

# Indoor Positioning via Three Different RF Technologies

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# Outline

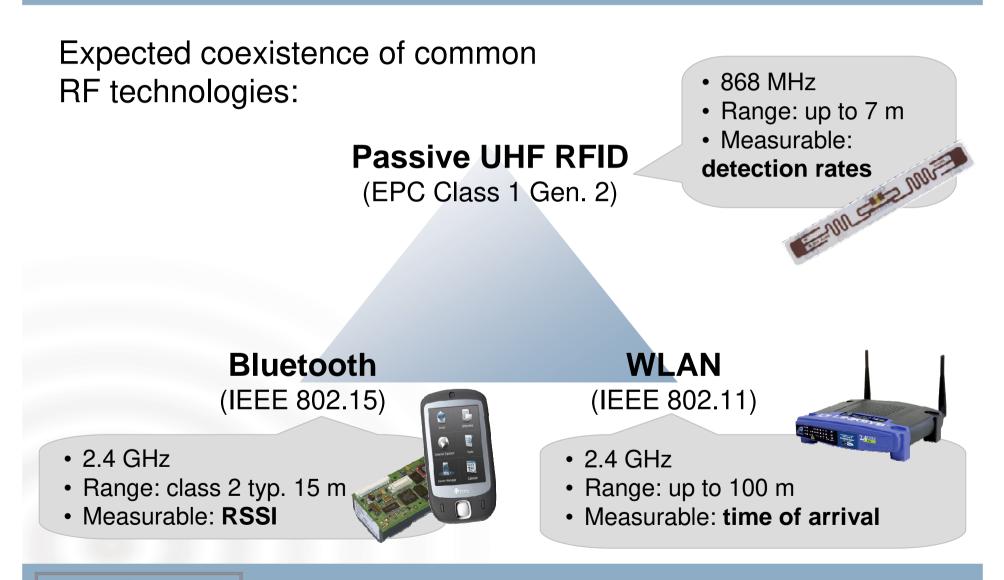
- Introduction
- Positioning methods based on
  - RFID
  - Bluetooth
  - WLAN
- Experiments (incl. video)
- Conclusion



#### Introduction

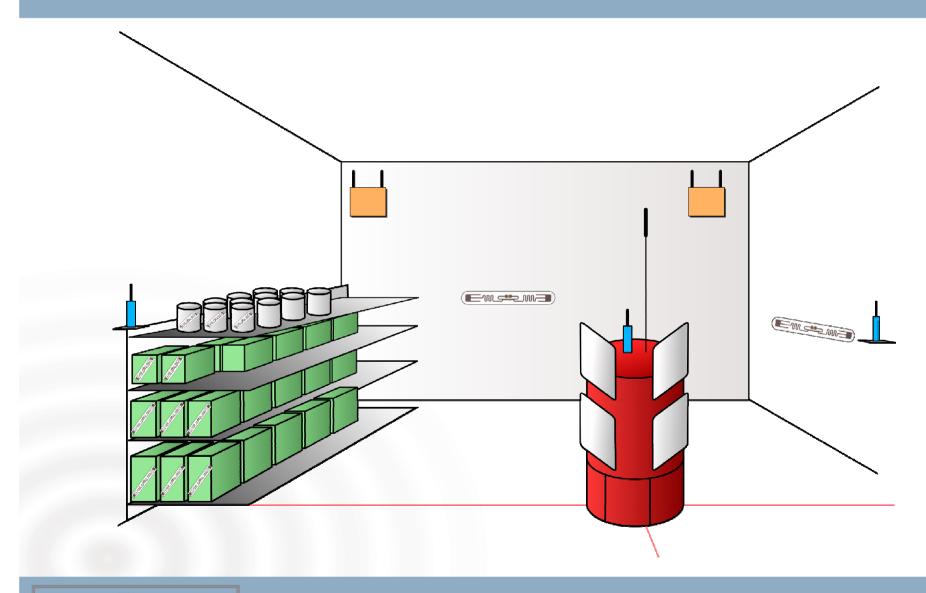
- Positioning: Position estimation in a given environment by means of sensor information
- Position information highly relevant for context-aware services and tracking purposes
- Potential scenarios
  - Patient and asset tracking
  - Product localization
  - Warehousing and logistics
  - Positioning for mobile systems, e.g. transport containers, autonomous vehicles, persons with laptops
- GPS fails indoors ⇒ requirement for alternatives
- Desirable: reuse of existing, inexpensive infrastructure

#### Focus on Radio Frequency Technologies



**AMBI SENSE** P. Vorst, J. Sommer, C. Hoene (University of Tübingen, Germany)

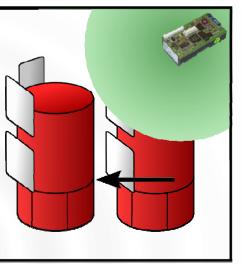
#### Positioning – General Idea

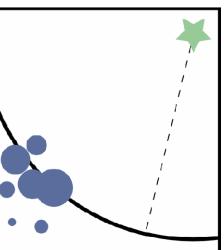


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# Particle Filtering

- Estimation of the state of a dynamic system
   Here: location of a mobile system
- Bayesian filtering technique, probability density function (PDF) over state space
- Discrete approximation of the PDF by set of weighted samples
- Robust and accurate, applicable to virtually any sensor
- Iterations of prediction, correction, normalization, and resampling





Prediction (motion model)

Correction (sensor model)

Normalization + resampling

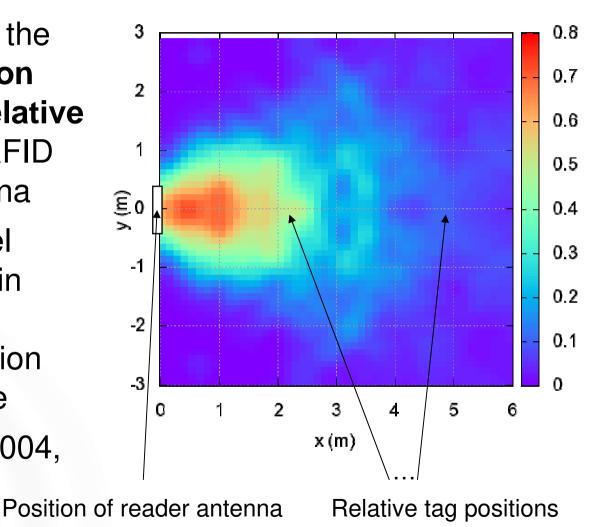
# 1. Positioning via Passive UHF RFID

- Near future: palettes, cartons, and products RFID tagged
- Mobile system carries RFID reader
   one reader only, lots of inexpensive tags
- Usual positioning method: proximity to tag of known position determines cell-based location
- Shortcomings:
  - Position resolved to coarse area only
  - Well-known problems of passive tags: false negatives, reflections, ...
- Our goal: accurate, metric localization

#### Positioning via Passive UHF RFID - cont'd

- Idea: Exploitation of the fact that tag detection rates depend on relative position between RFID tag and RFID antenna
- Detection rate model (see figure) is used in particle filtering
   probabilistic position refinement over time
- See (Hähnel et al. 2004, Vorst et al. 2008)

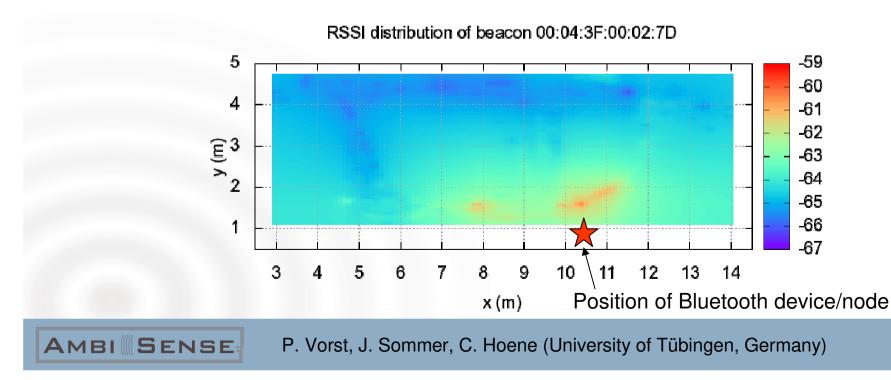
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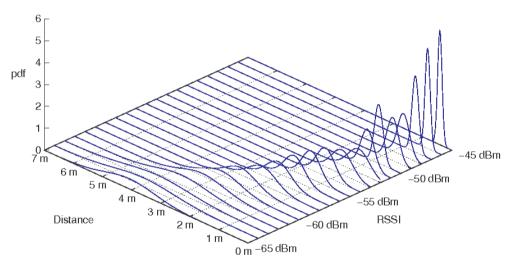
# 2. Positioning via Bluetooth

- Variety of mobile devices equipped with Bluetooth radio transceivers
- Received signal strength (RSSI) can be measured
- RSSI values decrease with distance between sender and receiver ⇒ distance estimation



# Positioning via Bluetooth – continued

- Each RSSI value can be assigned a PDF over possible distances
- Observation: noise, low resolution for small RSSI values
- Positioning: multilateration (e.g., MMSE), particle filtering
- PDF used for particle reweighting in correction step

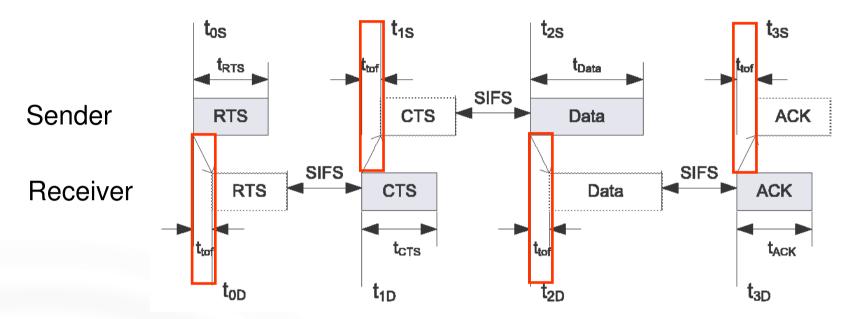


# 3. Positioning via Wireless LAN

- Usual positioning approach with WLAN: usage of RSSI values
- Alternative: time of arrival (TOA)
- Idea: Position has impact on the time of flight of WLAN packages between sender and receiver
- Advantage: TOA measurements scale linearly with open-air propagation distances
- Challenge: low clock resolution of off-the-shelf hardware (1µs ~ 300 m)



# Positioning via WLAN – continued

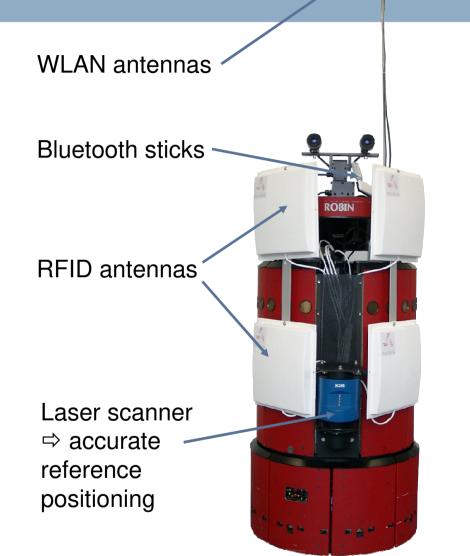


- Novel four-way TOA: TOA measurements conforming to IEEE 802.11 protocol using 4 transmission steps
- Improvement by averaging over 500-2000 packets
- Open-source software Goodtry provided freely
- See (Hoene et al. 2008)

# **Experimental Setup**

Mobile service robot (RWI B21)

- UHF RFID reader (ALR-8780)
- 2 Bluetooth USB sticks
- 2 WLAN PCI cards + antennas (for pings and TOA measurements)
- 240° laser scanner (reference positioning)



#### Experimental Setup – Environment

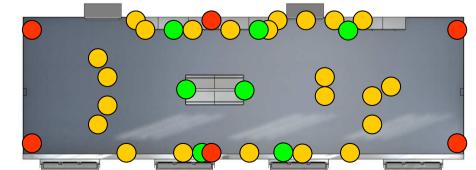
Laboratory with landmarks of known positions

- 24 RFID tags (Alien Techn. "Squiggle")
- 7 Bluetooth nodes (BTnodes, ETH Zürich)
- 6 WLAN access points



Additionally: 400 RFIDlabeled products of unknown positions in a supermarket shelf (metal)

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#### **Experimental Results**

- Data: 11+4 sample trajectories with RFID/BT+WLAN recordings plus accurate laser reference positions, > 5 min.
- Particle filter with 300 samples using odometry
- Investigation: Tracking, i.e., coarse initial pose estimate provided; mean absolute positioning errors over time

Method	Mean $\pm$ Std. dev.	Median	90th percentile	
RFID (model)	<b>0.432 m</b> ± 0.095 m	0.435 m	0.527 m	
Bluetooth	<b>0.494 m</b> ± 0.149 m	0.474 m	0.678 m	
WLAN	<b>3.315 m</b> ± 0.738 m	3.545 m	4.274 m	



#### [ play video ]

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# Conclusion

- Presented: Three RF-based positioning techniques
  - RFID tag detection rates
  - Bluetooth signal strength
  - WLAN time-of-arrival measurements
- Accuracies obtained in tracking a mobile robot:
  - $\approx$  0.5 m for RFID, Bluetooth
  - $\approx$  3-4 m for WLAN
- Low-cost, off-the-shelf hardware used in common RF infrastructures
- Future work:
  - Fusion of the techniques ⇒ easily possible due to particle filters
  - Refinements of methods and experiments in larger environments

#### Thank you for your interest!

#### Acknowledgments

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- Spokesperson of the research cooperation: Prof. Dr. Andreas Zell Wilhelm Schickard Institute for Computer Science Dept. Computer Architecture Sand 1 D-72076 Tübingen

### References

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- D. Hähnel, W. Burgard, D. Fox, K. Fishkin, and M. Philipose. *Mapping* and Localization with RFID Technology. In Proceedings of the 2004 IEEE International Conference on Robotics and Automation (ICRA 2004), pages 1015-1020, 2004.
- P. Vorst and A. Zell. European Robotics Symposium 2008, volume 44/2008 of Springer Tracts in Advanced Robotics, chapter Semi-Autonomous Learning of an RFID Sensor Model for Mobile Robot Selflocalization, pages 273-282. Springer Berlin/Heidelberg, February 2008.
- C. Hoene, J. Willmann. Four-way TOA and Software-Based Trilateration of IEEE 802.11 Devices, IEEE International Symposium on Personal, Indoor and Mobile Radio Communications (PIMRC), Cannes, Sep. 2008.

# **Picture Credits**

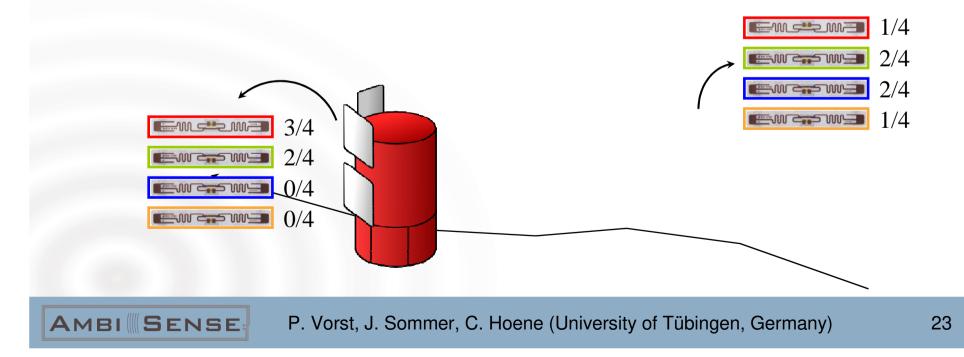
- RFID Tag (slides 2,4,5,23) from Alien Technology (http://www.alientechnology.com/tags/index.php)
- BTnode (slides 4,6) from ETH Zürich (http://www.btnode.ethz.ch/Main/Purchase)
- PDA (slides 2,4) from PIXmania (http://www.pixmania.lu/lu/de/554386/art/htc/pda-mit-telefonfunktiont.html)
- WLAN router (slides 2,4,11) from Litec Computer (http://www.litec-computer.de/popup\_image.php?pID=9611/imgID=0)
- Other pictures: courtesy of the corresponding AmbiSense subprojects

#### Extra slides

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# Positioning via RFID Snapshots

- Further possibility: fingerprinting
- Prior training: learning of a database of RFID measurements for different positions
- Positioning phase: Comparison of current list of detected tags with trained database
- Again: particle filtering to increase robustness



#### **Experimental Results**

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Method	Mean $\pm$ Std. dev.	Median	90th percentile		
RFID (model)	<b>0.432 m</b> ± 0.095 m	0.435 m	0.527 m		
RFID (snapshots)	0.264 m ± 0.047 m	0.267 m		higher tag (higher tag density)	
Bluetooth	<b>0.494 m</b> ± 0.149 m	0.474 m	0.678 m		
WLAN	<b>3.315 m</b> ± 0.738 m	3.545 m	4.274 m		