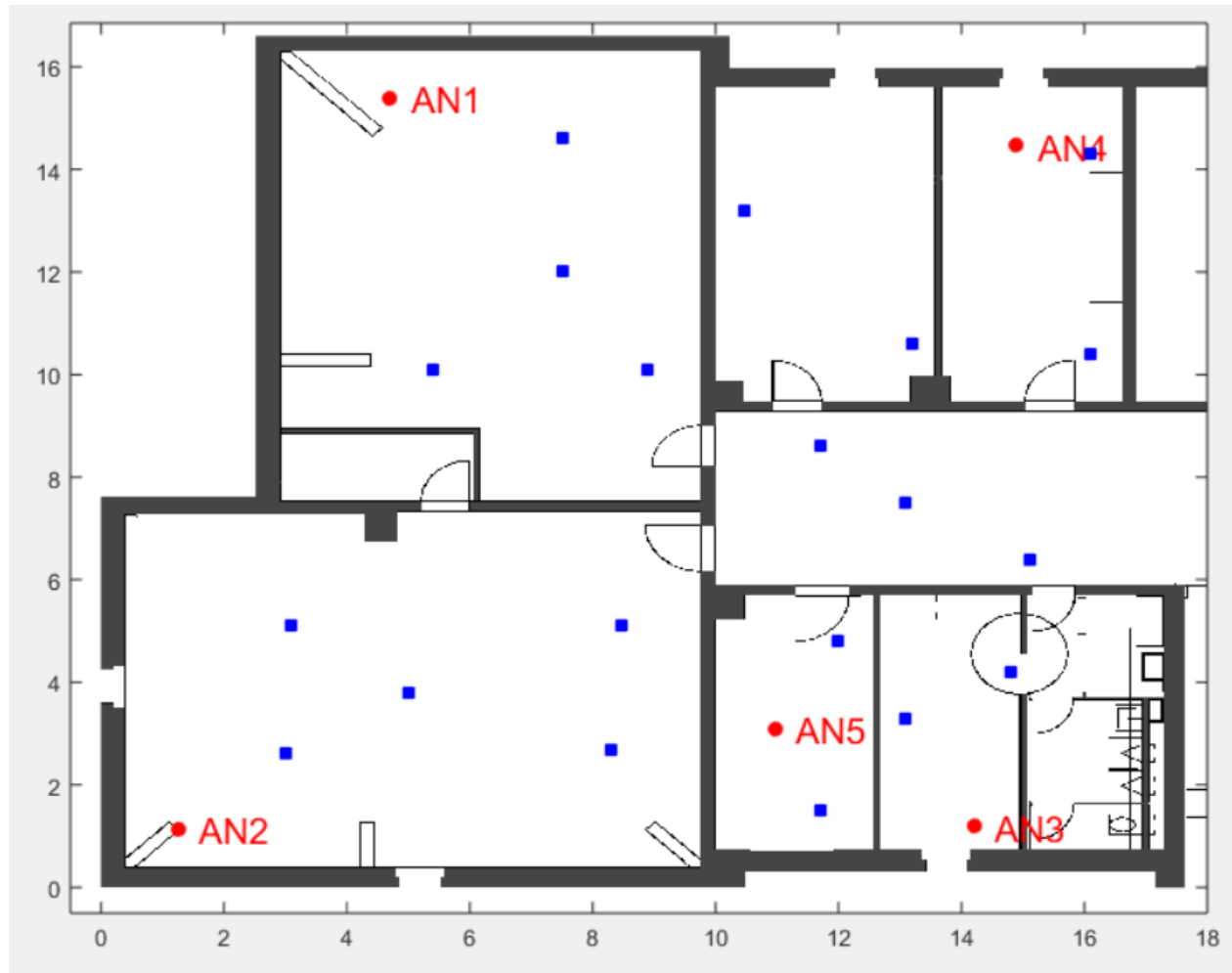


Bayesian Filter

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

Implementation

Ranging I

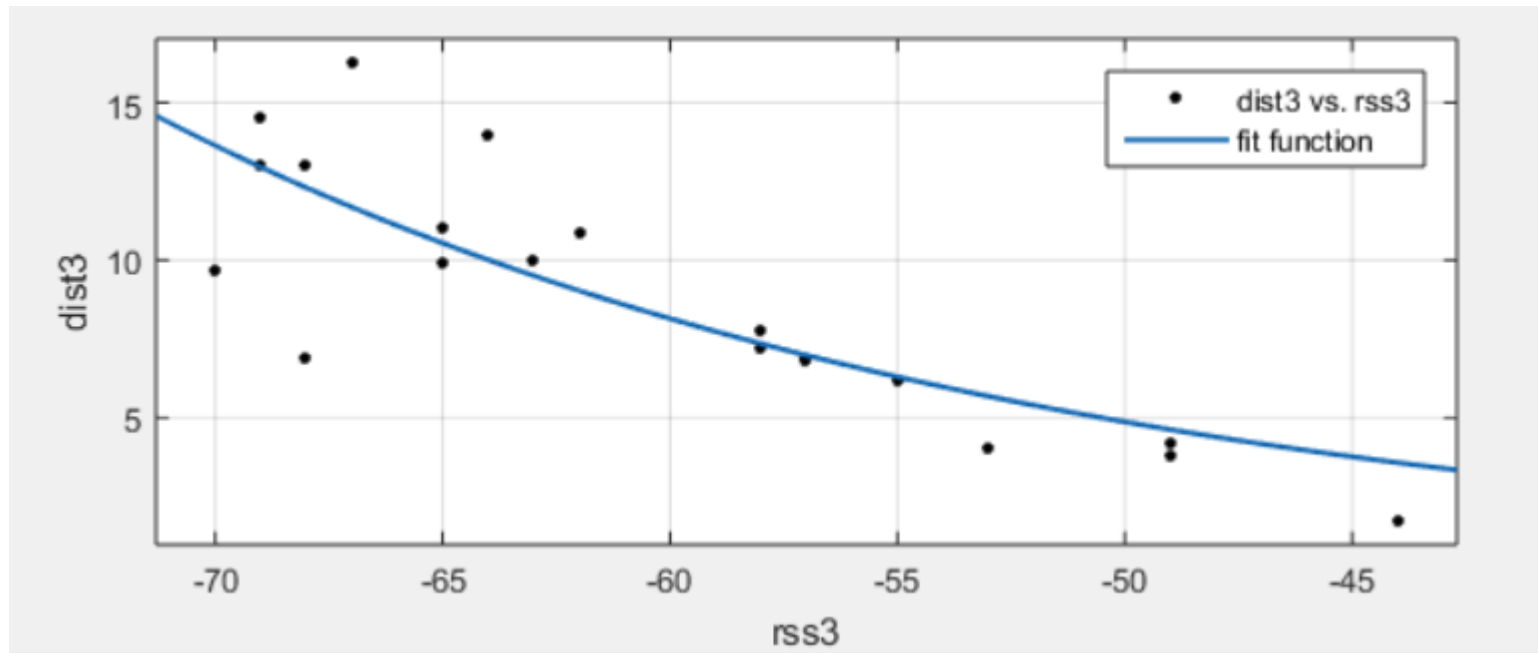


Implementation

Ranging II

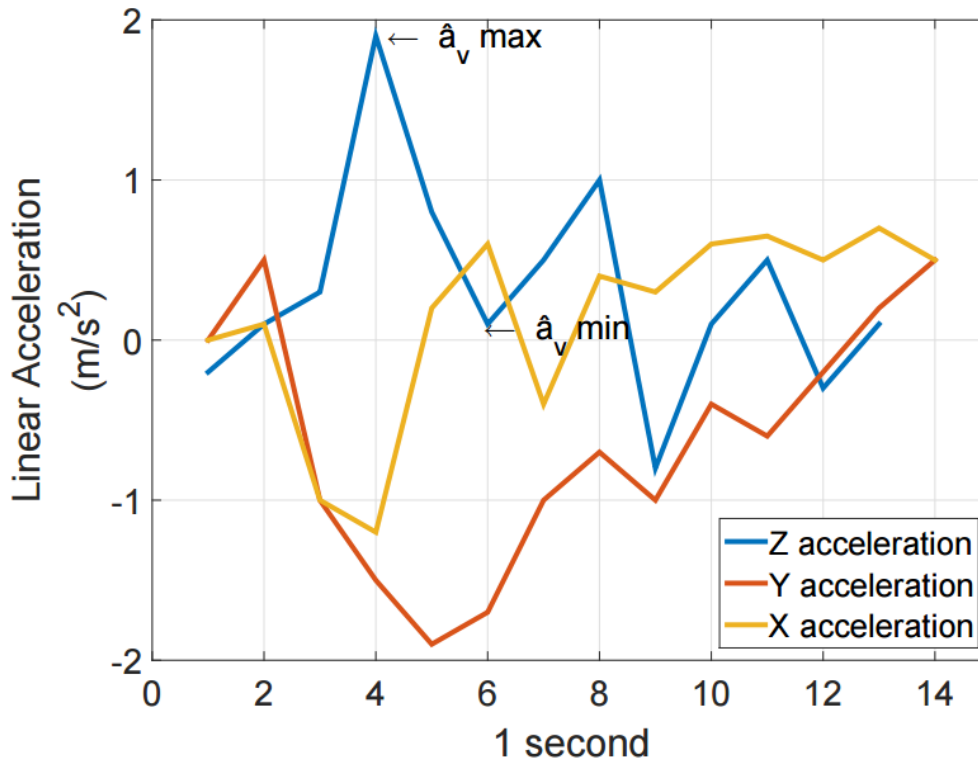
Non-Linear Regression Model

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



Implementation

Inertial Measurement Unit I



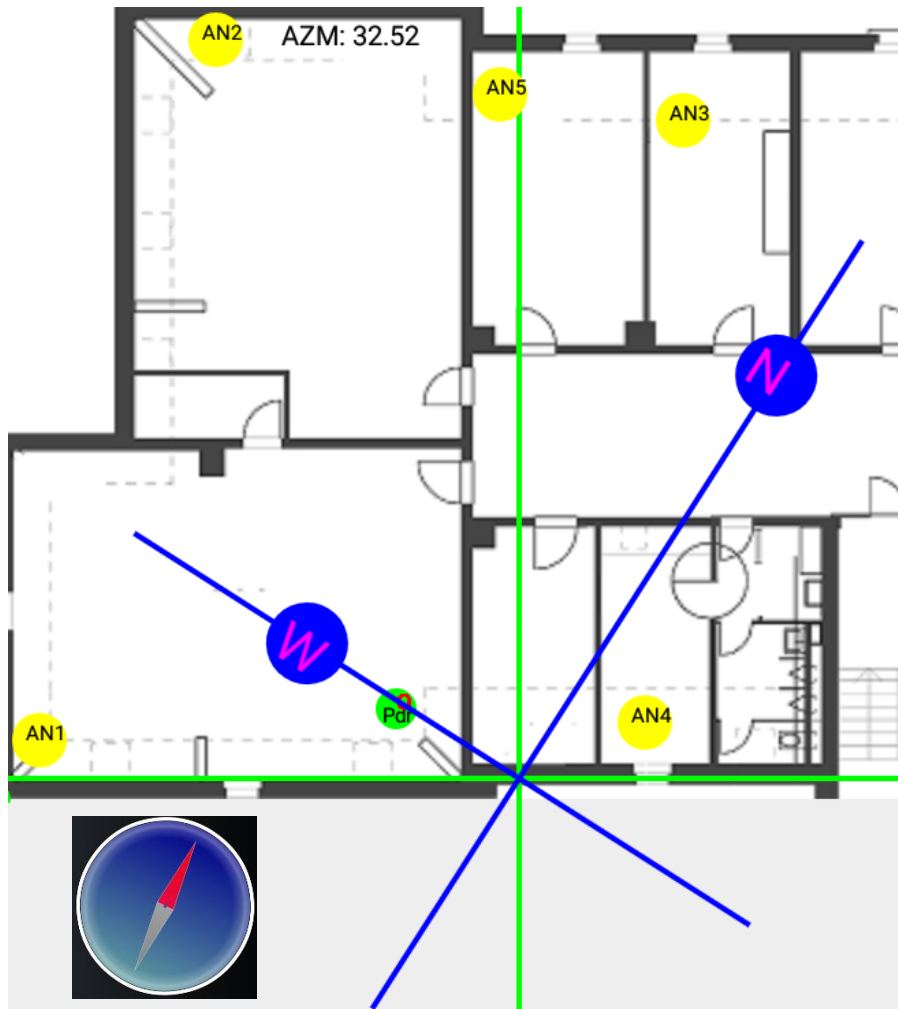
Accelerometer

Step Recognition

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

Implementation

Inertial Measurement Unit II



Magnetometer, Accelerometer

Heading Orientation

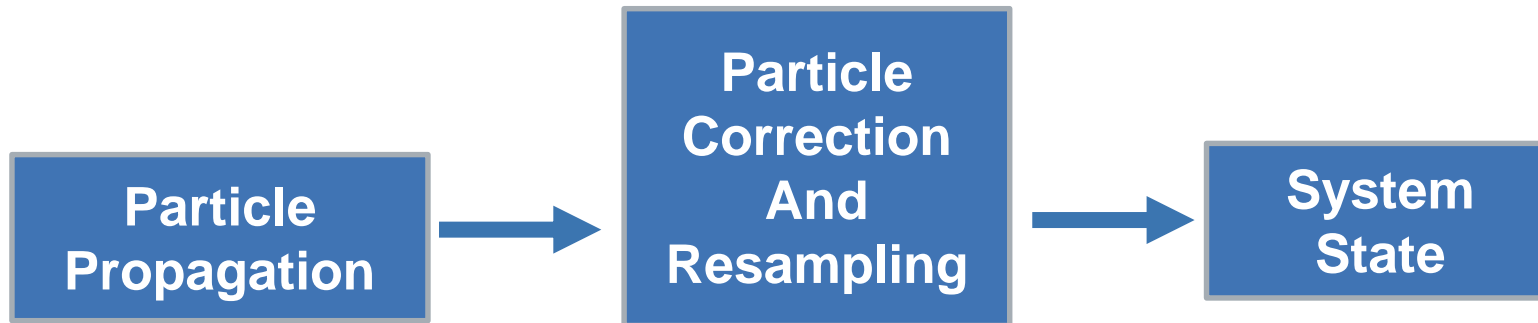
OffsetX: Inclination X axis Magnetic North

Azimuth: Magnetic North and Y axis

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

$st = \text{stride length}.$

Implementation Particle Filter



Motion Vector

$$X = \hat{st} * \cos(\theta).$$

$$Y = \hat{st} * \sin(\theta).$$



Likelihood

RSS observation

- Ranging

Floor Plan

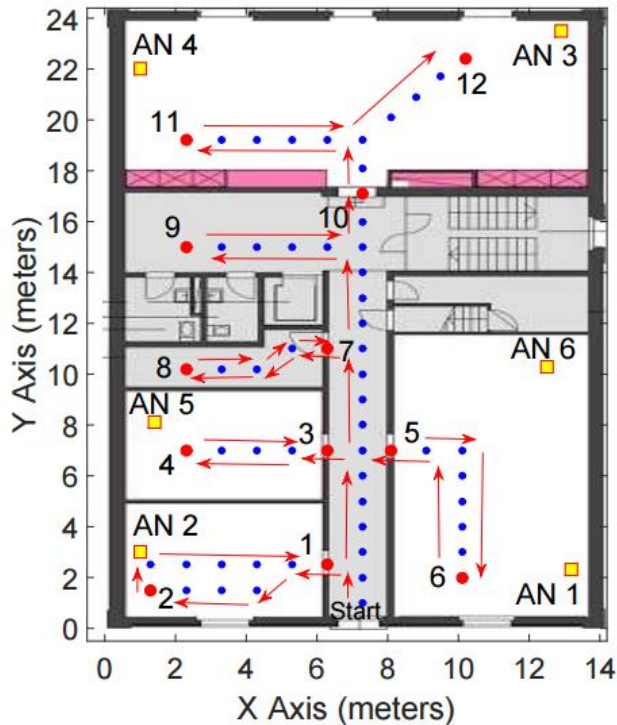
- Constraints



$$\mathbf{x}_k = \sum_{i=1}^{N_s} w_k^i \mathbf{x}_k^i$$

$$w_k^i = \hat{w}_k^i / \sum_{n=1}^{N_s} \hat{w}_n^i$$

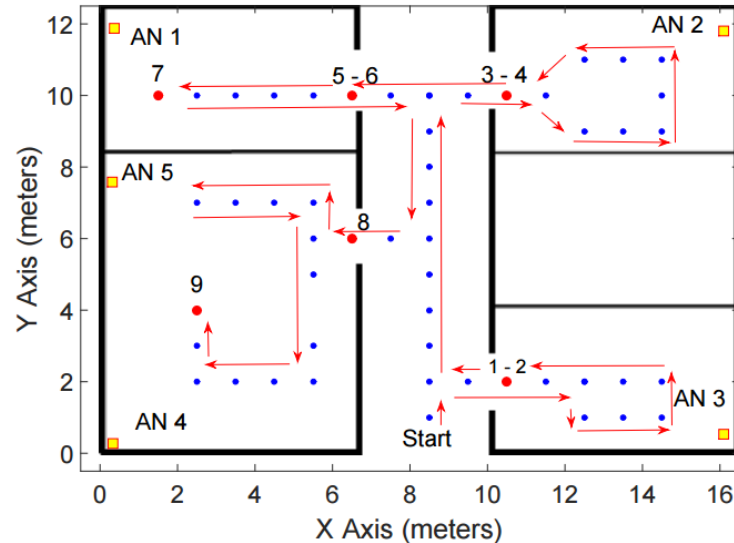
Experiments



(a) Scenario 1, trajectory 1

EXPERIMENT 1

- University of Bern.
- Target area: 336 m² (3 floors)
- 12 Check Points

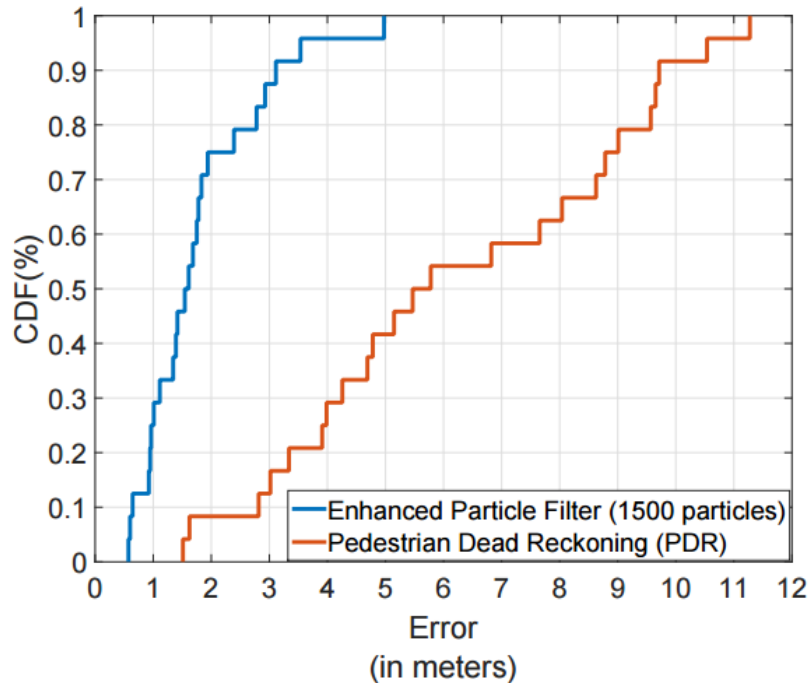


(b) Scenario 2, trajectory 2

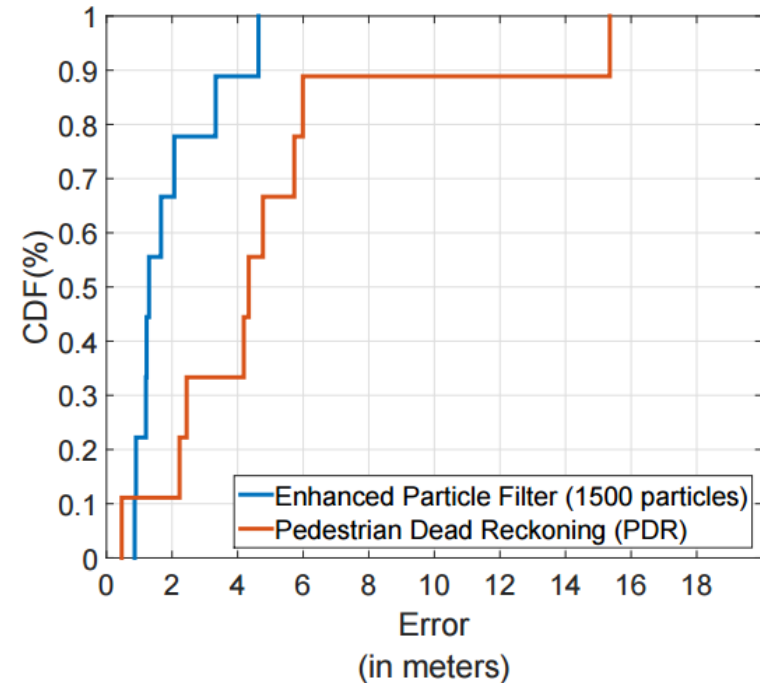
EXPERIMENT 2

- University of Geneva.
- Target area: 192 m²
- 9 Check Points

Results



(a) Particle Filter vs PDR, Scenario 1



(b) Particle Filter vs PDR, Scenario 2

Tracking Approach	Mean error	S.D	90% Acc.
Particle Filter Scenario 1	1.7m	1.0m	3.2m
Particle Filter Scenario 2	1.9m	1.27m	4.3m
PDR Scenario 1	6.2m	2.9m	10.5m
PDR Scenario 2	5.1m	4.25m	15.7m

Conclusions

- > Tested complex scenario. Room entrance prone to error.
- > Proposed Ranging-PF assisted approach higher accuracy, stable than PDR.
- > PF outperforms PDR around 72.6% with 90% accuracy.
- > Use RSSI-based ranging information to recalibrate PDR to deal with accumulative errors.
- > RSSI-based ranging information requires ANs position.
- > Remarks from competition
 - > Outdated AP locations/MAC information provided
 - > Large scenarios (50000 m²) take long survey period
 - > Ranging or fingerprinting?

Questions?

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