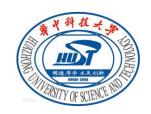
## LiFS: Little Human-Effort, Device-Free Localization with Fine-grained Subcarrier Information

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### **Motivation**

### Location is the key for many applications

#### **Academia**

[RADAR'00], [Cricket'04],

#### **Industry**





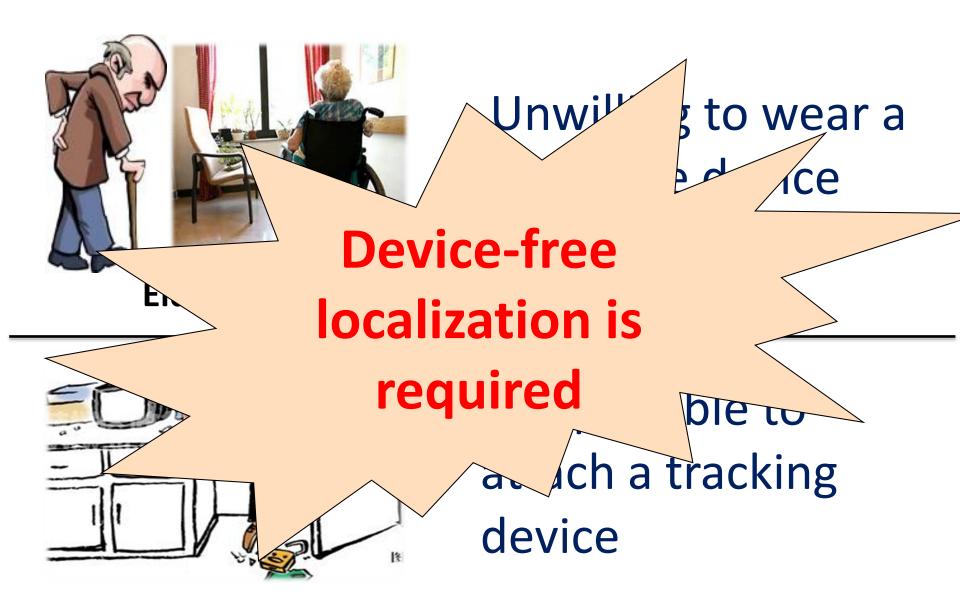
### Why device-free localization?

rack'15],[Chronos'16]...





### **Motivation**



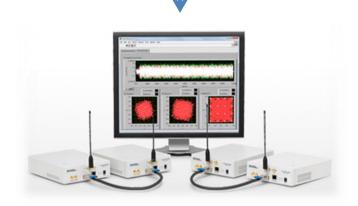
**Intruder Detection** 

### **Limitations of Existing Work**

**High labor-cost/hardware-cost** 



Build Fingerprint
Database



Demand Software-Defined-Radio

Can we do better?

### Our Approach: LiFS

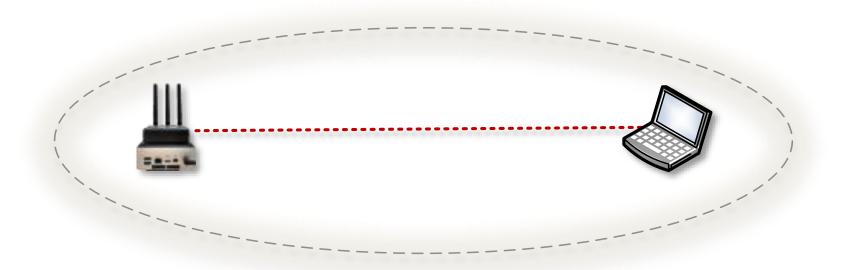
### A model-based device-free localization system implemented on commodity Wi-Fi device

• Low labor cost: no labor-intensive offline training.

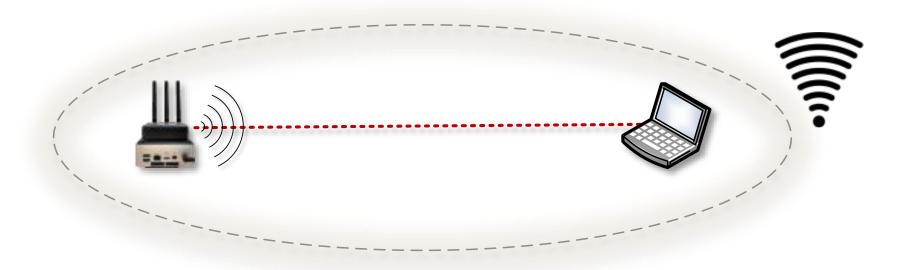
Low hardware cost: cheap commodity Wi-Fi device.

 High accuracy: around 1 m median accuracy for device-free passive localization.

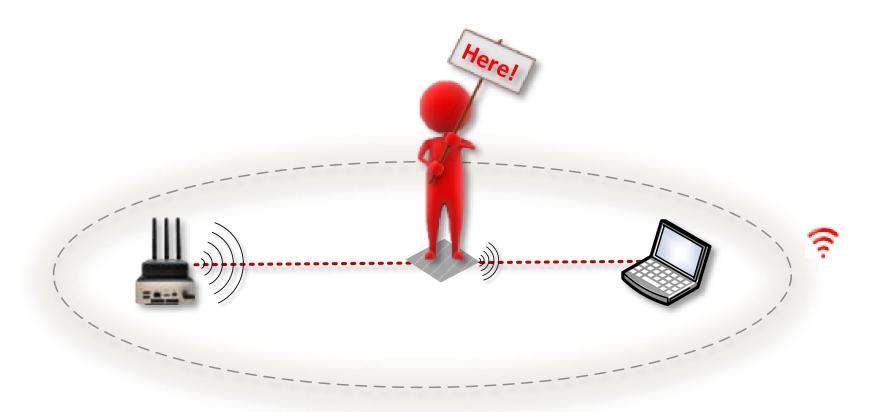
### **How does LiFS Work?**



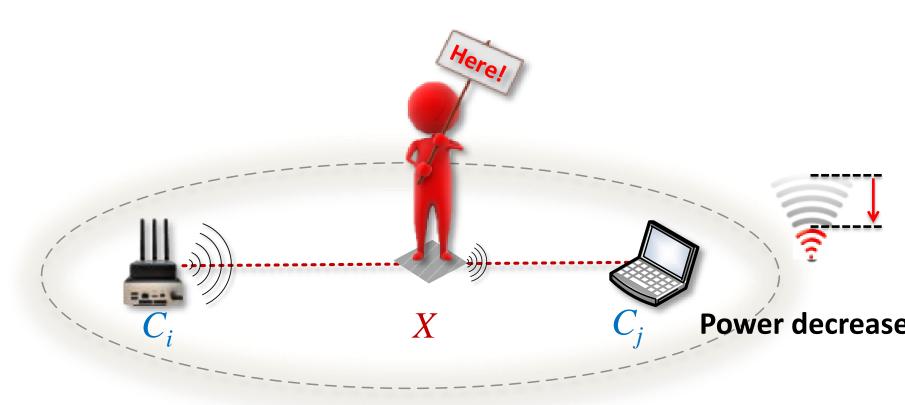
Considering a Wi-Fi link



We can easily measure the signal power



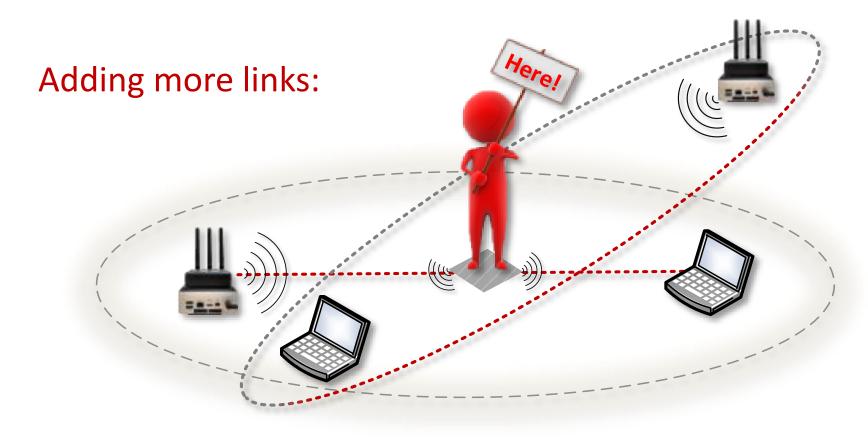
The signal power decreases if a target appears



Wireless Propagation theory shows:

**Power decrease** is  $R_{ij} = f(C_i, C_j, X)$ 

where  $C_i$ ,  $C_i$  and X are locations of AP, client and target



Employing models to determine the target's location, avoiding offline training

### The idea is straightforward!

But, does it work in reality?

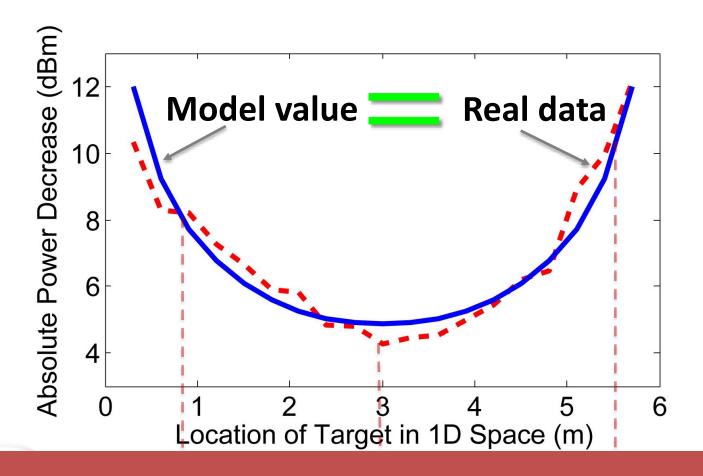
### **Outdoors: Few Multipath**





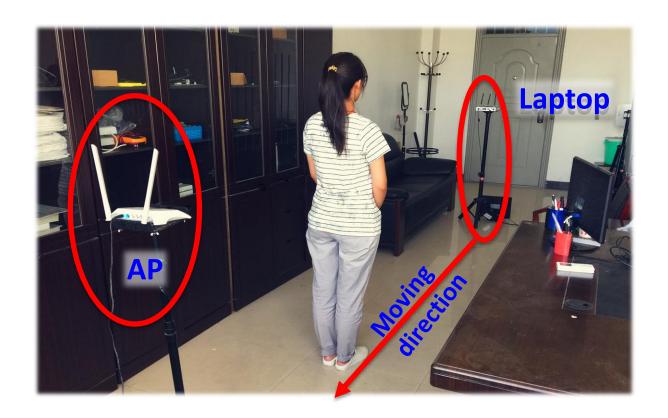


### **Outdoors: Few Multipath**



Target can be localized accurately with models due to few multipath

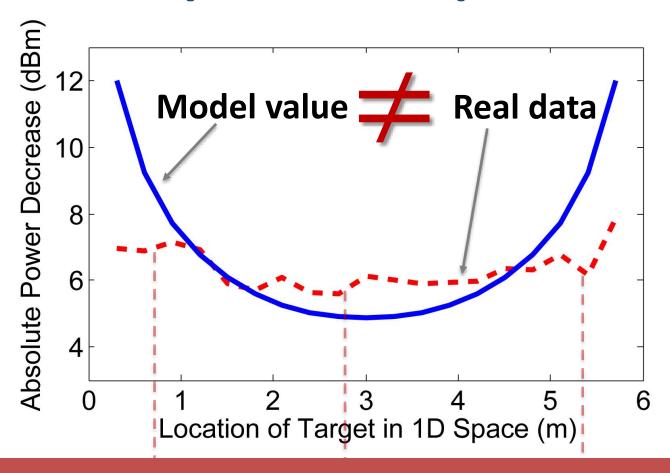
### **Indoors: Rich Multipath**





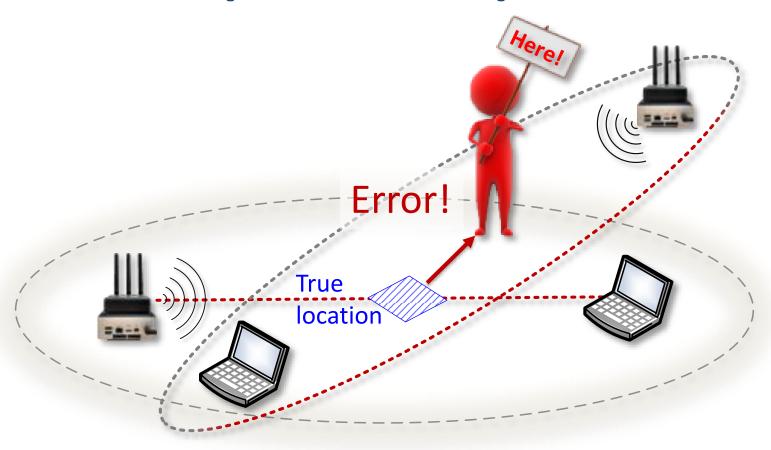


### Multipath is the problem



Real data does not match model values indoors due to rich multipath!

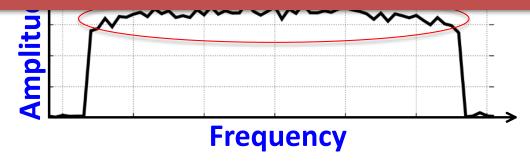
### Multipath is the problem



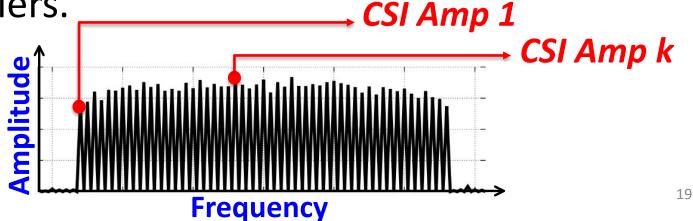
Large errors occur if employing raw data for localization indoors!

### **Understand Power Decrease**

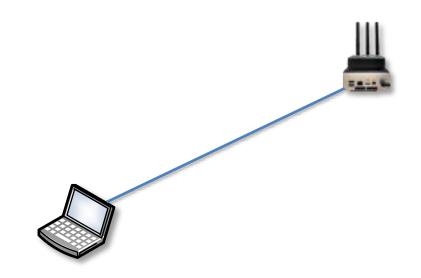
### CSI provides more detailed information across multiple subcarriers



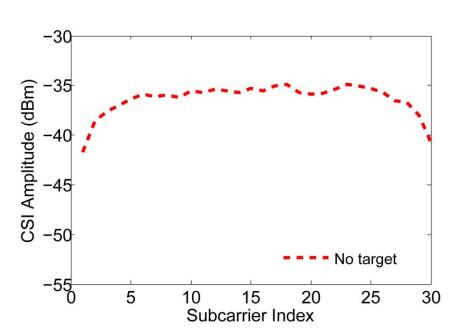
• What if, looking power at *CSI* over multiple subcarriers.

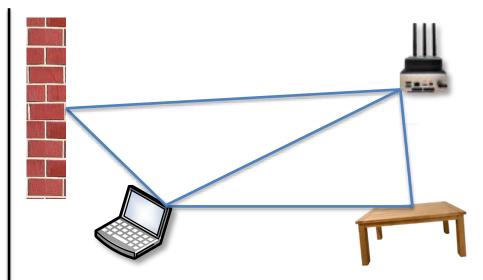


### **Understand Power Decrease**

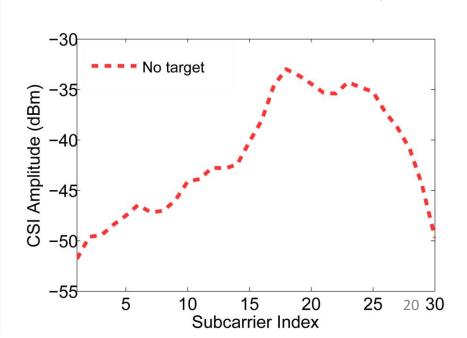


### **Outdoor with few multipath**





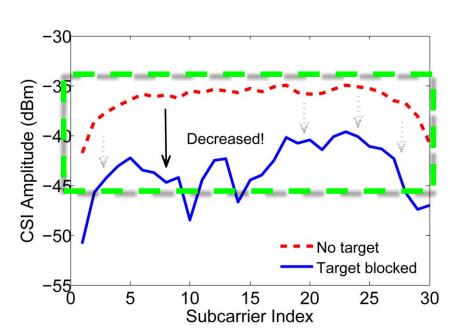
#### Indoor with rich multipath



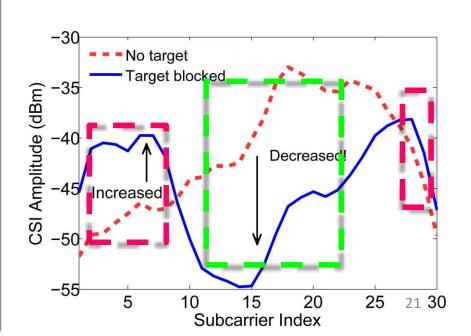
### **Understand Power Decrease**

### Not all subcarriers are affected equally by multipath!

#### **Outdoor with few multipath**



#### Indoor with rich multipath



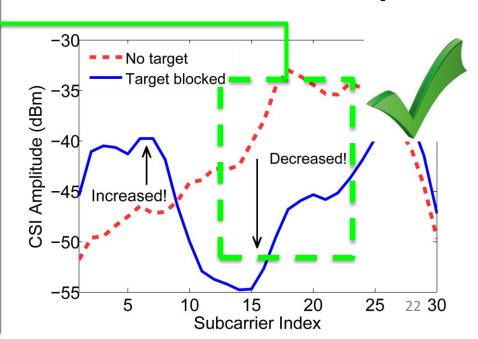
### **Key Idea**

### Identify "clean" subcarriers not affected by multipath for localization

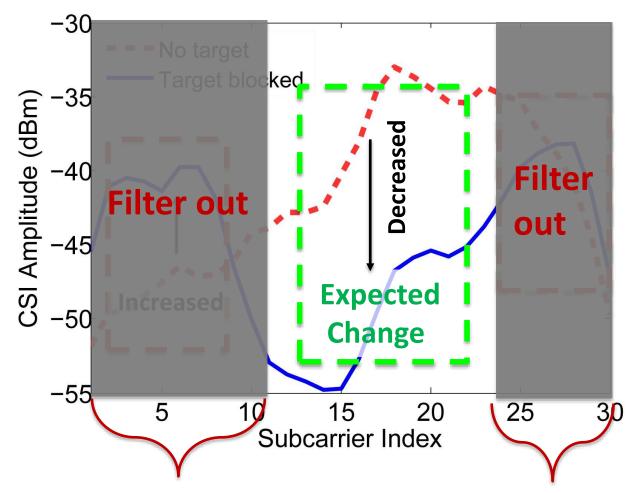
#### **Outdoor with few multipath**

# (mgp) ephilidury Decreased! Decreased! Clean Subcarriers 50 5 10 15 20 25 30 Subcarrier Index

#### Indoor with rich multipath



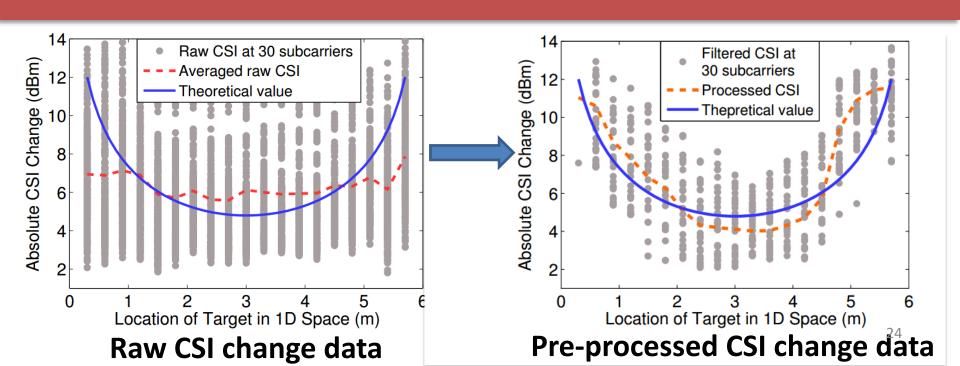
### Find "Clean" Subcarriers: Pre-Processing Method for CSI



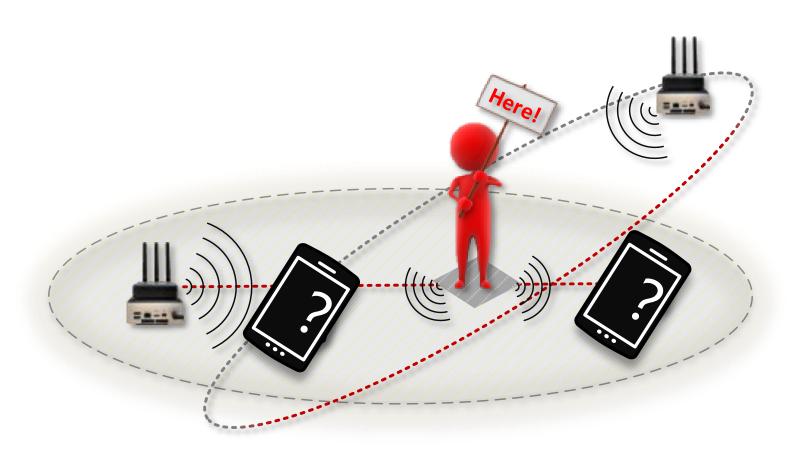
Subcarriers greatly affected by multipath

### **CSI Pre-Processing Method Verification**

### Pre-processed CSI measurements match the theoretical values



### Challenge: locations of some transceivers (e.g. mobiles) are unknown!



### **Unknown Transceiver Locations**

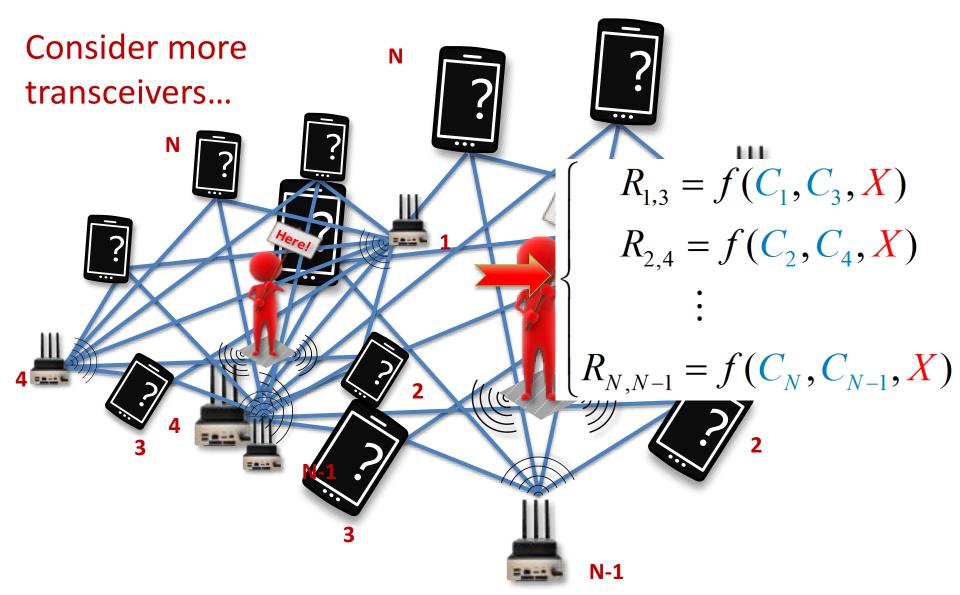
$$R_{1,3} = f(C_1, C_3, X)$$

$$R_{2,4} = f(C_2, C_4, X)$$

$$C_4$$

$$C_3$$

### **Unknown Transceiver Locations**



### **Key Observation**

$$R_{1,3} = f(C_1, C_3, X)$$
 $R_{2,4} = f(C_2, C_4, X)$ 
 $\vdots$ 
 $R_{N,N-1} = f(C_N, C_{N-1}, X)$ 

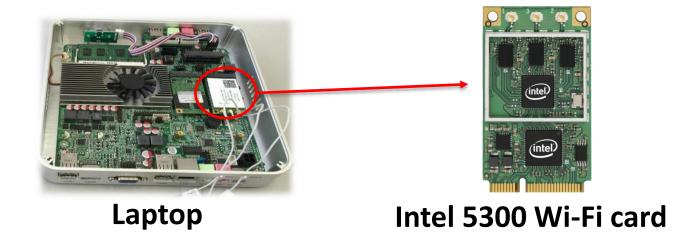
N(N-1)/2 equations >> 2(N+1) unknowns

Given enough number of transceivers, all unknown locations can be estimated!

### Implementation & Evaluation

### **Implementation**

 Implemented: 11 laptops, which are equipped with Intel 5300 Wi-Fi card.



- 4 laptops act as APs and the rest are clients.
- APs and one client have known locations.
- Objective: device-free passively localize a target.

### **Compared Approaches**

	Signal Feature	Methodology	Requirement
Pilot [1]	CSI	Kernel-based MAP	Fingerprinting
RASS [2]	RSS	SVM	Fingerprinting
RTI [3]	RSS	Maximum attenuation	Knowledge of deployment

<sup>[1]</sup> J. Xiao, K. Wu, Y. Yi, L. Wang, and L. M. Ni. Pilot: Passive device-free indoor localization using channel state information. In Proc. IEEE International Conference on Distributed Computing Systems (ICDCS), pages 236–245, 2013.

<sup>[2]</sup> D. Zhang, Y. Liu, X. Guo, and L. M. Ni. RASS: A real-time, accurate, and scalable system for tracking transceiver-free objects. IEEE Trans. on Parallel and Distributed Systems (**TPDS**), 24(5):996–1008, 2013.

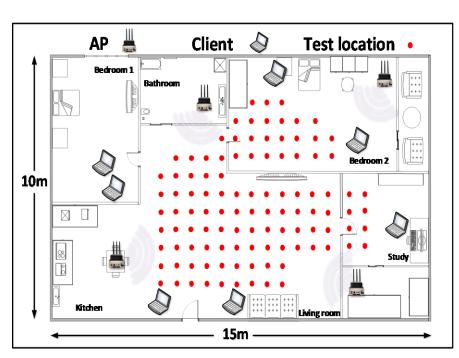
<sup>[3]</sup> J. Wilson and N. Patwari. See-through walls: Motion tracking using variance-based radio tomography networks. IEEE Trans. on Mobile Computing (**TMC**), 10(5):612–621, 2011.

### **Evaluation in Medium Multipath**

Home environment: medium multipath



Home environment



Testbed floorplan

### **Evaluation in LoS & NLoS**

 Empty classroom and Library environments, corresponding to *low* and *high* multipath



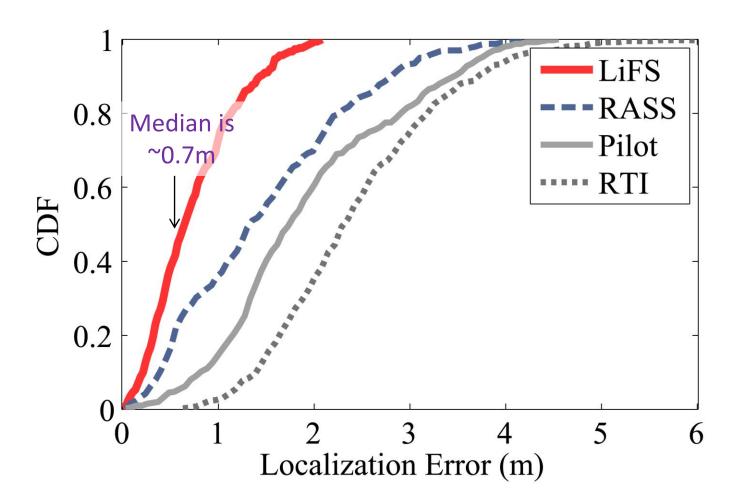
Strong LoS (9 m x 12 m)



Strong NLoS (7 m x 12 m)

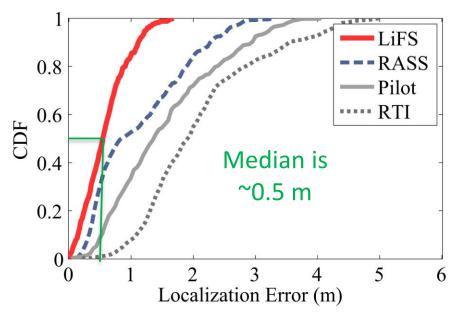
### **Evaluation in Medium Multipath**

Home environment: medium multipath

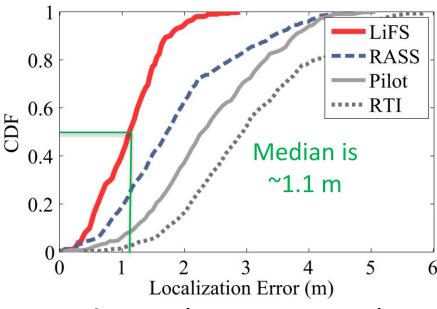


### **Evaluation in LoS & NLoS**

 Empty classroom and Library environments, corresponding to *low* and *high* multipath



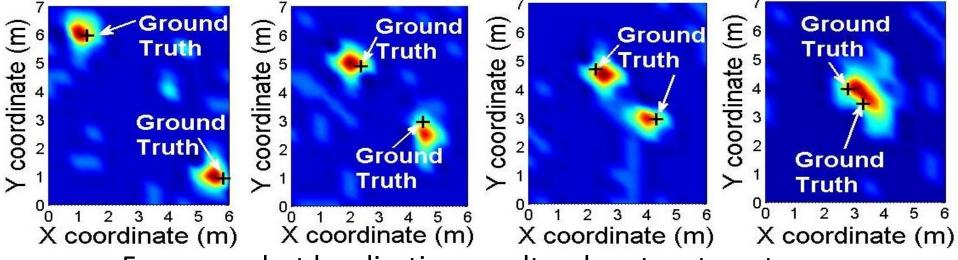
Classroom (Strong LoS)



Library (Strong NLoS)

### **Two-Target Localization**

 Intuition: a target is not able to affect all the wireless links simultaneously.



Four snapshot localization results when two targets are 5.4 m, 3 m, 1.8 m and 0.6 m apart.

LiFS can localize each individual target accurately when they are not too close!

### Conclusions

- Device-free localization is important for many applications.
- Even in a rich multipath environment, it's possible to identify "clean" subcarriers for model-based localization.
- Extensive experiments demonstrate the effectiveness of LiFS.