

Leveraging RF-channel fluctuation for activity recognition

On active and passive systems; Continuous and RSSI-based signal features

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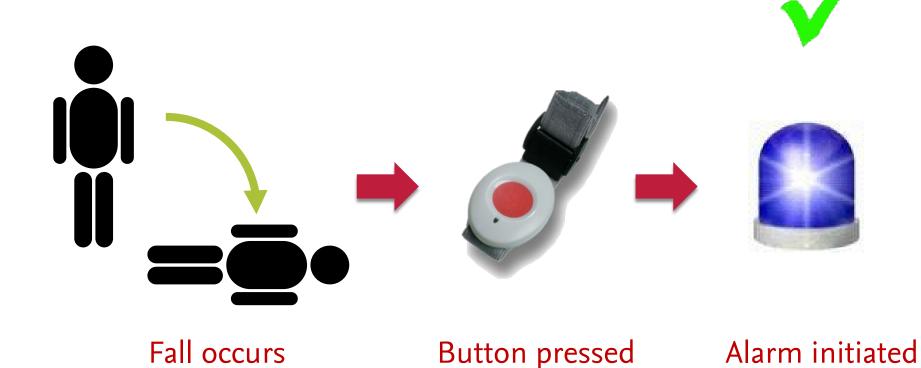
- Motivation: Why should anyone want to do this?
- Test Setup
- Results
- Conclusion



Personal Emergency Response System (PERS)

Observed behavior 1: Normal usage

Works – since several years



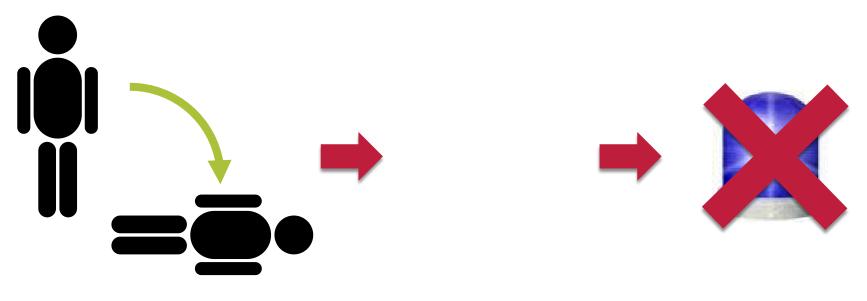


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Personal Emergency Response System (PERS) – Non-Usage

Observed behavior 2: Non-usage

- Forgets to wear sender
- Doesn't even know where it is



Fall occurs

No button to press

No alarm



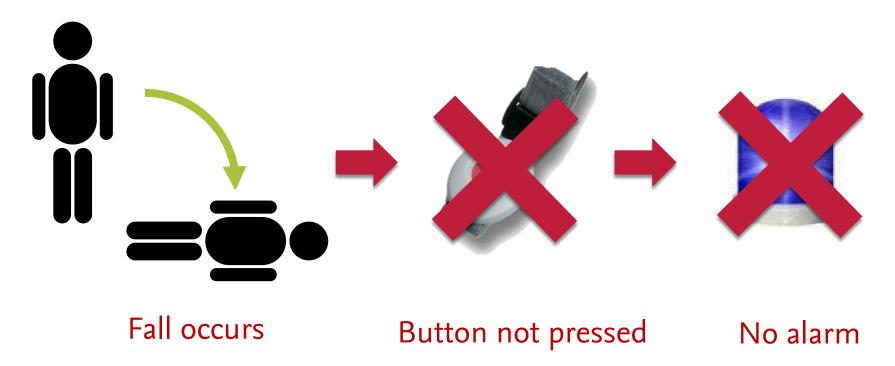
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Personal Emergency Response System (PERS) – Refused Usage

Observed behavior 3: Refused usage

- "The mobile nursing service will come anyway in 2-4 hours"
- "I'll wait here on the floor don't want to bother anyone"





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Personal Emergency Response System (PERS) – Anxious Usage

Observed behavior 4: Anxious usage

- "I feel safe I better wear that thing 24/7"
- Many false alarms during nighttime



Bad dream occurs

Button accidentally pressed



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False Alarm

Summary: Personal Emergency Response System (PERS)

- 1. Normal usage
 - OK, this is boring...
- 2. Non-usage
 - Additional stationary sensors for activity detection
- 3. Refused usage
 - Additional stationary sensors for activity detection
- 4. Anxious usage
 - Only stationary sensors for activity detection

\rightarrow Using RF-channel fluctuation for activity recognition may be a good idea!







Motivation for RF-based Activity Detection

- Worn devices are not always present when needed
- Most IoT-devices include a radio transceiver
- No interaction is required
- Not intrusive like cameras
- Higher acceptance



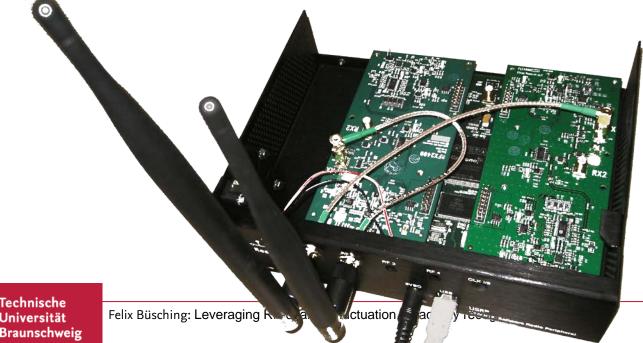
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USRP1: Active SDR-based DFAR

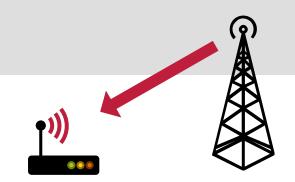
- 900 MHz (RFX900 board)
- Vert900 Antenna
- Sine signal, continuously modulated on carrier
- 80 Hz sample rate





USPR N210: Passive SDR-based DFAR

- 82.5 MHz (WBX board),
- Vert900 Antenna
- Environmental FM radio from nearby radio station
- 64 Hz sample rate





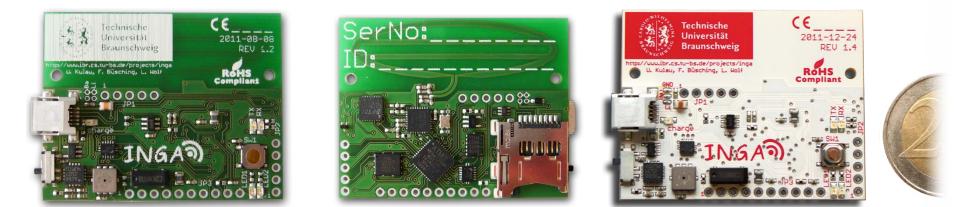
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INGA WSN-Node: Active RSSI-based DFAR

- 2.4 GHz IEEE802.15.4 radio
- PCB-Antenna
- UDP-Packets
- 100 packets per second
- RSSI-values from radio (received MAC-Frames)





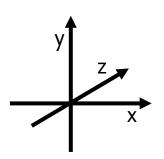
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Iphone 4: Accelerometer-based activity recognition

- 3 axis accelerometer
- Attached to body
- 40 Hz sampling rate
- Recorded simultaneously to other experiments



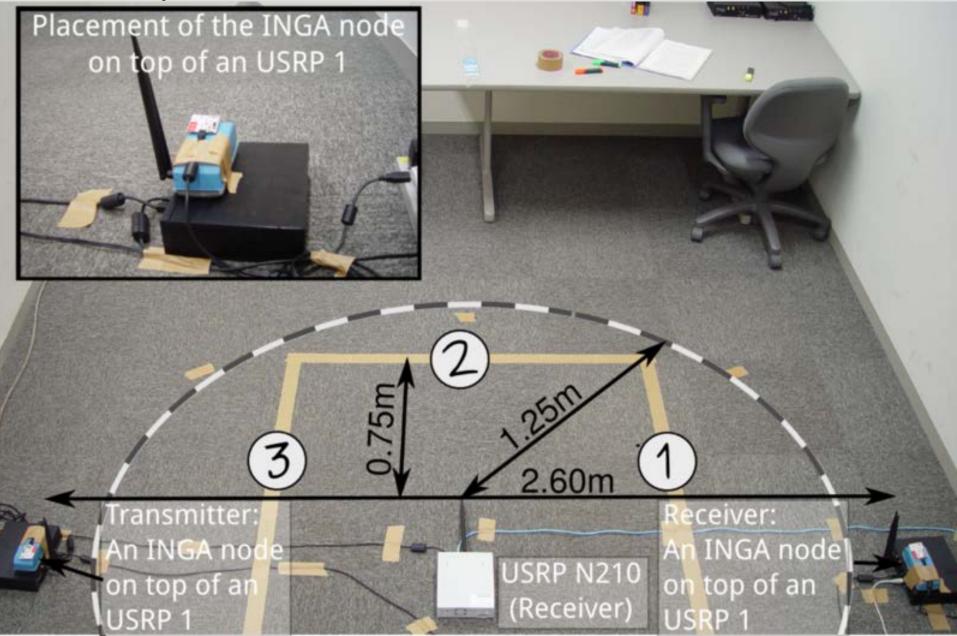


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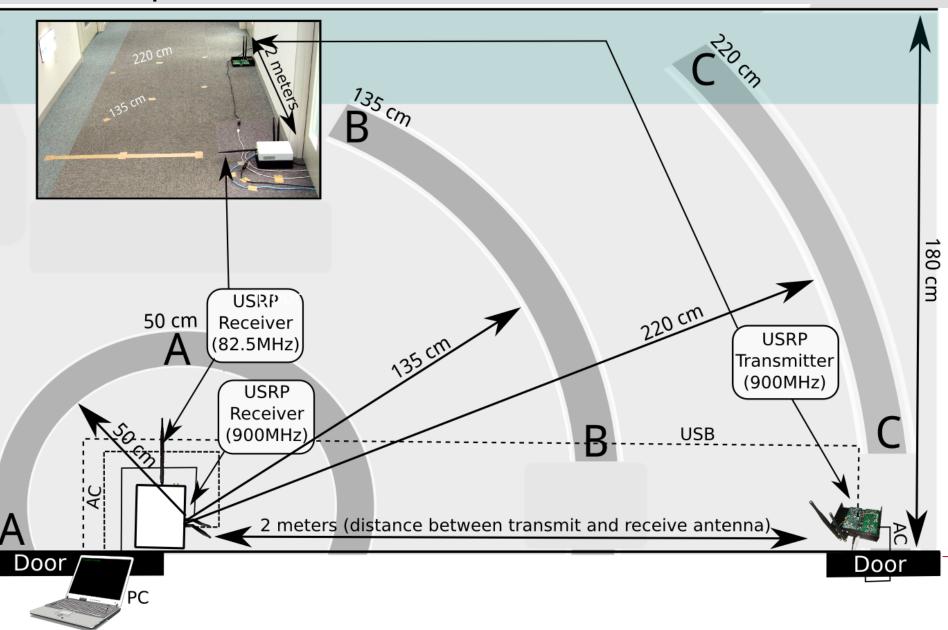


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Test Setup: Scenario 1



Test Setup: Scenario 2



Performed Activities

Empty room

Standing

At 3 different locations

Lying

At 3 different locations

Walking

At 3 different speeds

Crawling



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Summary: Different Approaches and Scenarios

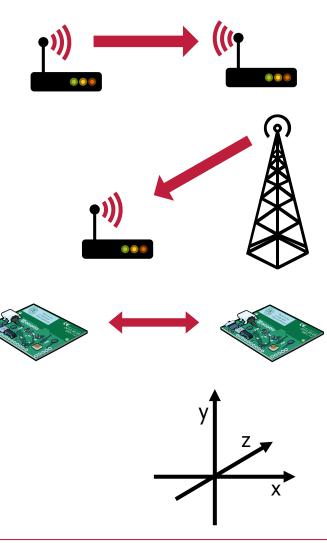
Approaches

- Active SDR-based DFAR
- Passive SDR-based DFAR
- Active RSSI-based DFAR
- Accelerometer-based activity recognition

2 Scenarios

- Room
 - All four approaches evaluated
- Hallway
 - No WSN evaluation





- Motivation
- Test Setup
- Results
- Conclusion



Confusion Matrices – Classification Accuracy – k-NN algorithm 1/2

Accelerometer-based activity recognition

	Lying	Standing	Walking	Crawling
Lying	.976	.024		
Standing		1.0		
Walking			•955	.045
Crawling			.253	.748

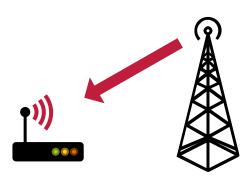
Passive SDR-based DFAR

	Lying	Standing	Walking	Crawling
Lying	1.0			
Standing	.056	.980	.022	
Walking	.023		.874	.102
Crawling	0.44		.144	.811



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Confusion Matrices – Classification Accuracy – k-NN algorithm 2/2

Active SDR-based DFAR

	Lying	Standing	Walking	Crawling
Lying	.904	.096		
Standing	.096	.898	.006	
Walking		.013	.962	.025
Crawling		.038	.212	.750



Active RSSI-based DFAR

	Lying	Standing	Walking	Crawling
Lying	.882	.118		
Standing	.120	.869	.007	.004
Walking			•953	.047
Crawling		.010	•439	.551



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Device-free, no worn sensors, no interaction required Monitoring of elderly by RF-Channel Activity Recognition

- works!
- ... nearly as good as accelerometer based
- (Passive is not that good)

Monitoring smart spaces works for simple activities.

Outlook

Only RSSI-based information will be needed for activity detection

