

Pazl: A Mobile Crowdsensing based Indoor WiFi Monitoring System



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informatics

Introduction

- **6 billion mobile subscriptions** in the world [source: UN report, 2013].



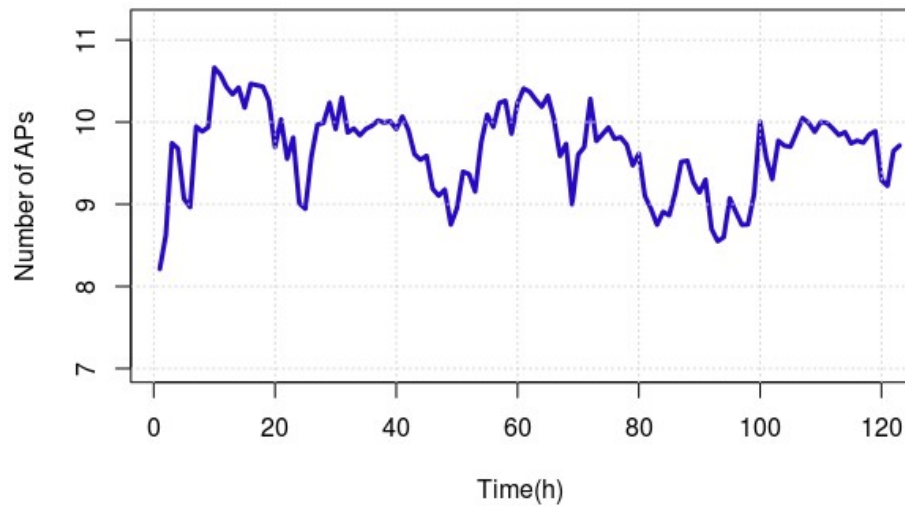
- **1.4 billion smartphones** will be in use by December 2013 [source: ABI, 2013] and expected to reach 2 billion by 2015 [source: Strategy Analytics, 2013].



Motivation

WLANs require permanent monitoring to capture all the dynamic aspects.

In the Informatics Forum:

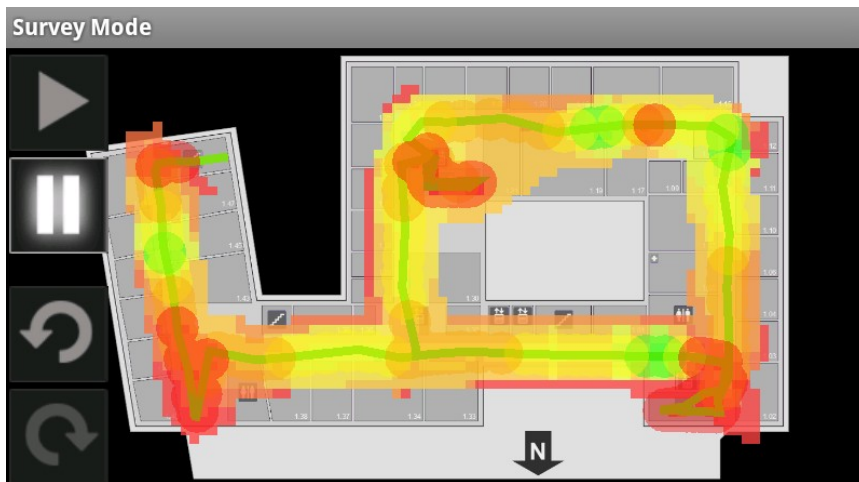


Dynamic fluctuation of APs number
at a single location

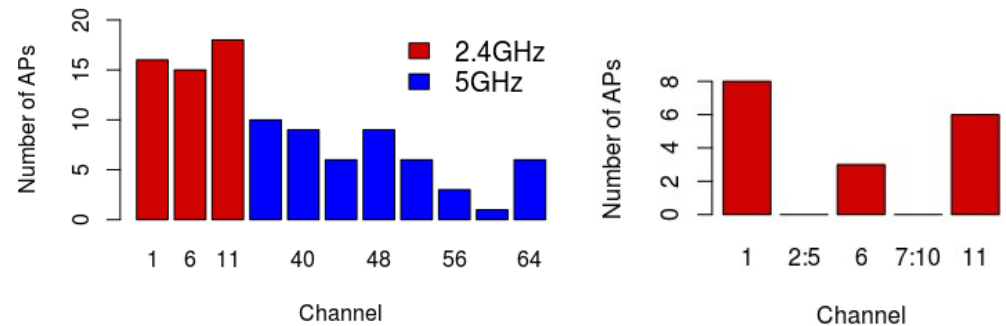


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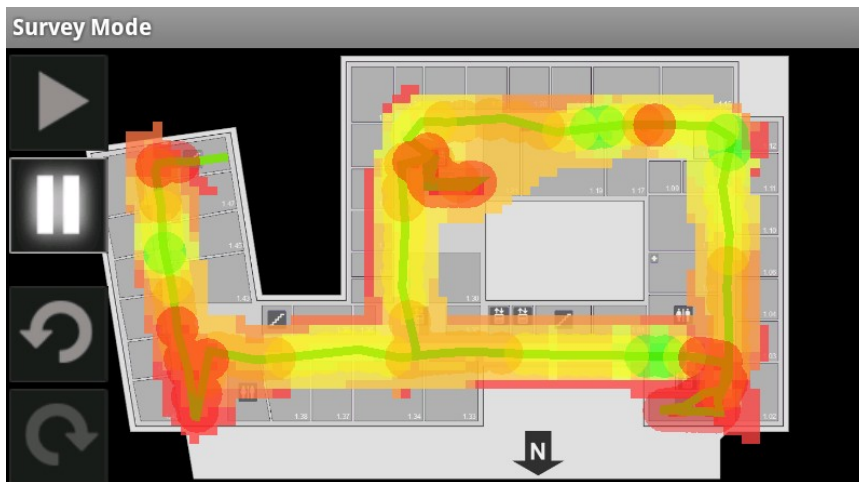
Manual site survey with Ekahau
observing some coverage holes



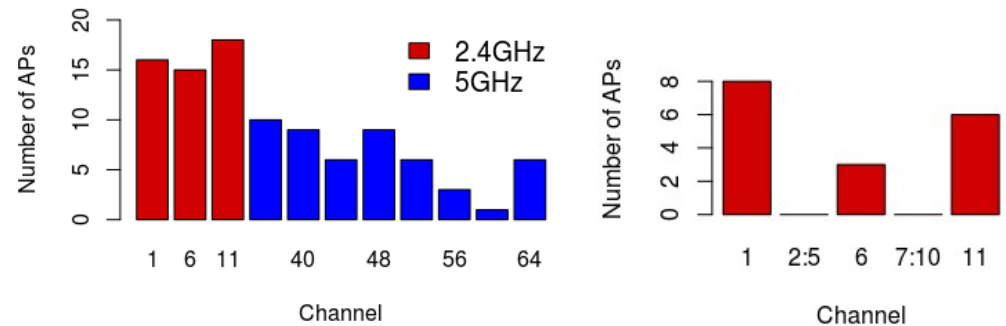
Channel imbalance

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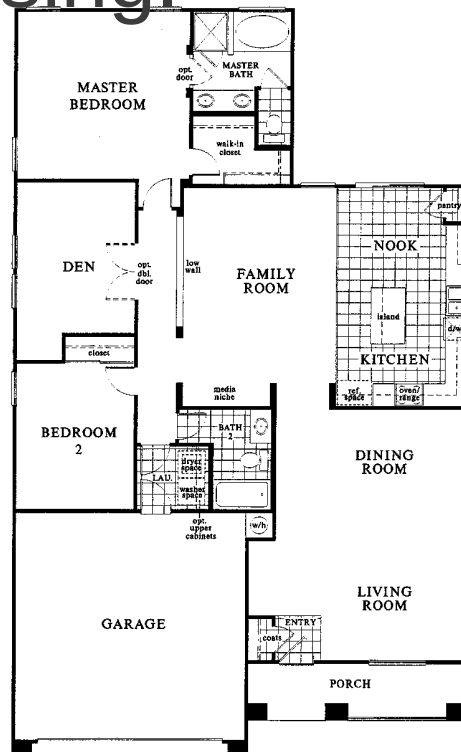
Need for continuous monitoring in space and time.

Motivation

- Traditional site surveys are expensive, intrusive and time consuming.
- People carry smartphones that can perform ubiquitous sensing.

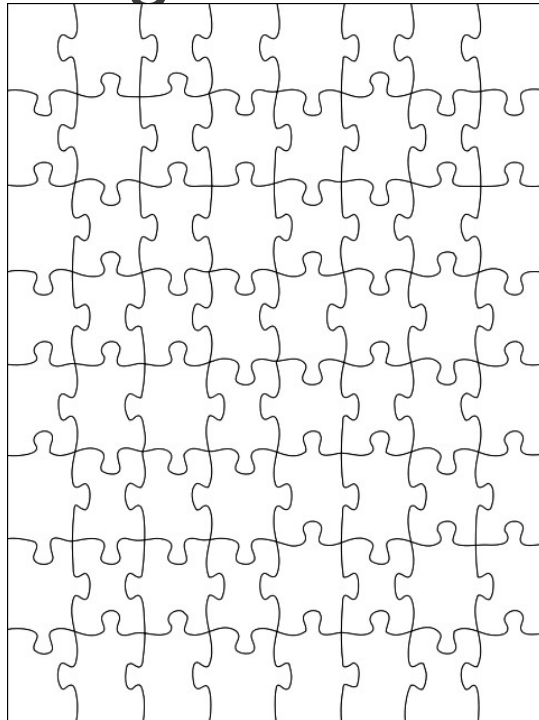
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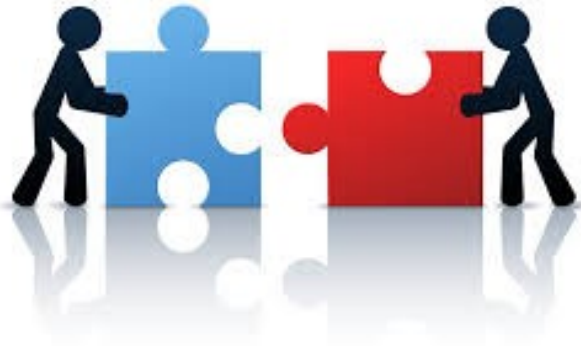
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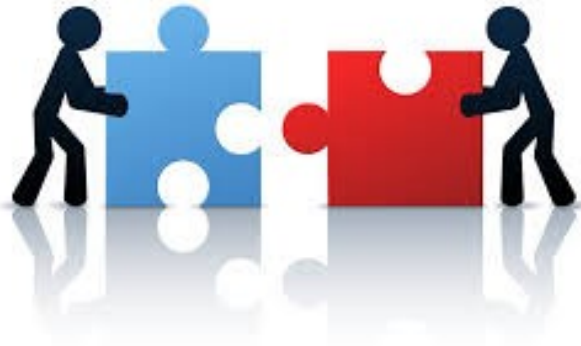
Pazl

Pazl - a mobile crowdsensing based indoor WiFi monitoring system.



Pazl

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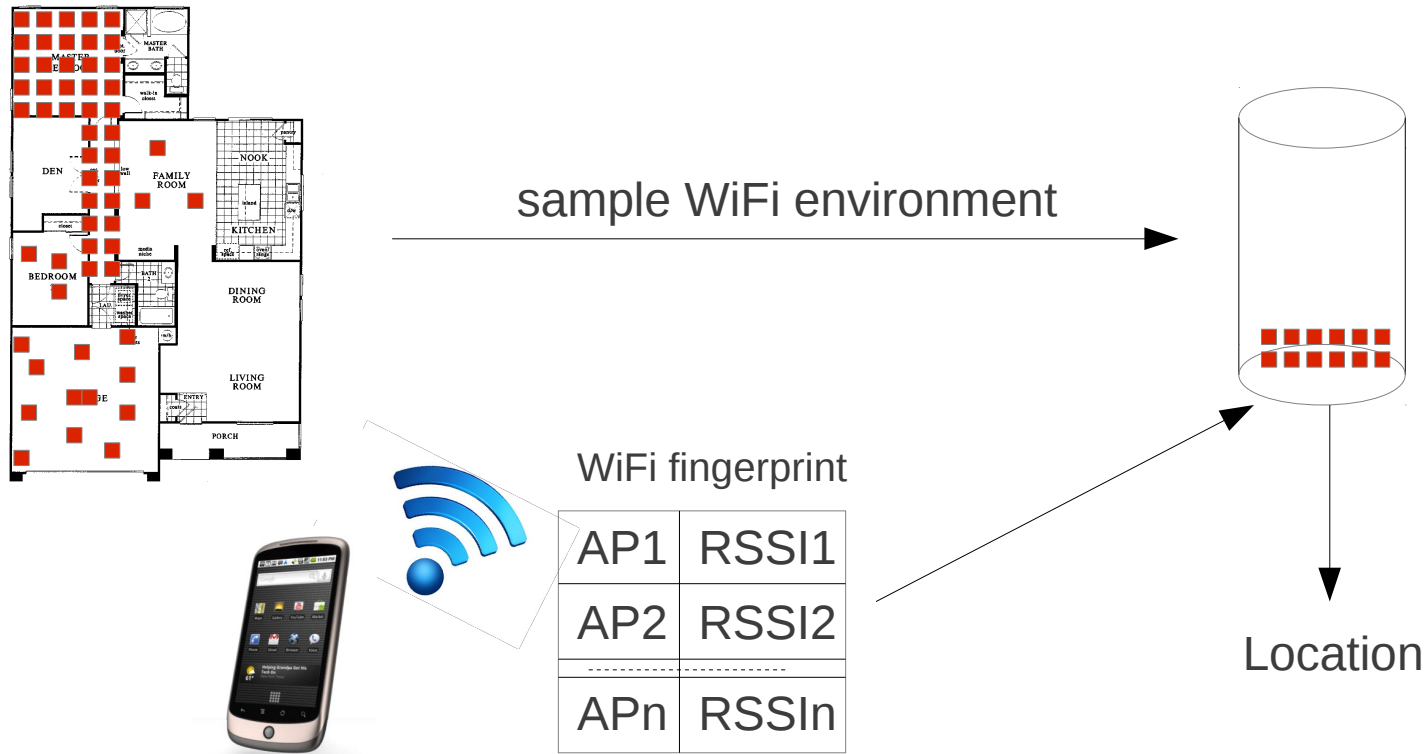
Continuous monitoring
of the WiFi environment

Challenges

- To map any data we need to annotate it with its location.
- GPS is an established localization solution for outdoors, but not very reliable inside a building.
- Indoor localization:
 - WiFi fingerprinting or
 - Pedestrian Dead Reckoning (PDR).

Background – WiFi fingerprinting

Offline phase – data collection



Online phase – localization

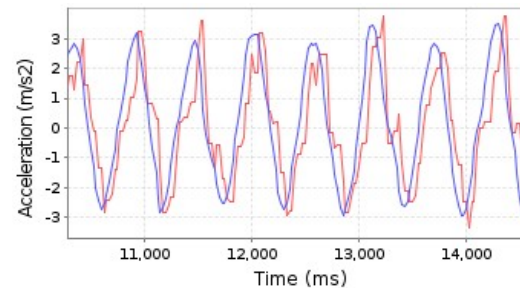
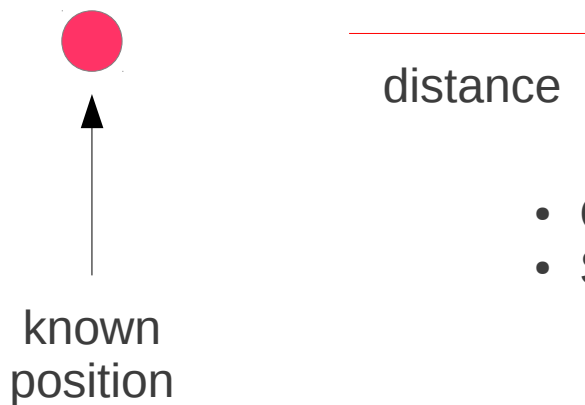
Disadvantages of WiFi Fingerprinting

1. needs WiFi coverage.
2. Scanning the WiFi environment requires substantial amount of energy.
 - one order of magnitude more than the energy requirements for the accelerometer and compass.
 - not suitable for continuous tracking.
3. Many interferences (microwave ovens, people).
4. Disruptions in communication when done excessively.

Background - PDR

How it works?

- Compute consecutive positions starting from a known position
- Distance estimation
- Direction estimation

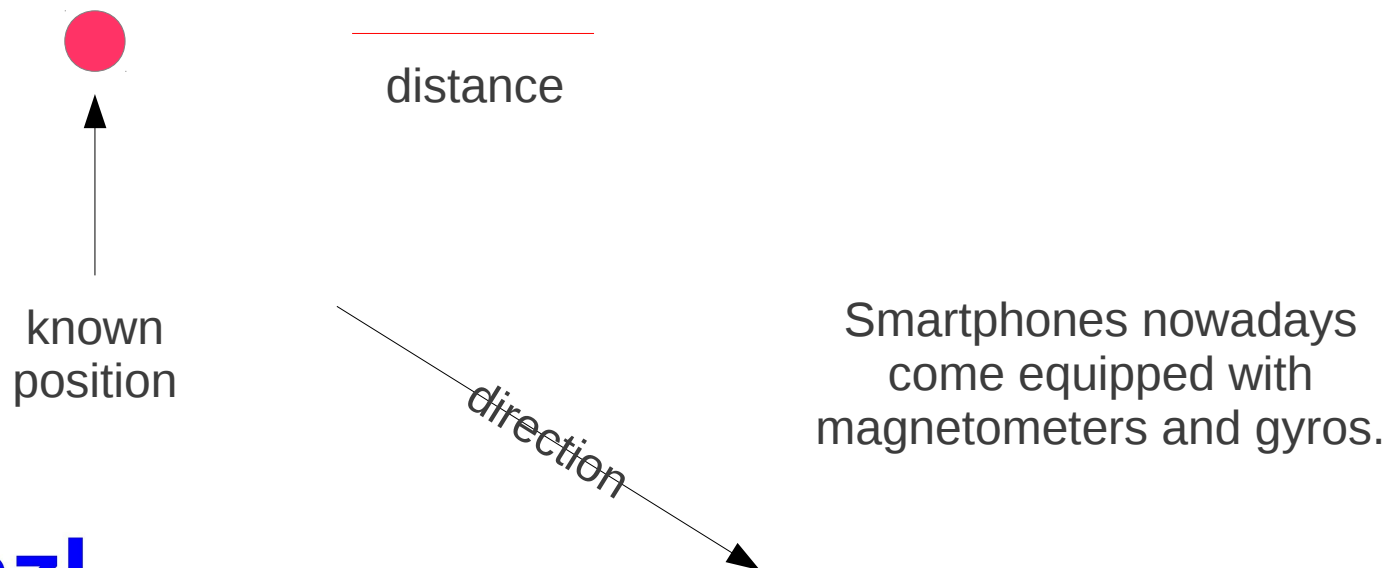


- Counting the number of steps.
- Step detection from acceleration:
 - Zero-crossing – count the number of acceleration crossing 0 value.
 - Peak detection
 - Auto-correlation – repetitiveness of human walking.
- Step length as a linear function of stepping frequency (R. Harle, 2012)

Background - PDR

How it works?

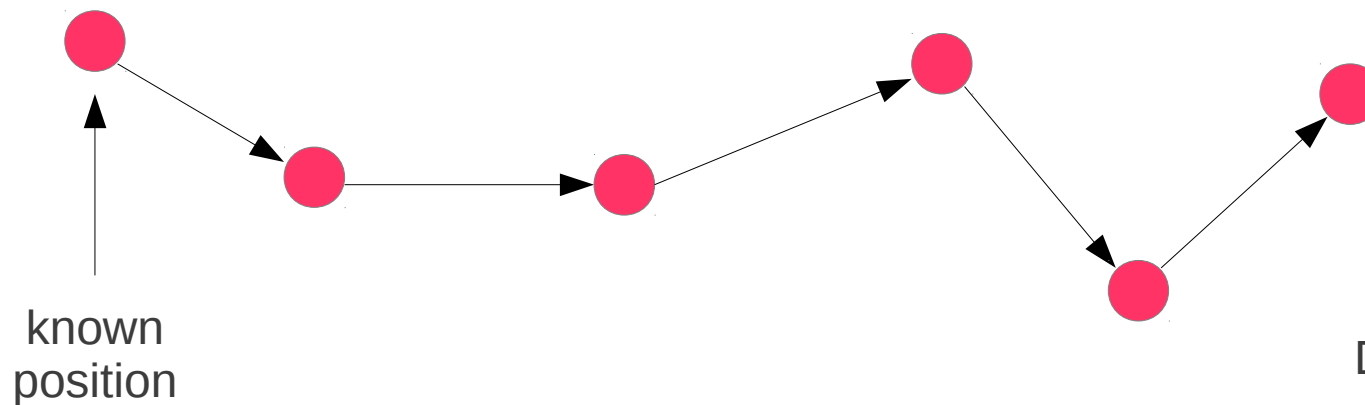
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Background - PDR

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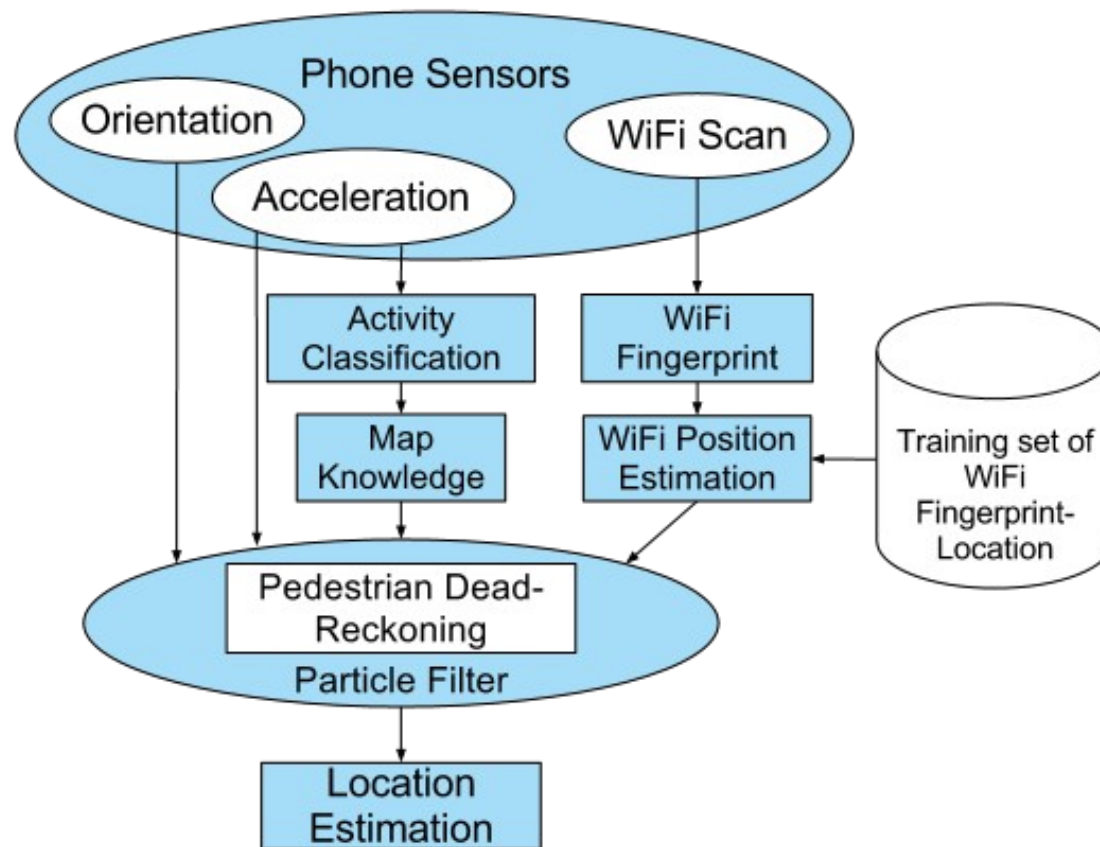
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Disadvantages:
- noisy sensors
- error accumulation

Pazl's localization solution

Application specific – PDR with periodic WiFi fingerprint and map knowledge assistance.



Activity recognition

- Activity recognition based on acceleration magnitude:

$$a = \sqrt{a_x^2 + a_y^2 + a_z^2} - g$$

- Feature extraction: in time domain (mean, standard deviation, variant, correlation between axes) and in frequency domain (energy and entropy).
- Activity classifier trained for:

- Walking
- Static
- Going up on stairs
- Going down on stairs
- Elevator moving up
- Elevator moving down
- Opening and closing doors
(both in hand and in pocket)

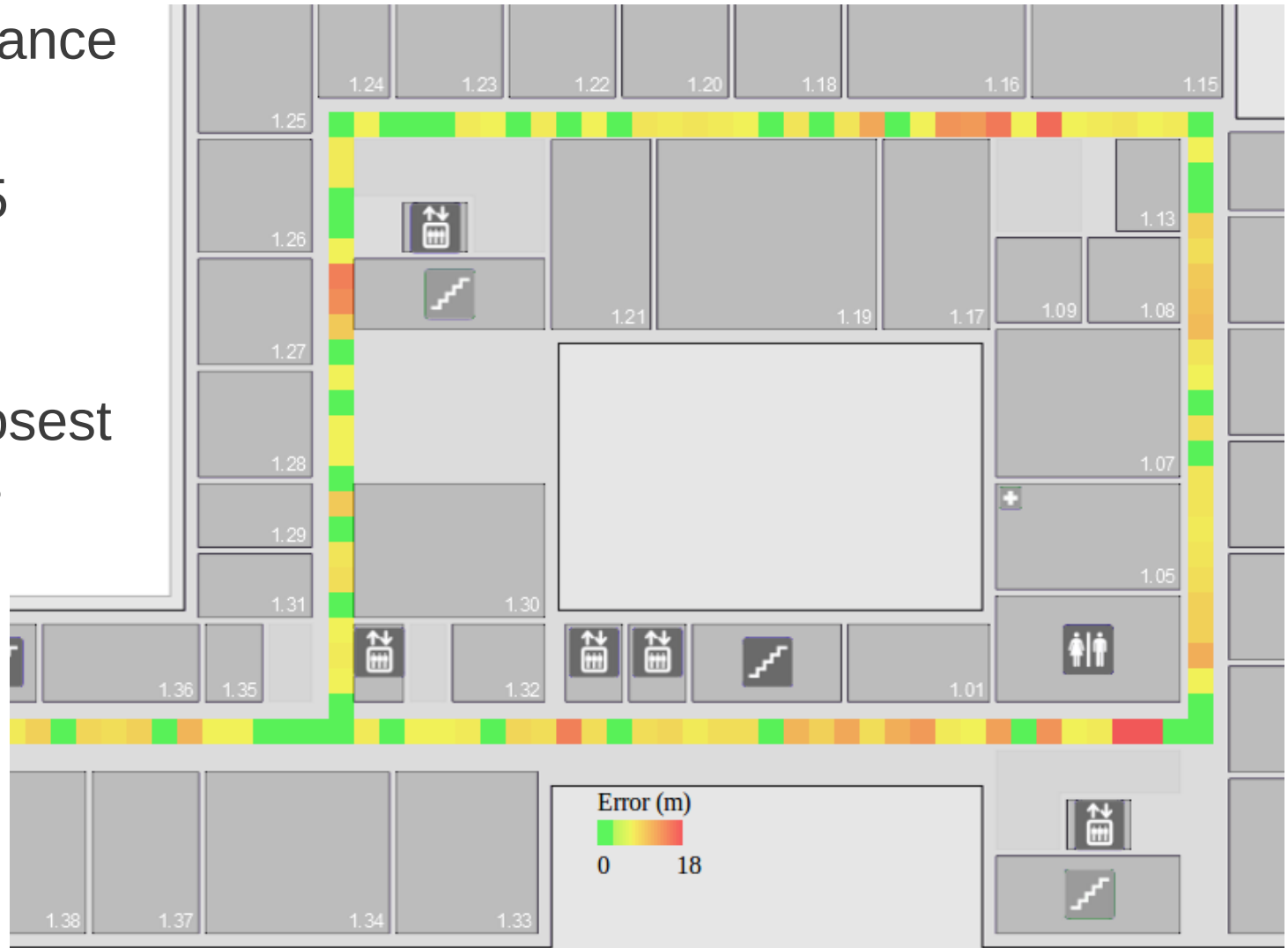
Window size	J48	Naive-Bayes	FT(tree)
128 samples	70.5%	81.7%	80.5%
256 samples	74.2%	85.3%	81.9%

- On the server Weka toolkit was used to classify the acceleration samples to activities.

WiFi fingerprinting

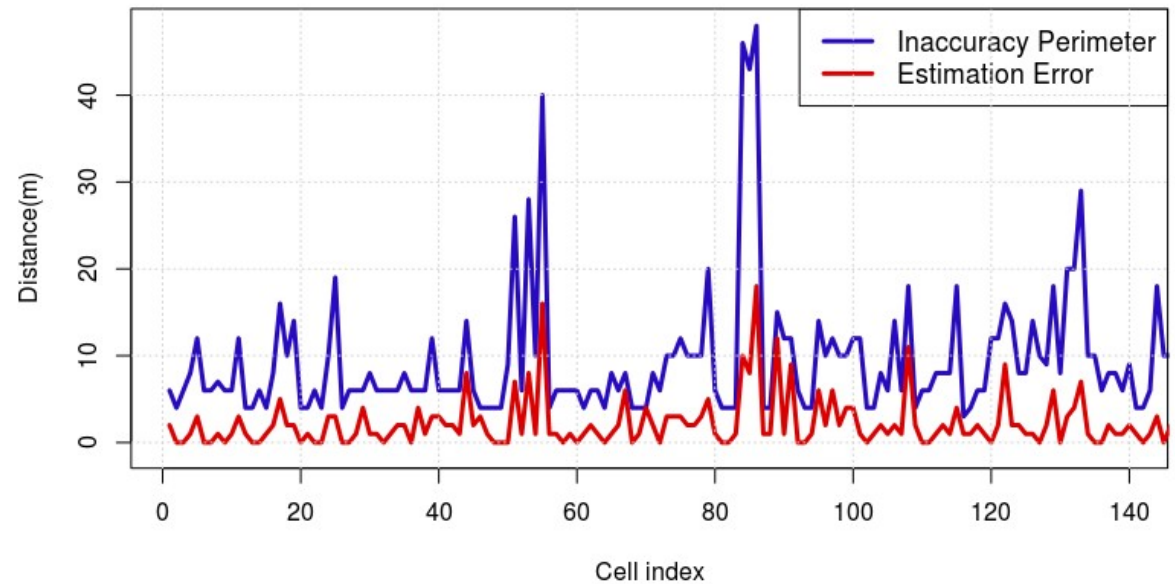
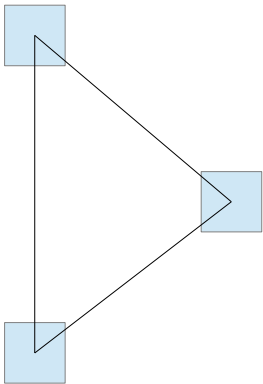
- Euclidean distance approach.
- Vector of top 5 APs in signal strength
- Centroid of closest three matches
- Cells 1x1m

We have observed that some locations have consistently better accuracy.



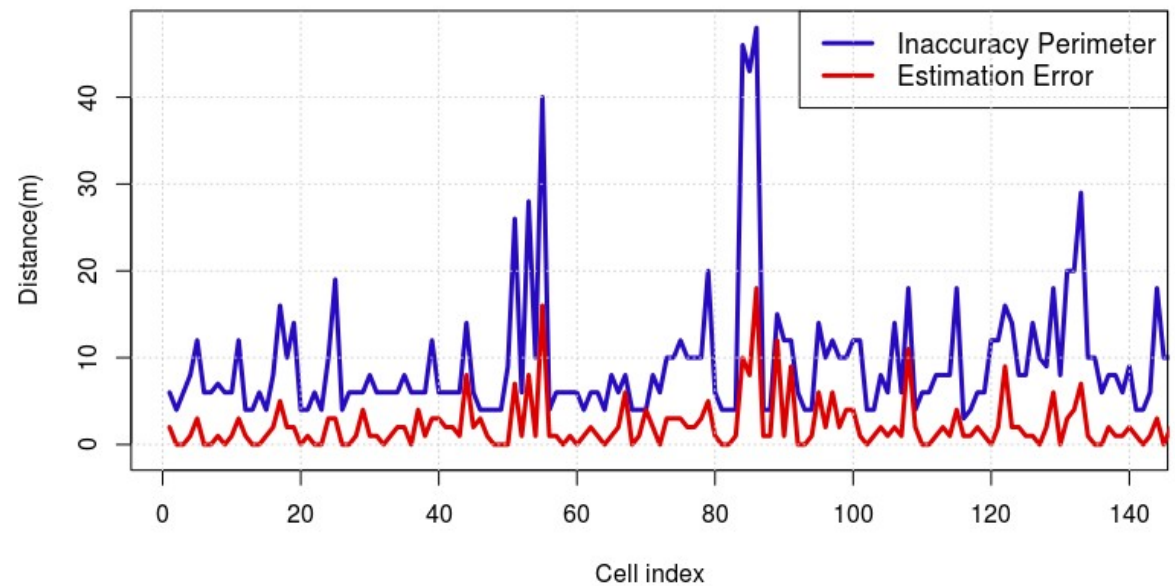
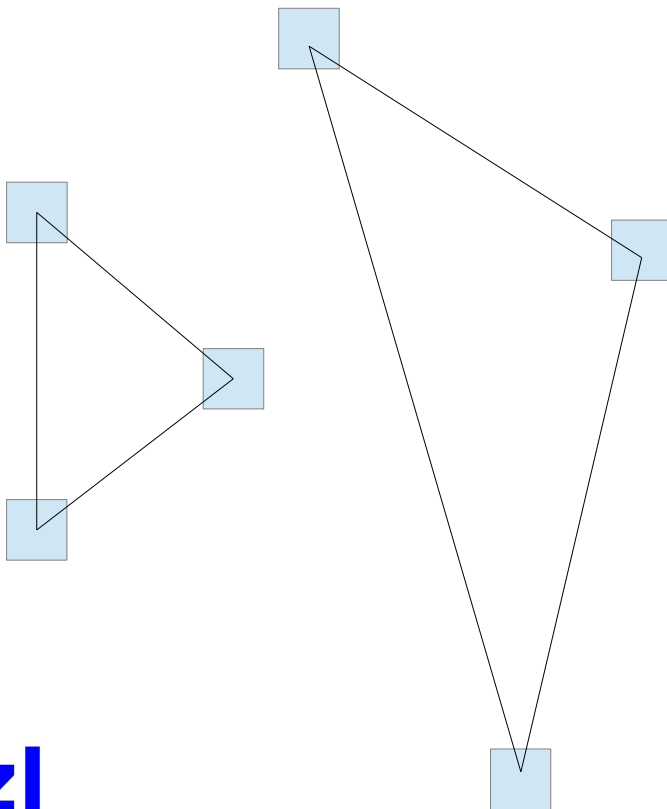
WiFi fingerprinting

- Inaccuracy perimeter – the perimeter defined by the first three closest matching fingerprints in the database.



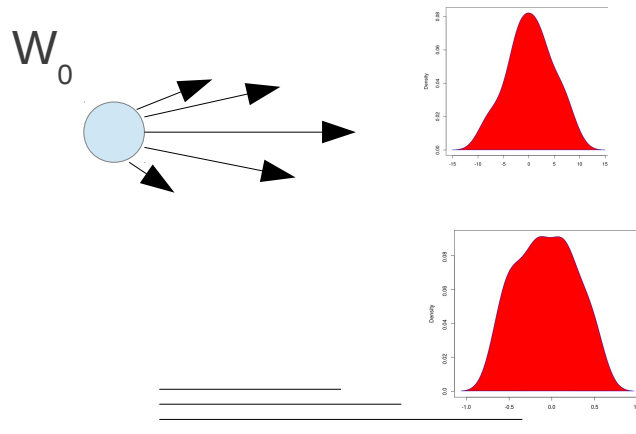
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Particle filter

- In the PDR, we observed that compass deviations and distance deviations between estimations and ground truth follow a close to normal distribution.



$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

W_c – Weight on the orientation choice

W_d – Weight on the distance choice

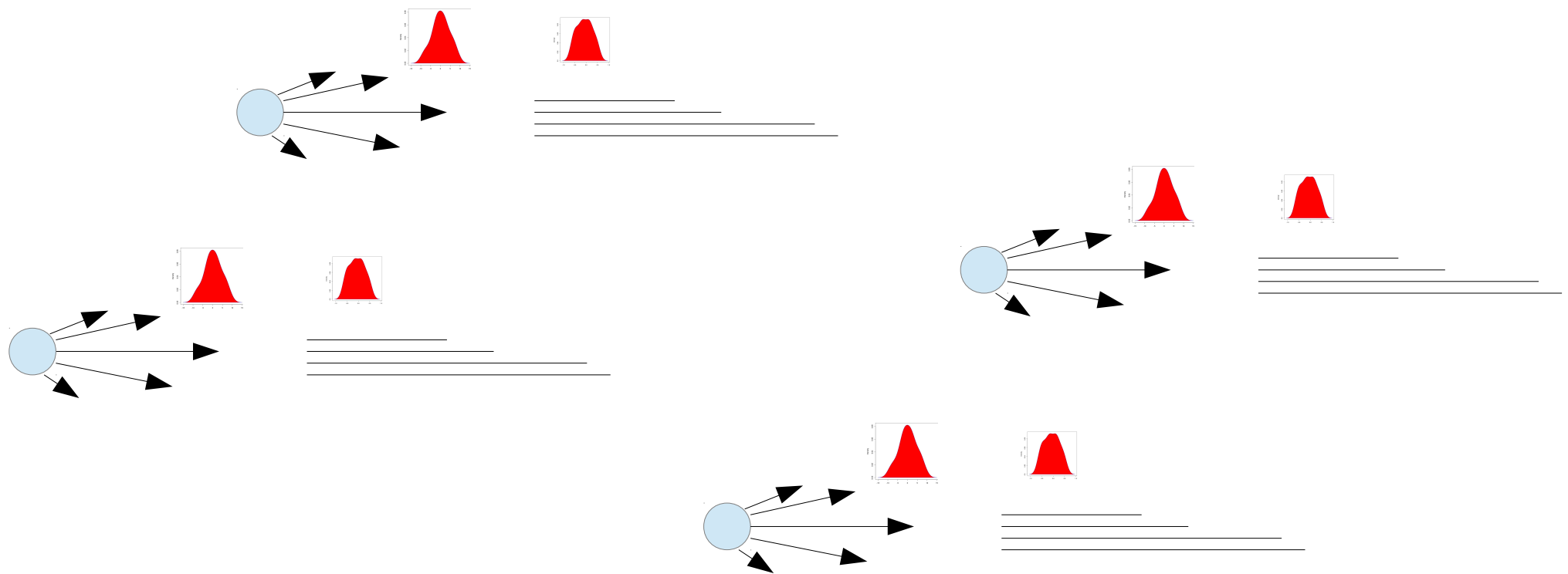
W_a – Weight from activity confidence.

W_f – Weight on distance to WiFi fix

Activity selection

$$W = W_0 + W_d + W_c + W_a + W_f$$

Particle filter

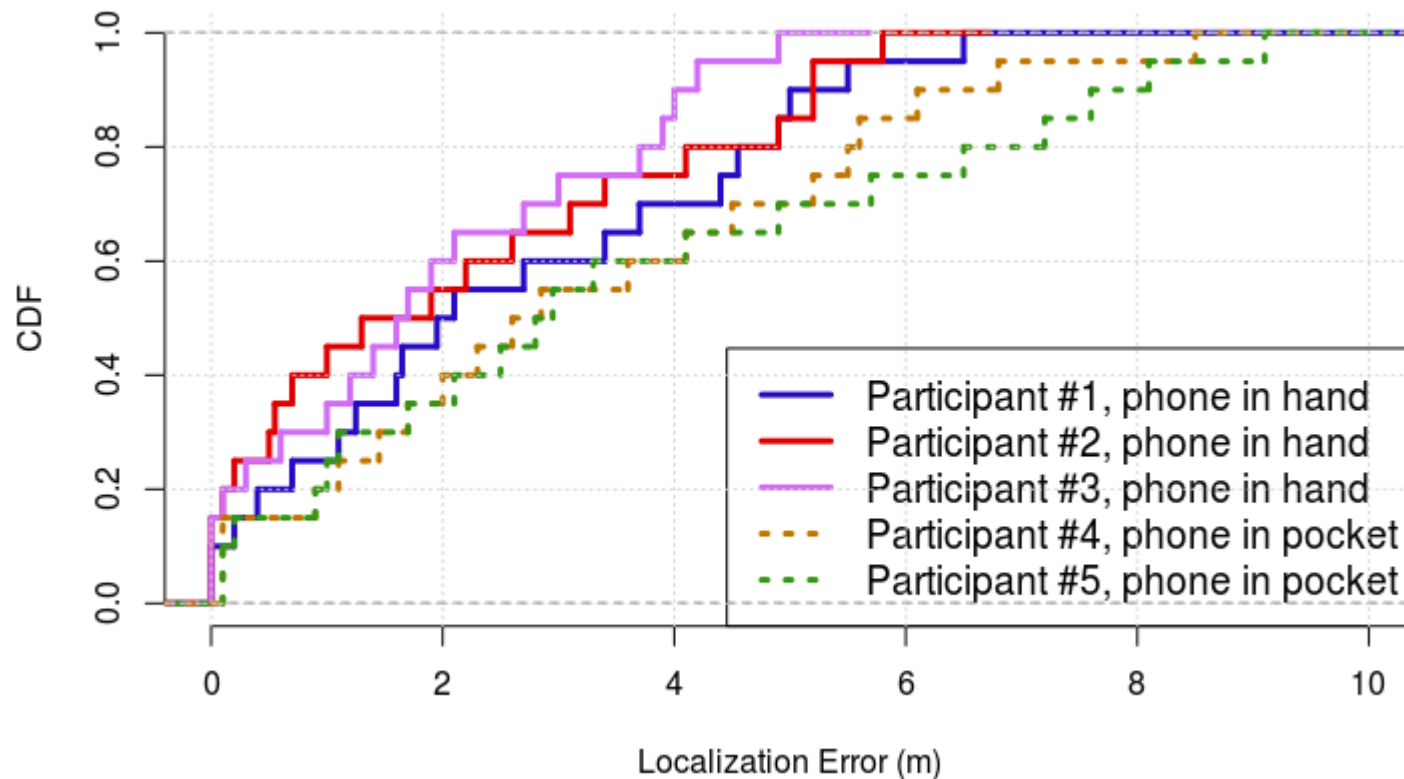


Related work

- Building radio maps for WiFi fingerprinting using PDR.
 - WiFi-SLAM (B. Ferris et al., 2007)
 - ZEE (A. Rai et al., 2012).

Evaluation – Pazl localization system

- 5 participants on a track of 100 meters.



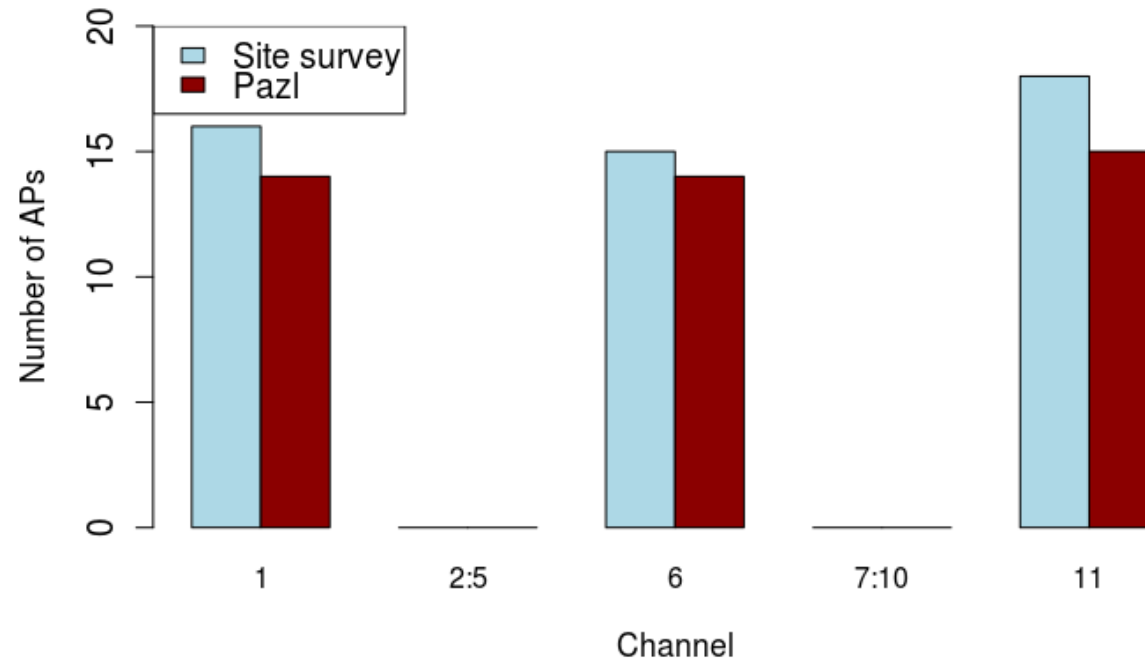
System Design

- Mobile application collecting sensor data on the phone
- Opportunistic data upload to the server for computations
- Server application running in the cloud
- Data is annotated with a location
- Create WiFi status reports

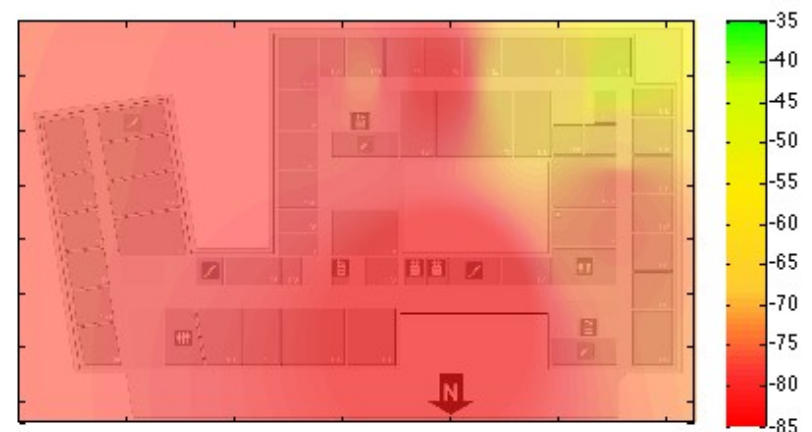
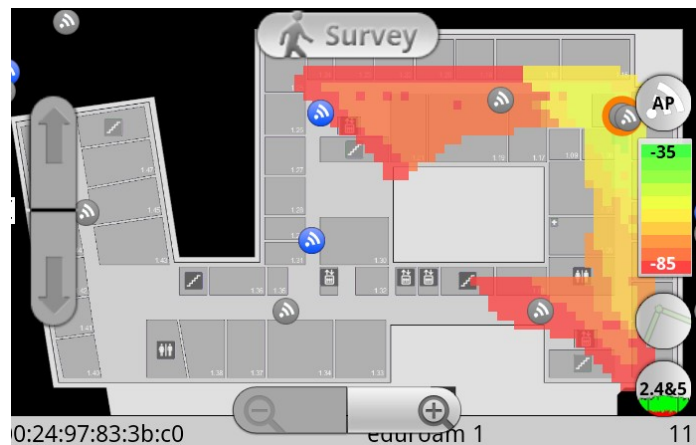
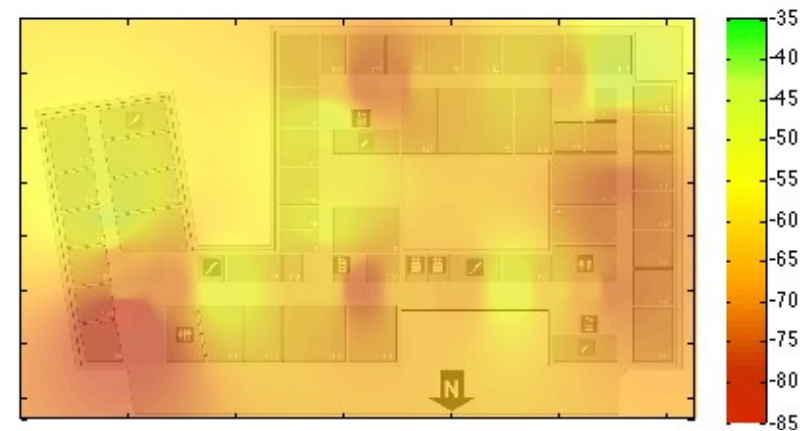
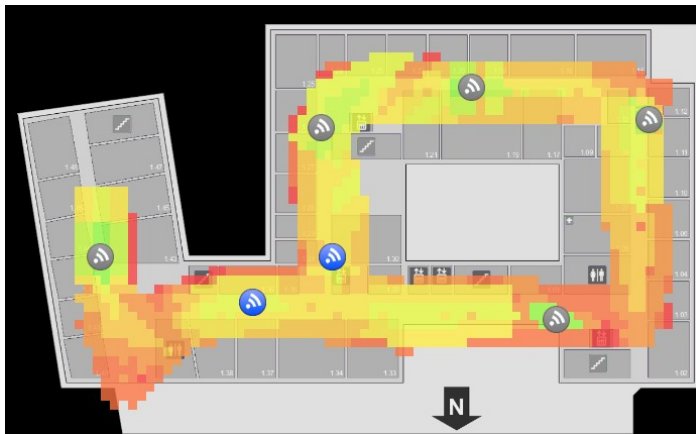
Evaluation – Pazl

- WiFi database was built by two participants.
- Activity classifier trained with the samples from two participants.
- In the experiment, 5 participants moved freely in the building for the period of a working day (10am-6pm).
- Monitoring at first floor in Informatics Forum.

Pazl site survey



Pazl compared to Ekahau



Future Work

- Remaining challenges
 - Energy-efficiency for long term running systems
 - Bootstrapping the application with indoor-outdoor transition detection
- Automate network management decisions using Pazl reports.
- Monitor other wireless environments.

Conclusions

- We move the monitoring perspective from the infrastructure to the client.
- Continuous monitoring through users mobility.
- Crowds map phenomena of common interest.
- Application specific indoor localization using a hybrid approach.

Thank you!

Questions?

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WiMo Group

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