ECE 598HH: Advanced Wireless Networks and Sensing Systems

Lecture 12: Wireless Sensing Part 3

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*Slides Courtesy of Prof. Fadel Adib

Interest in Sensing the Human Body

Heart Rate

Breathing

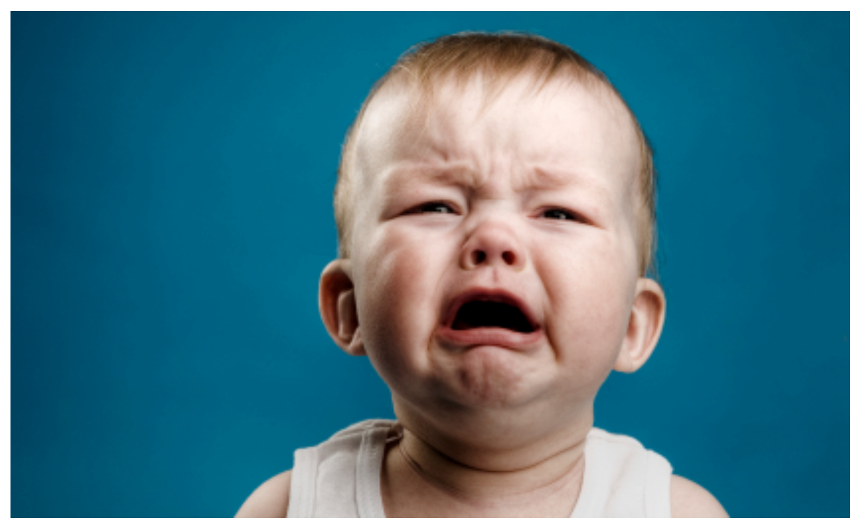
Locations

Gestures

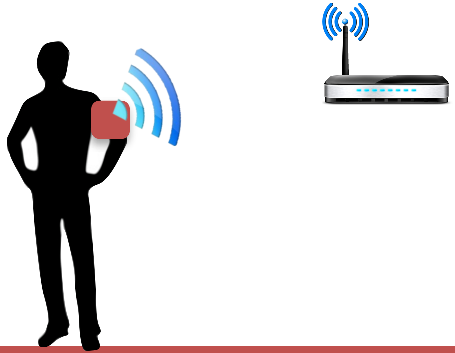
Heart Rate Breathing Locations Gestures

On-body sensors can be cumbersome

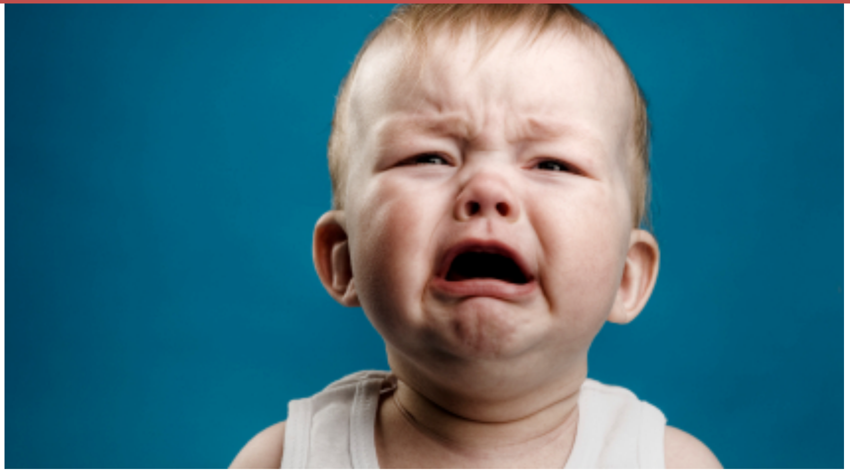
Not suitable for elderly & babies

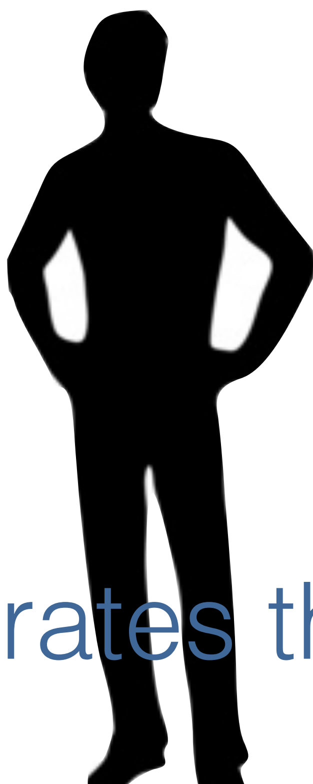


Heart Rate   Breathing   Locations   Gestures



Imagine enabling these applications without sensors on the human body





- Location
- Vital Signs
- Imaging

Operates through occlusions

Last Lecture

WiVi: Sensing humans through walls with WiFi

- MIMO Nulling
- Inverse SAR

WiTrack 1.0 & 2.0: Localizing & Tracking through walls

- FMCW
- Background Subtraction
- Dynamic Multipath
- Multi-Shift FMCW
- Successive Silhouette Cancellation
- Multi-Resolution Subtraction Window

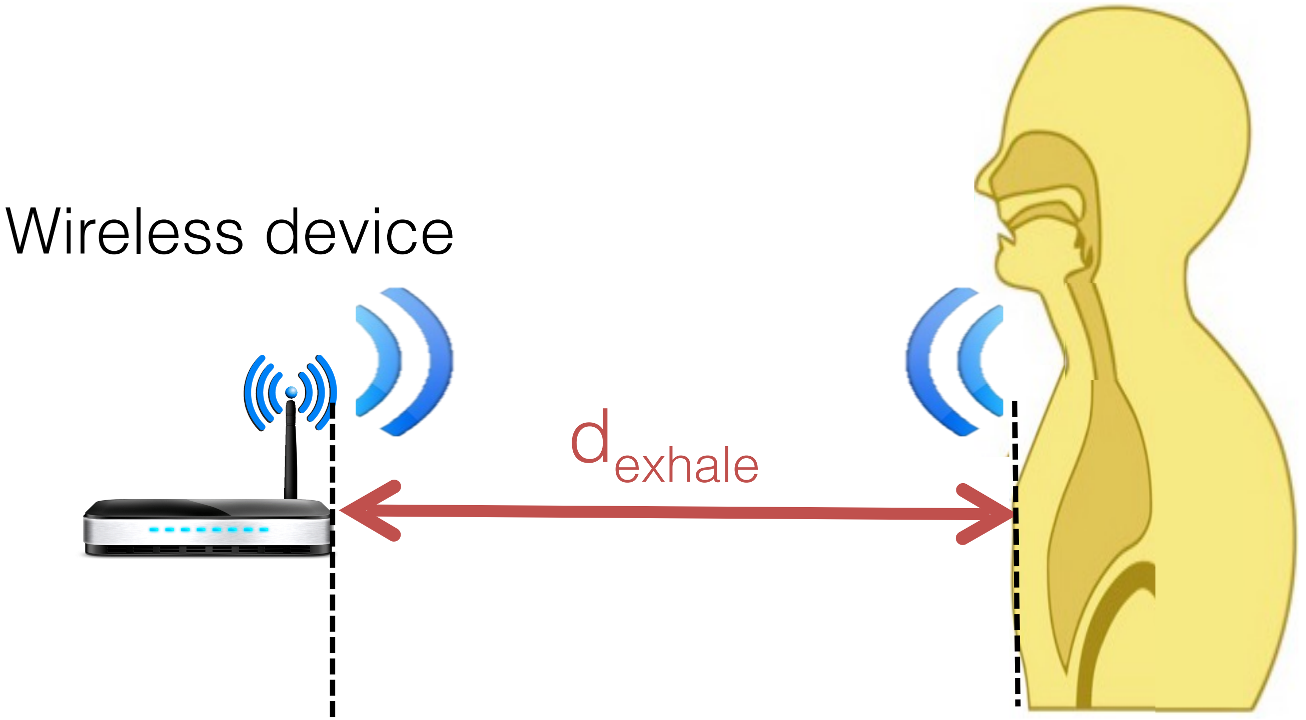
This Lecture

Vital Radio: Extracting vital signs

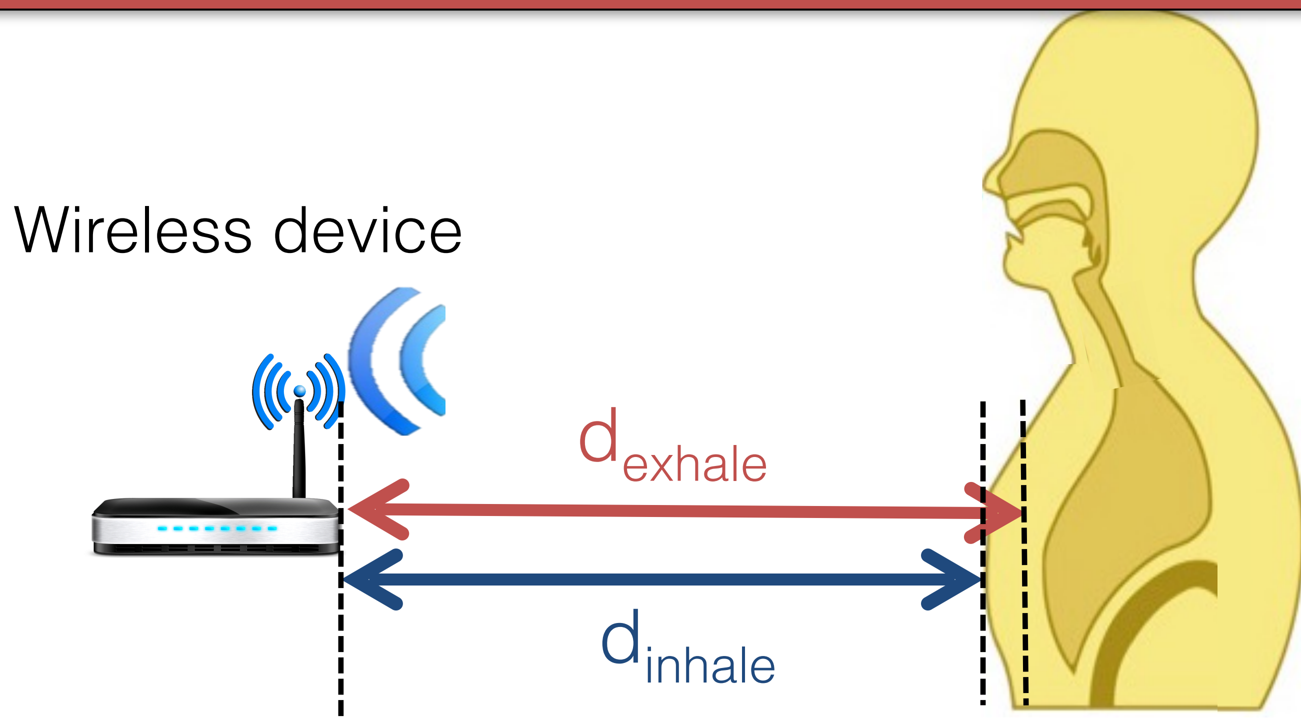
- Breathing Rate
- Heart Rate

RF-Capture: Capturing human figure through wall.

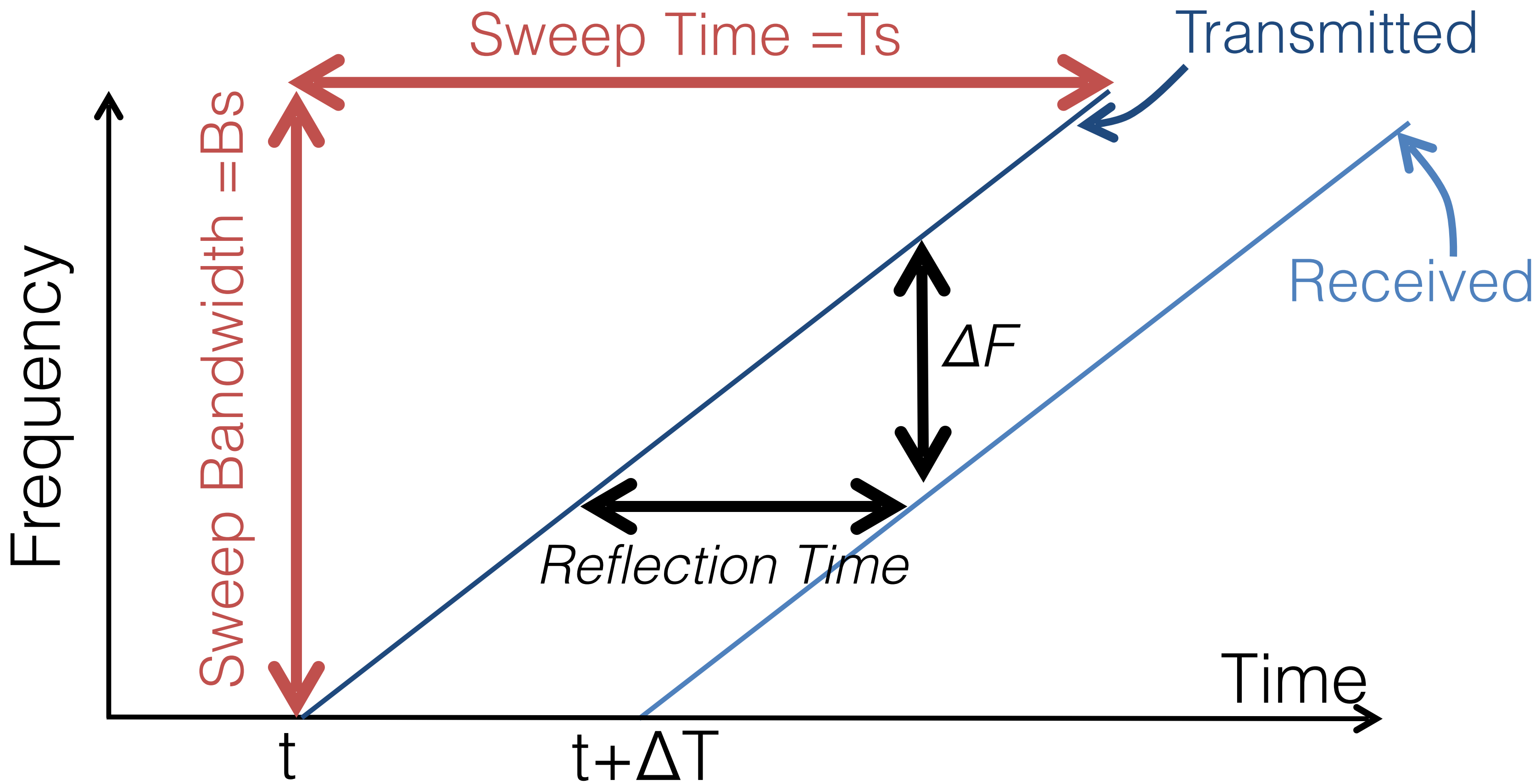
Vital Radio: Use wireless reflections off the human body to monitor breathing and heart rate


Wireless device
d_{exhale}

Problem: Localization accuracy is only 12cm and cannot capture vital signs



FMCW: Measure time by measuring frequency



Slope = k = Bs/Ts

Reflection Time =ΔF/k

FMCW

- FMCW Transmitted Signal:

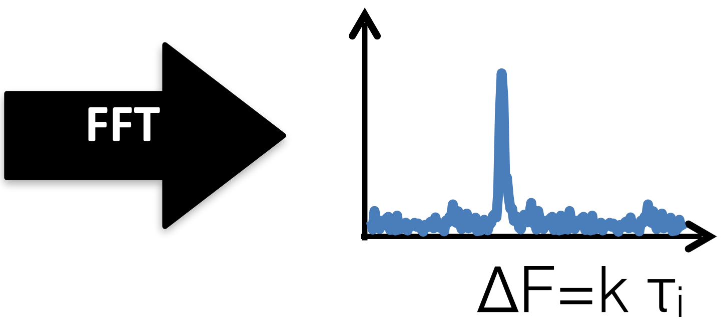
$$x(t) = e^{j2\pi(\frac{k}{2}\ t^2 + f_0 t)}$$

- FMCW Received Signal:

$$y(t) = \sum_i A_i e^{j2\pi(\frac{k}{2}\ (t-\tau_i)^2 + f_0(t-\tau_i))}$$

- FMCW after downconversion:

$$y_b(t) = \sum_i A_i e^{j2\pi(k\tau_i t + f_0 \tau_i)}$$



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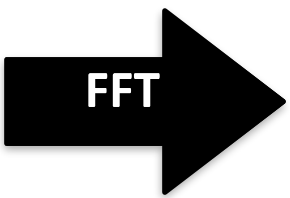

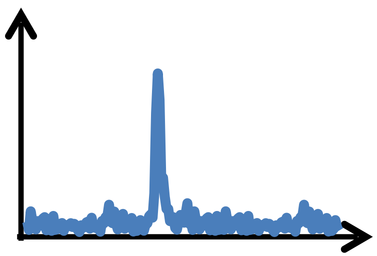

$\Delta F = k\ \tau_i$

- Sampling Rate = B

$\Delta F < B \longrightarrow \tau_{max} = B/k = BxTs/Bs \longrightarrow d_{max} = cxBxTs/2Bs$

FMCW

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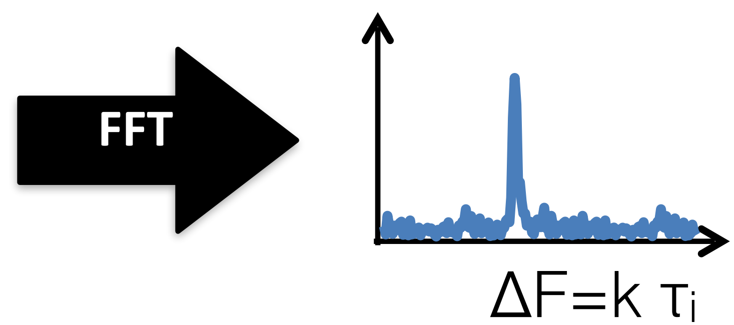

- Sampling Rate = B

$\Delta F < B \longrightarrow \tau_{max} = B/k = BxTs/Bs \longrightarrow d_{max} = cxBxTs/2Bs$

- Sampling Window = Ts

$dF > 1/Ts \longrightarrow \tau_{min} = 1/(kxTs) = 1/Bs \longrightarrow d_{min} = c/2Bs$

FMCW

- FMCW Transmitted Signal:

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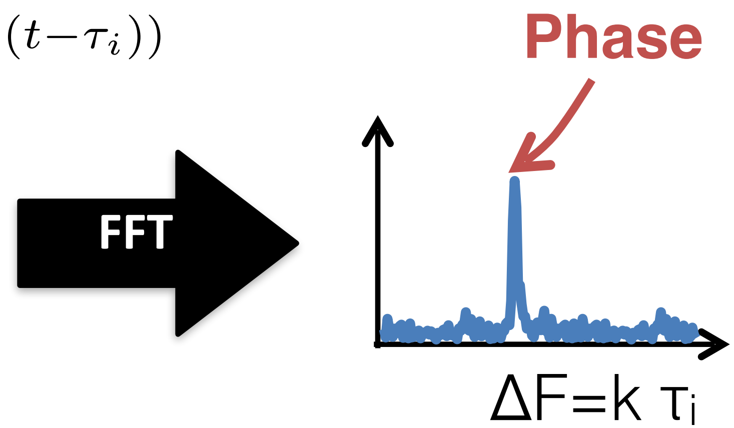


- Phase of peak = $f_0\tau_i$
  - Phase wraps around 2pi
  - Use peak position $\Delta F = k\,\tau_i$ for course estimate of τ_i
  - Use peak phase $f_0\tau_i$ for fine estimate of τ_i

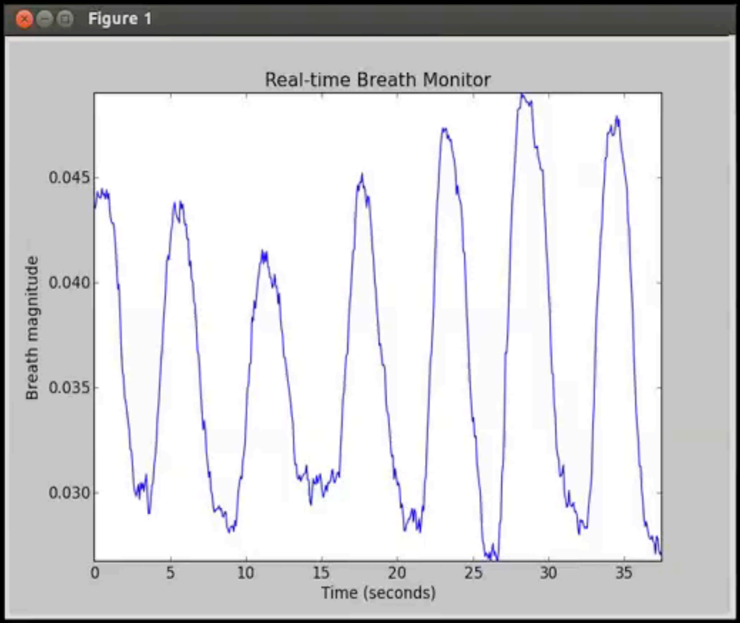
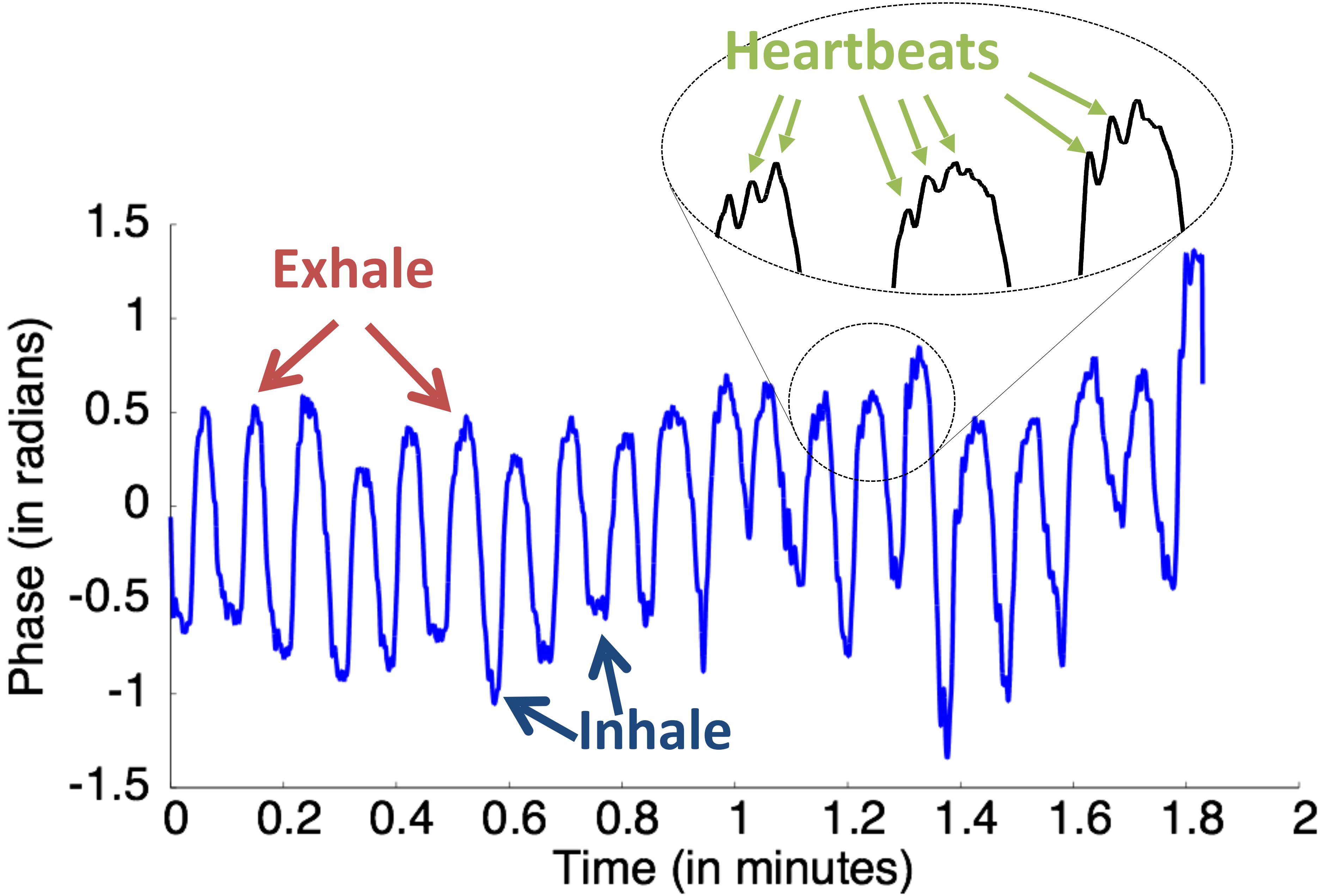


Figure 1
Real-time Breath Monitor
Breath magnitude
0.045
0.040
0.035
0.030
0
5
10
15
20
25
30
35
Time (seconds)

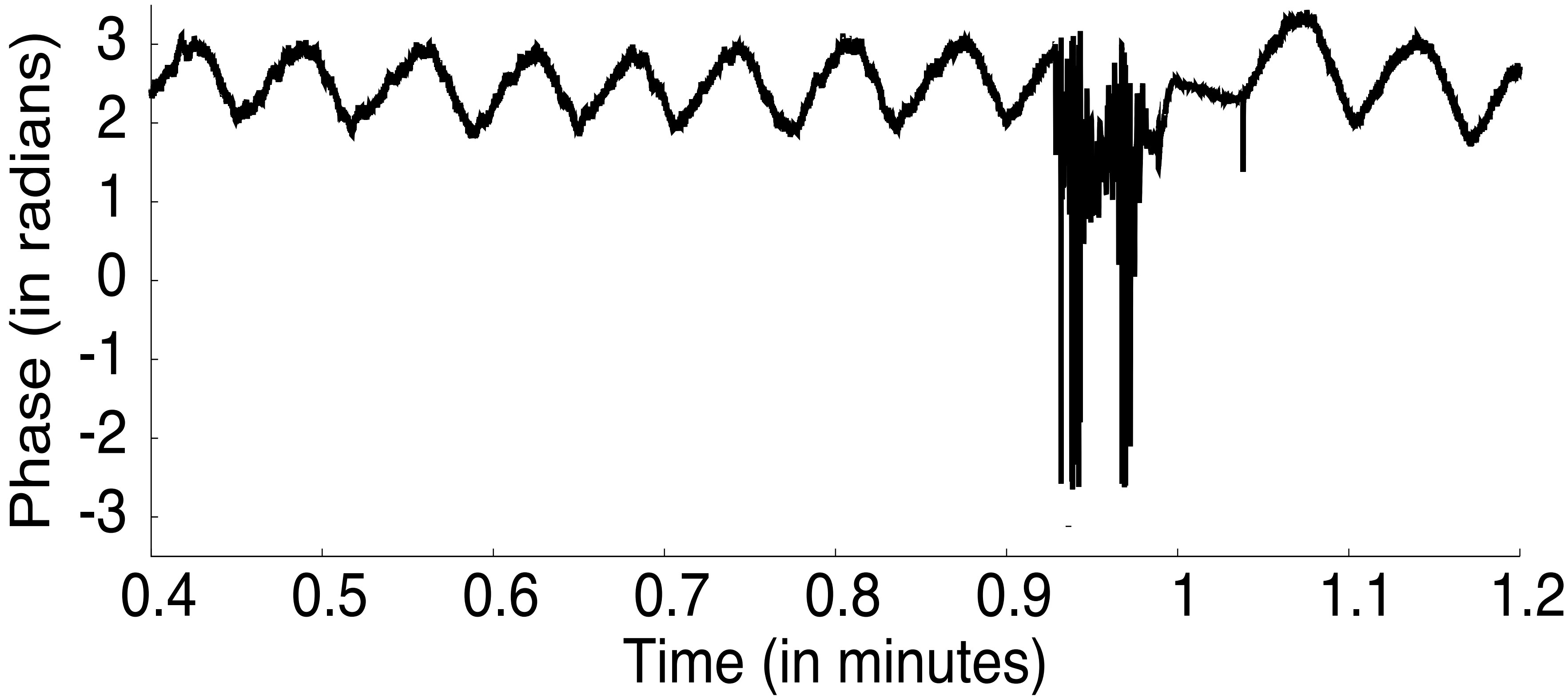
Let's zoom in on these signals



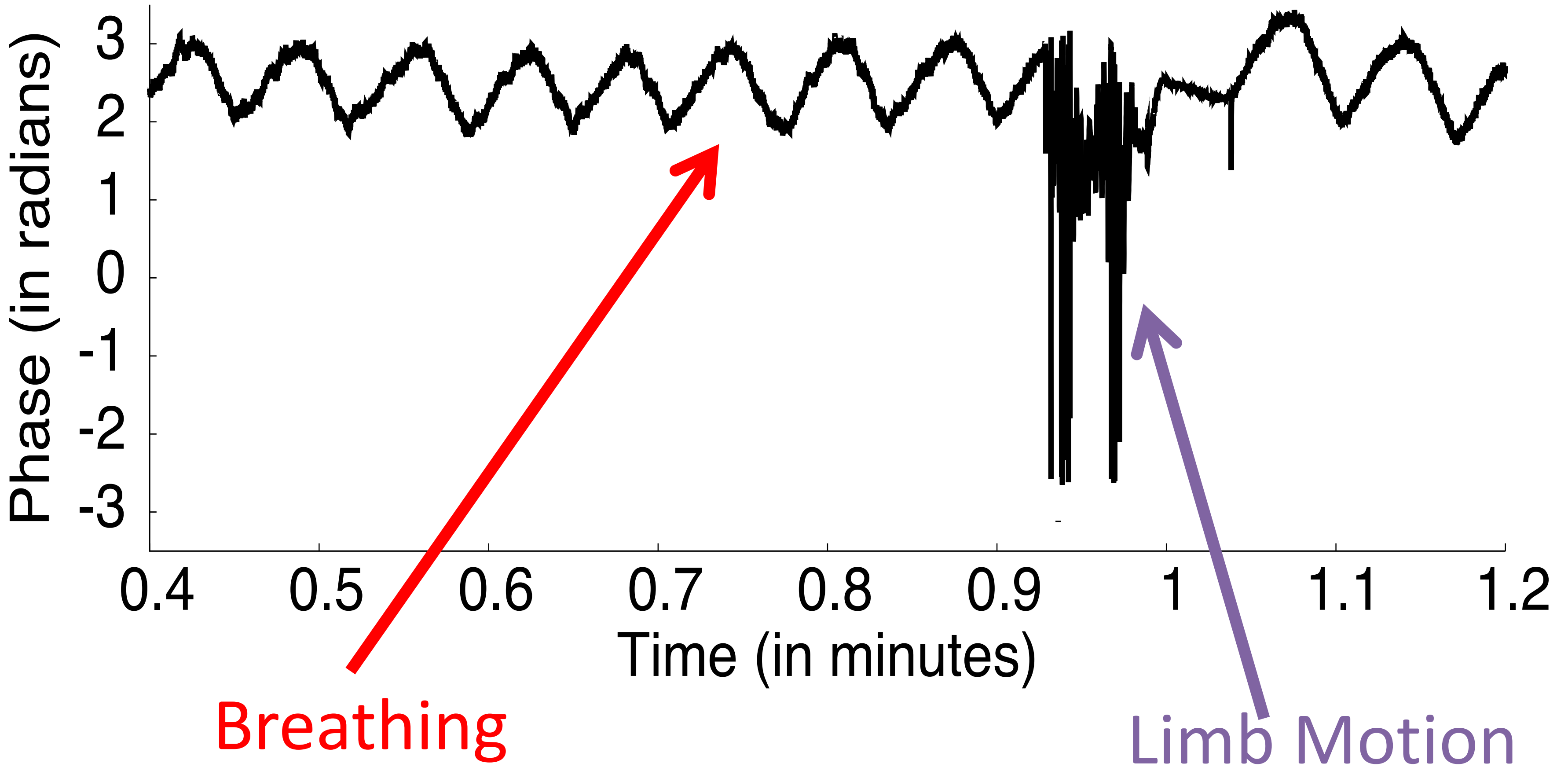
Heartbeats
Exhale
Inhale
1.5
1
0.5
0
-0.5
-1
-1.5
Phase (in radians)
0
0.2
0.4
0.6
0.8
1
1.2
1.4
1.6
1.8
2
Time (in minutes)

What happens when a person moves his limb?

What happens when a person moves his limb?



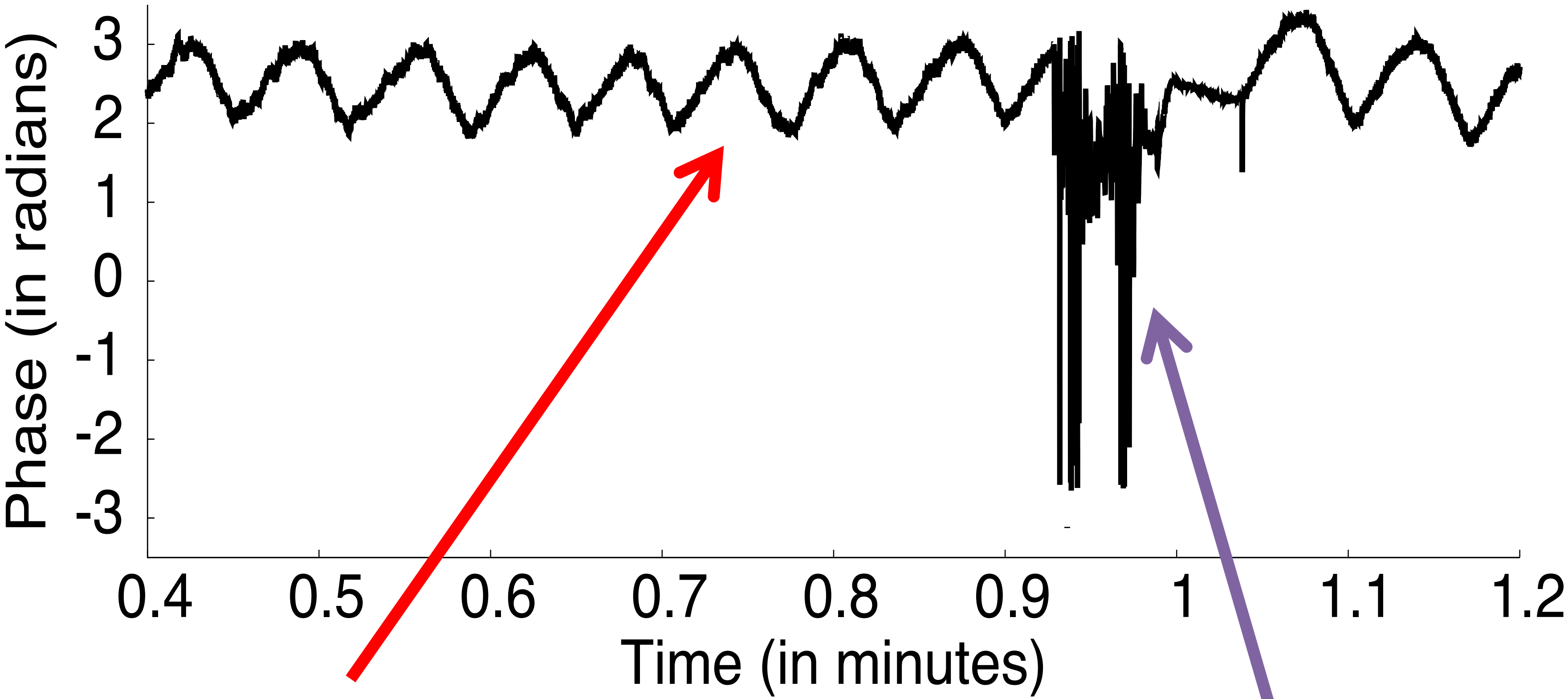
What happens when a person moves his limb?



Periodic

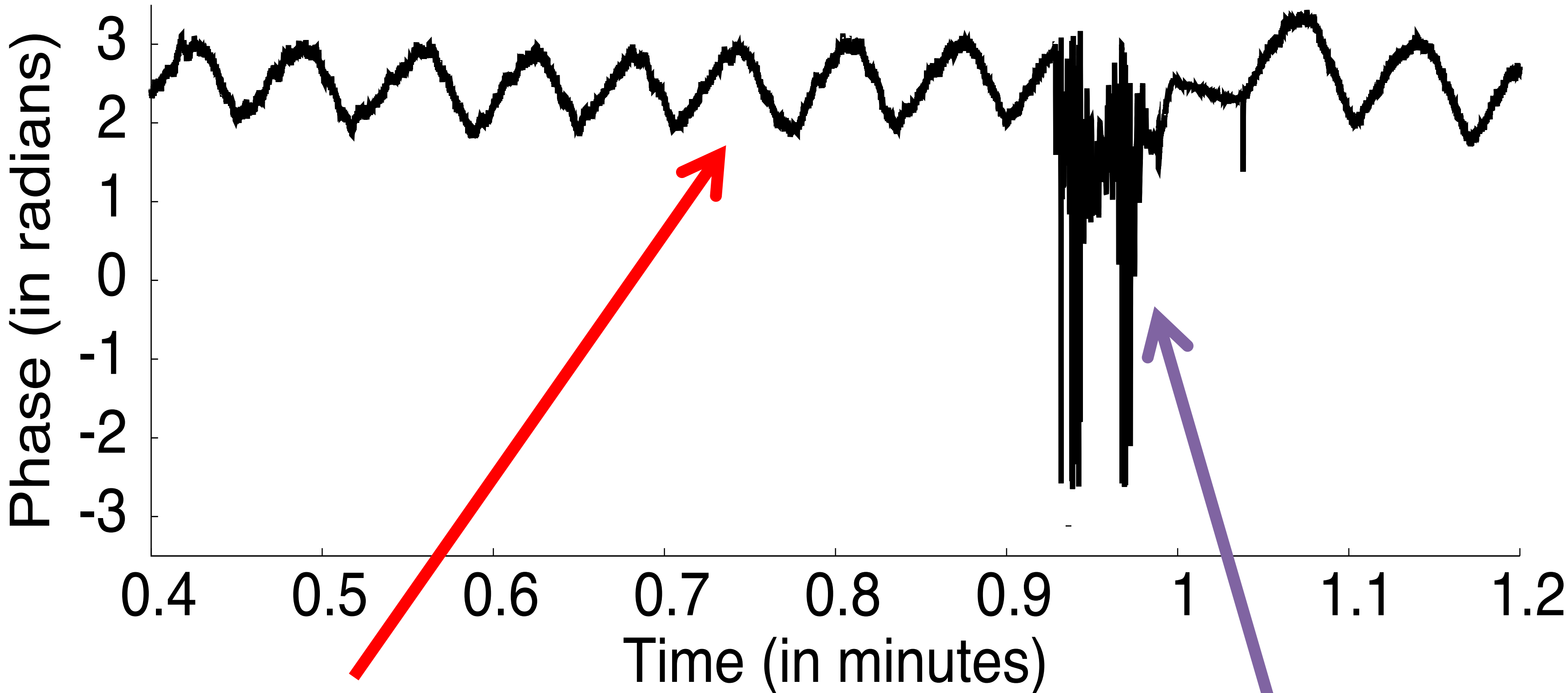
Not periodic

What happens when a person moves his limb?



Use periodicity test to eliminate variations that are not due to breathing/heartbeats

What happens when a person moves his limb?

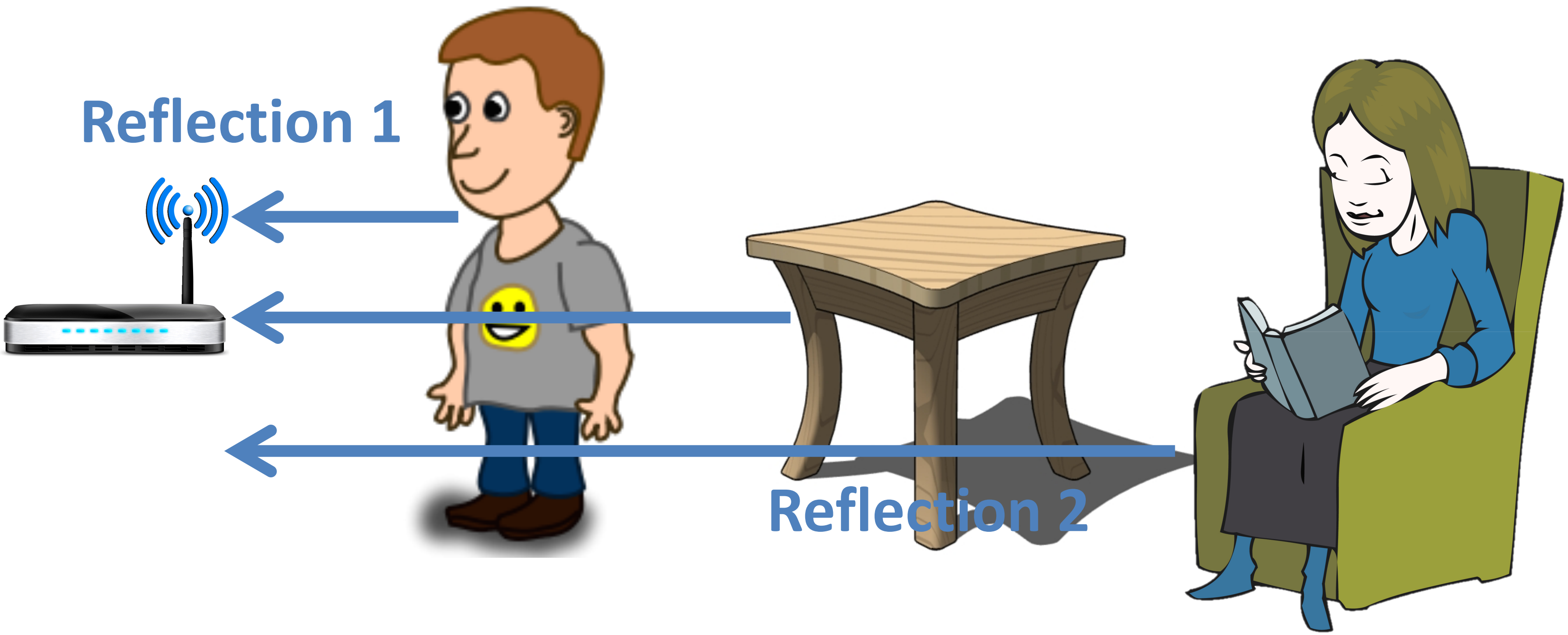


Band-pass filter the cleaned signals to extract breathing and heart rate

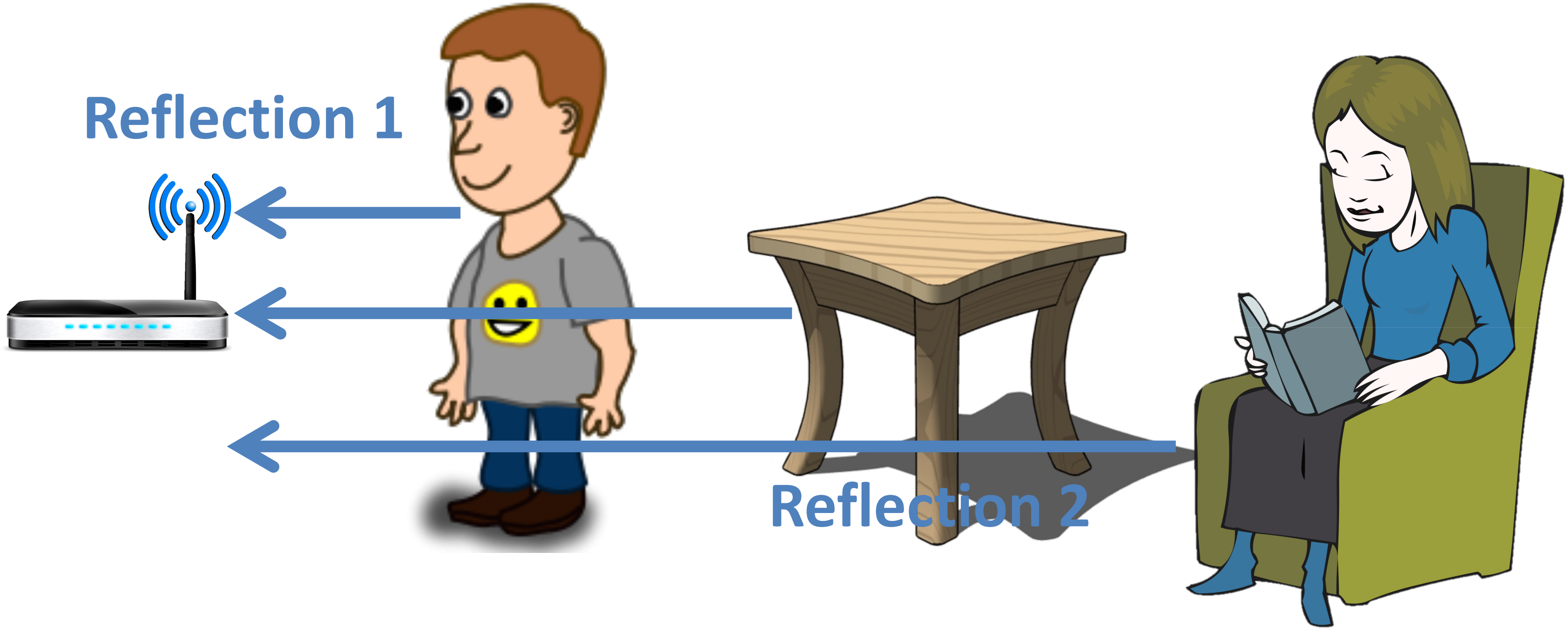
What happens with multiple users in the environment?

Reflections from different objects collide

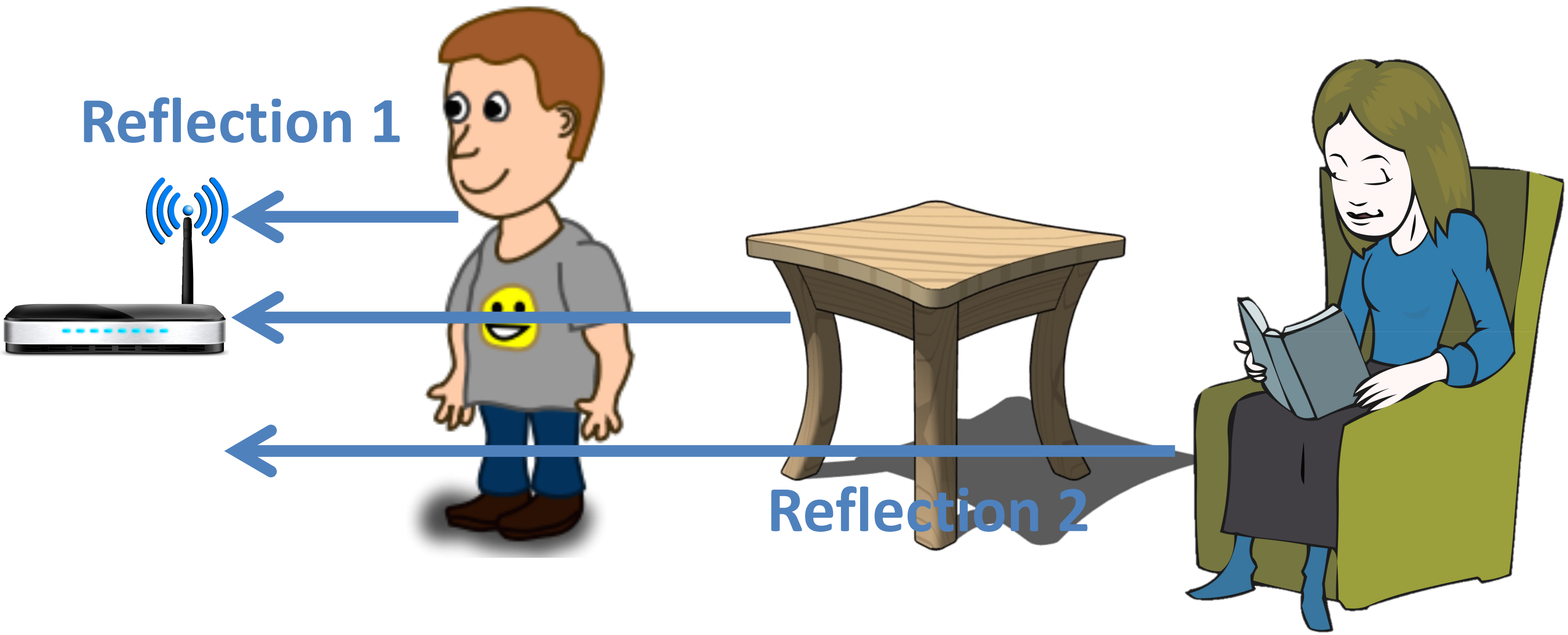
<u>Problem:</u> Phase becomes meaningless!



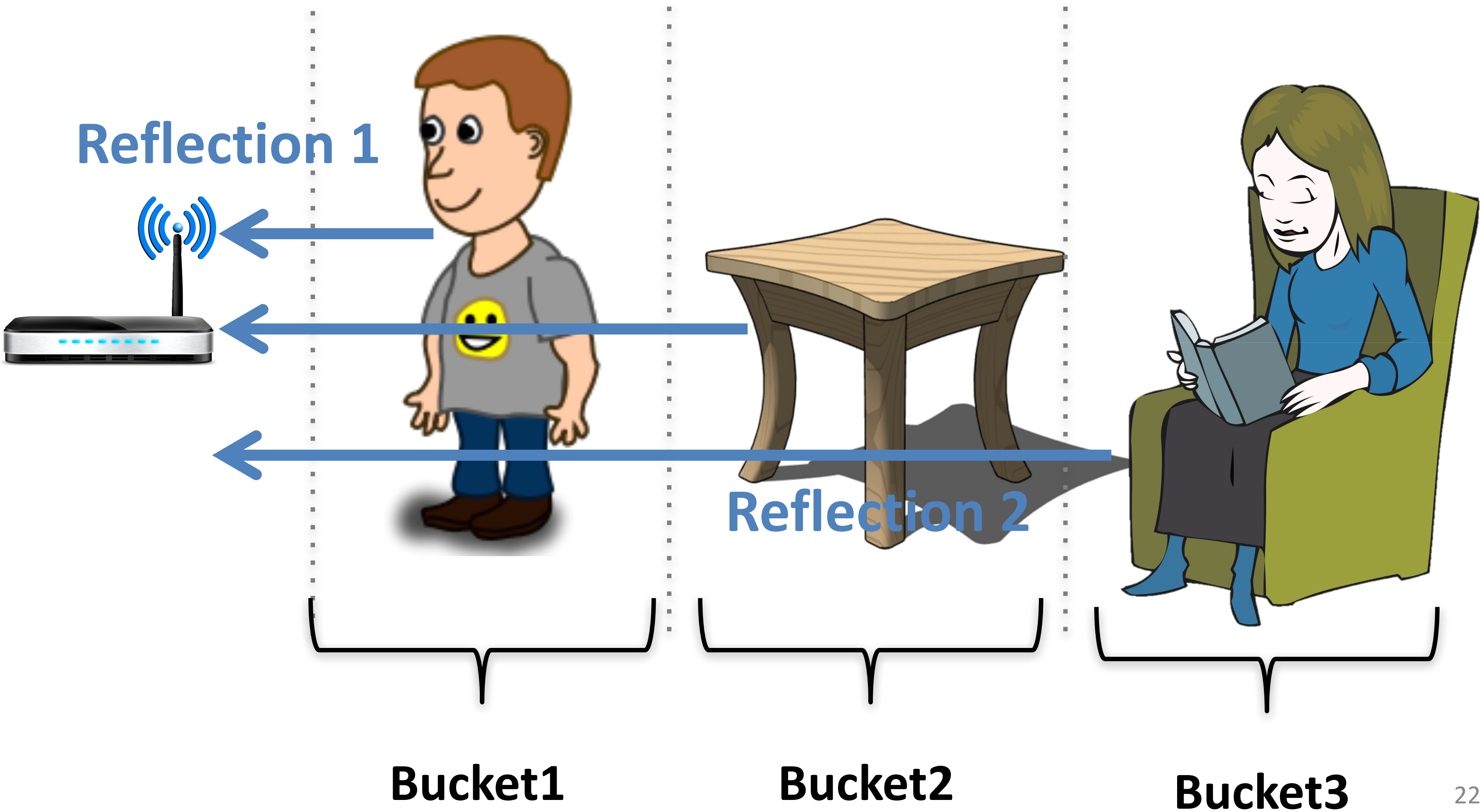
Idea: Wireless localization can be used to locate various devices



Solution: Use wireless localization as a filter to isolate reflections from different positions



Solution: Use wireless localization as a filter to isolate reflections from different positions



Putting It Together

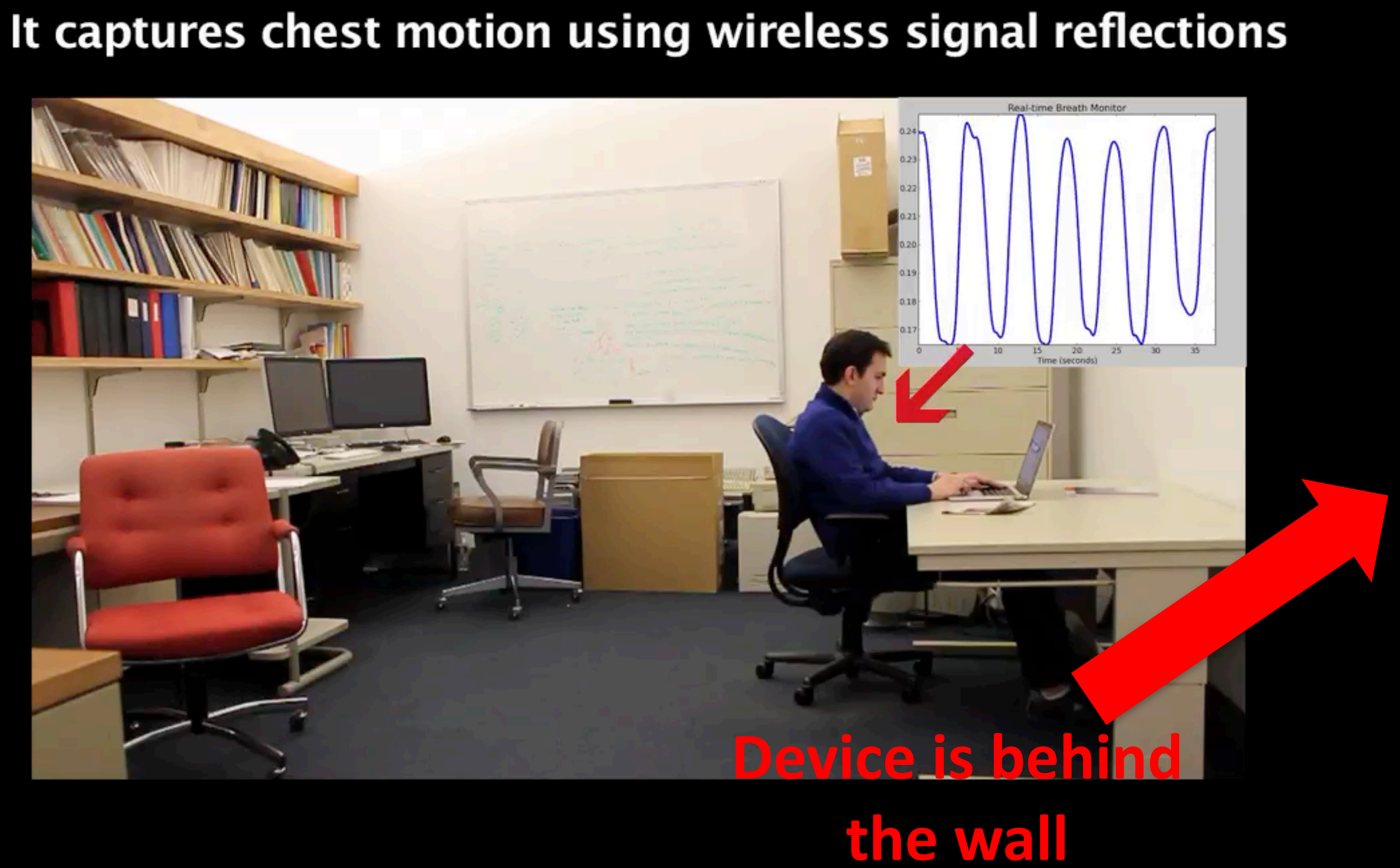
Step 1: Transmit a wireless signal and capture its reflection

Step 2: Isolate reflections from different objects based on their positions

Step 3: Zoom in on each object’s reflection to obtain phase variations due to vital signs

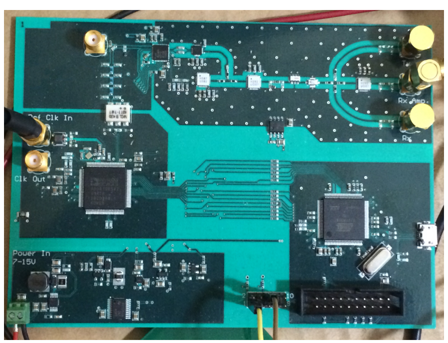
Step 4: Use frequency analysis to separate breathing and heart rate signals

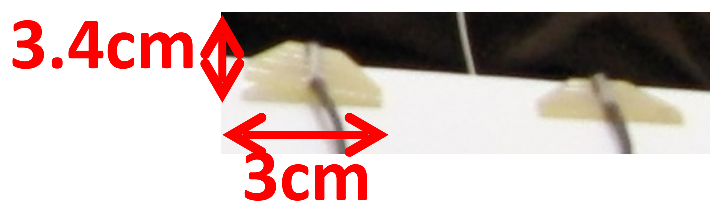
Through-wall breath monitoring of multiple users



Vital-Radio Implementation

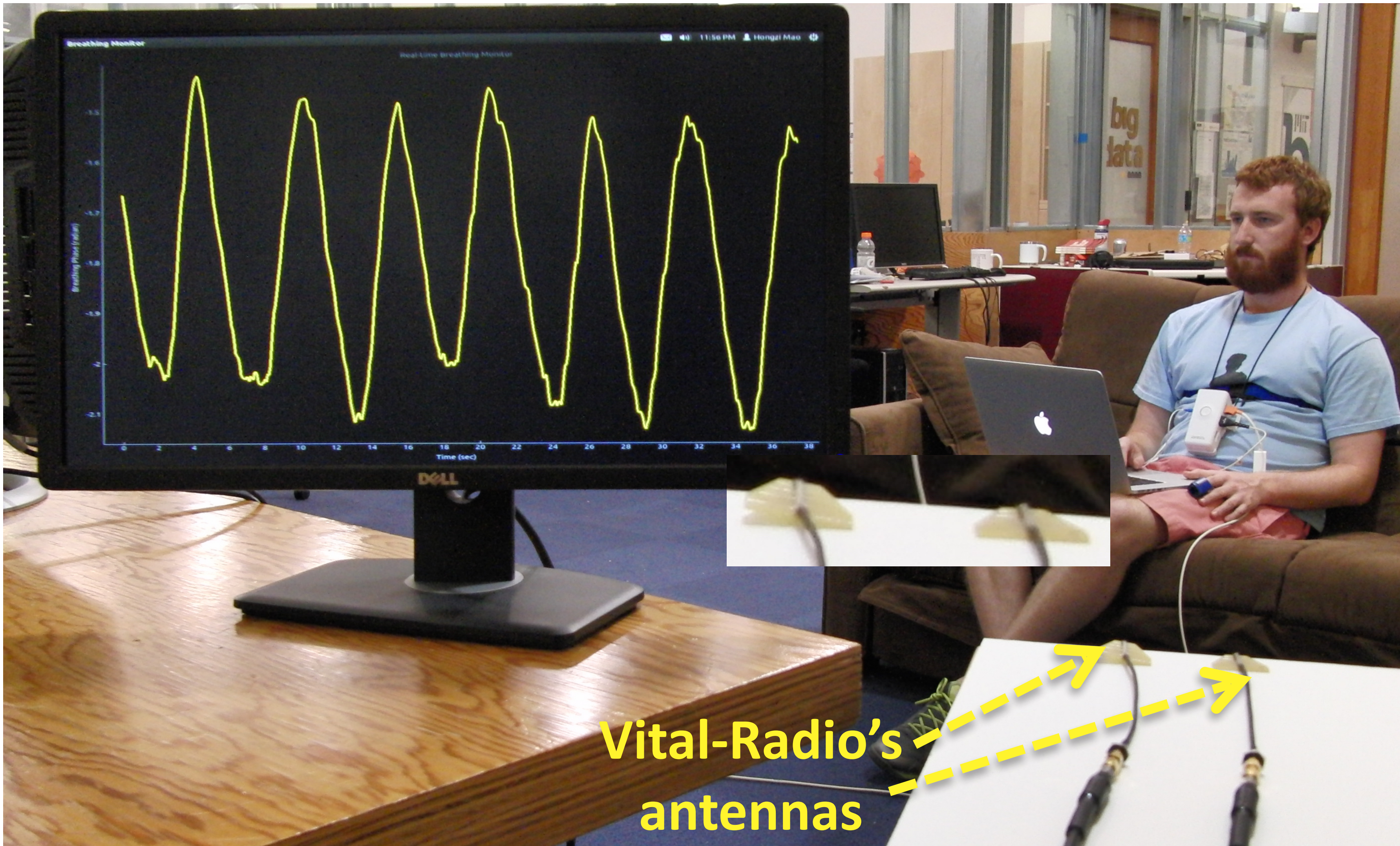
- Wireless positioning device to transmits and receives wireless signals
  - 10,000x lower power than cellphones
  - 1 transmit & 1 receive antenna





- Signal is analyzed in software to extract vital signs

Vital-Radio Implementation



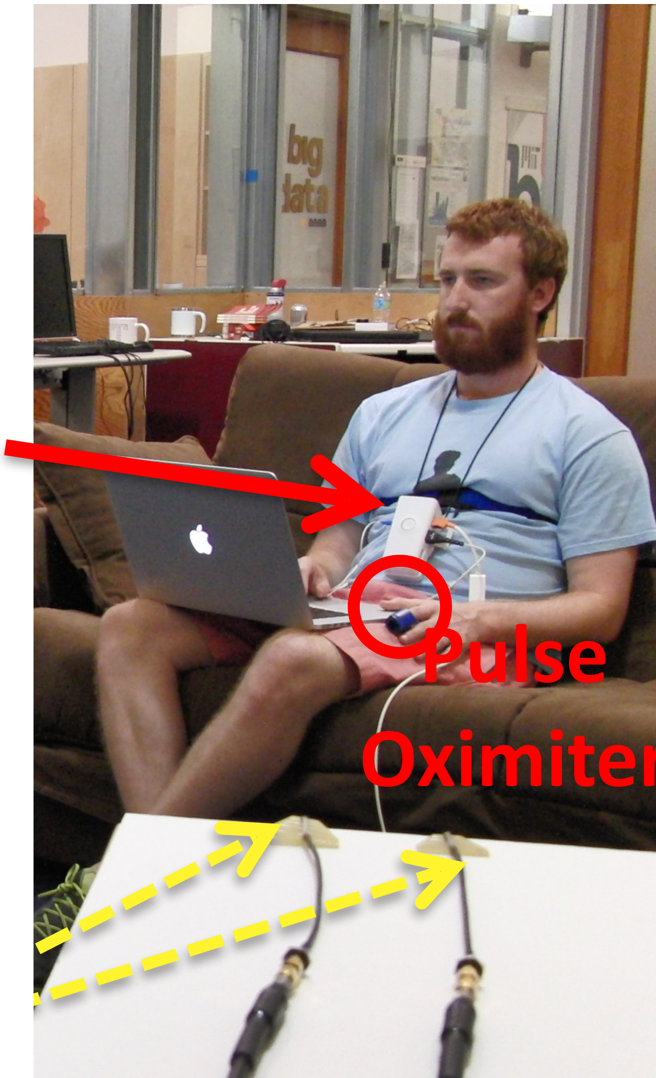
Vital-Radio Evaluation

Baseline:

- FDA-approved breathing and heart rate monitor

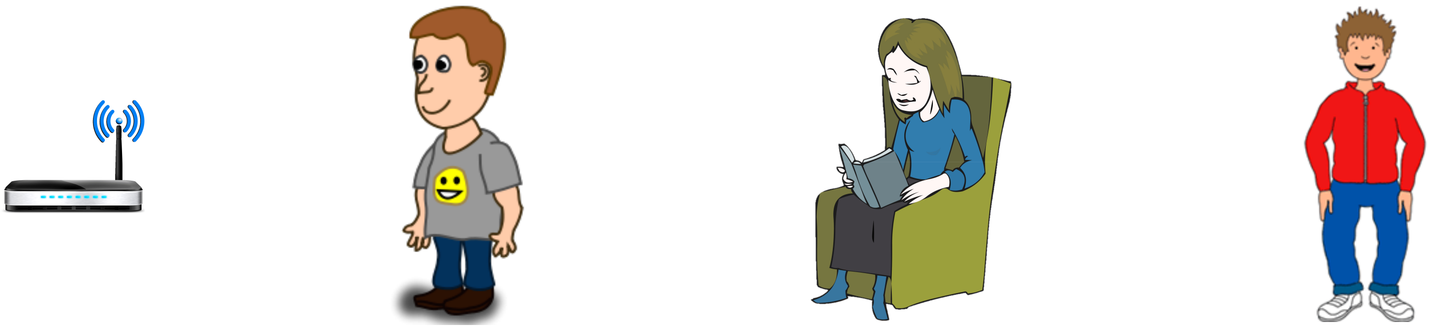
Experiments:

- 200 experiments
- 14 participants
- 1 million measurements

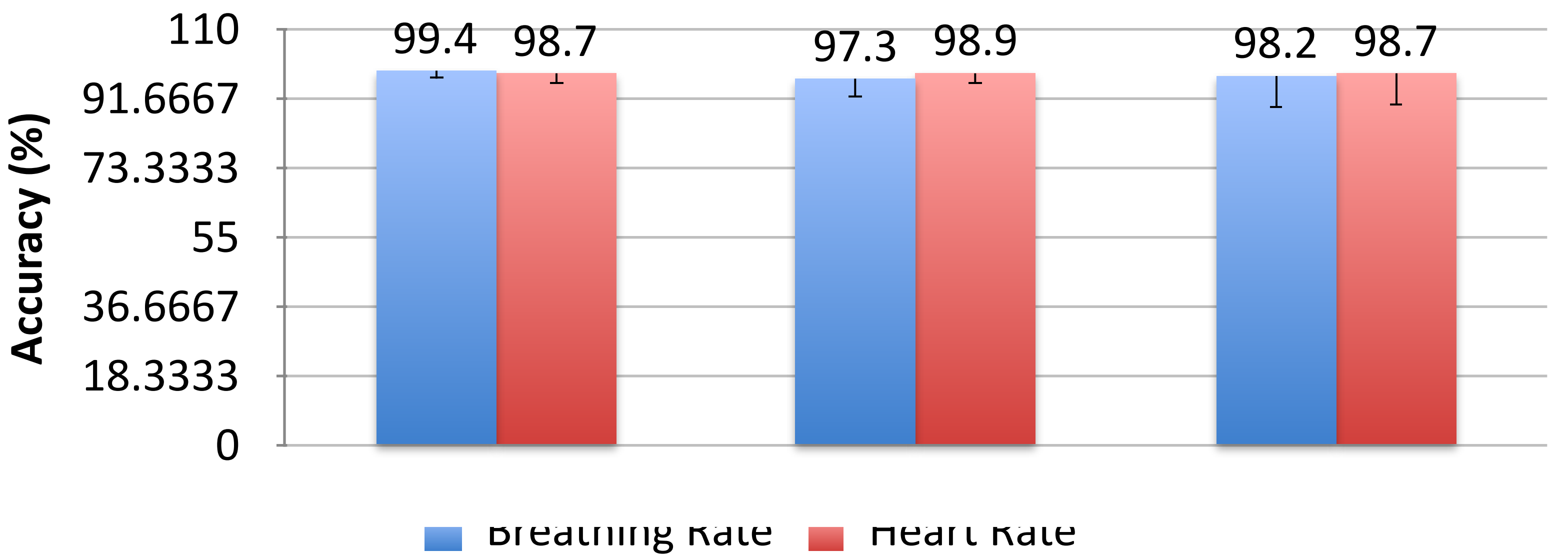


Accuracy for Multi-User Scenario

Multiple users sit at different distances

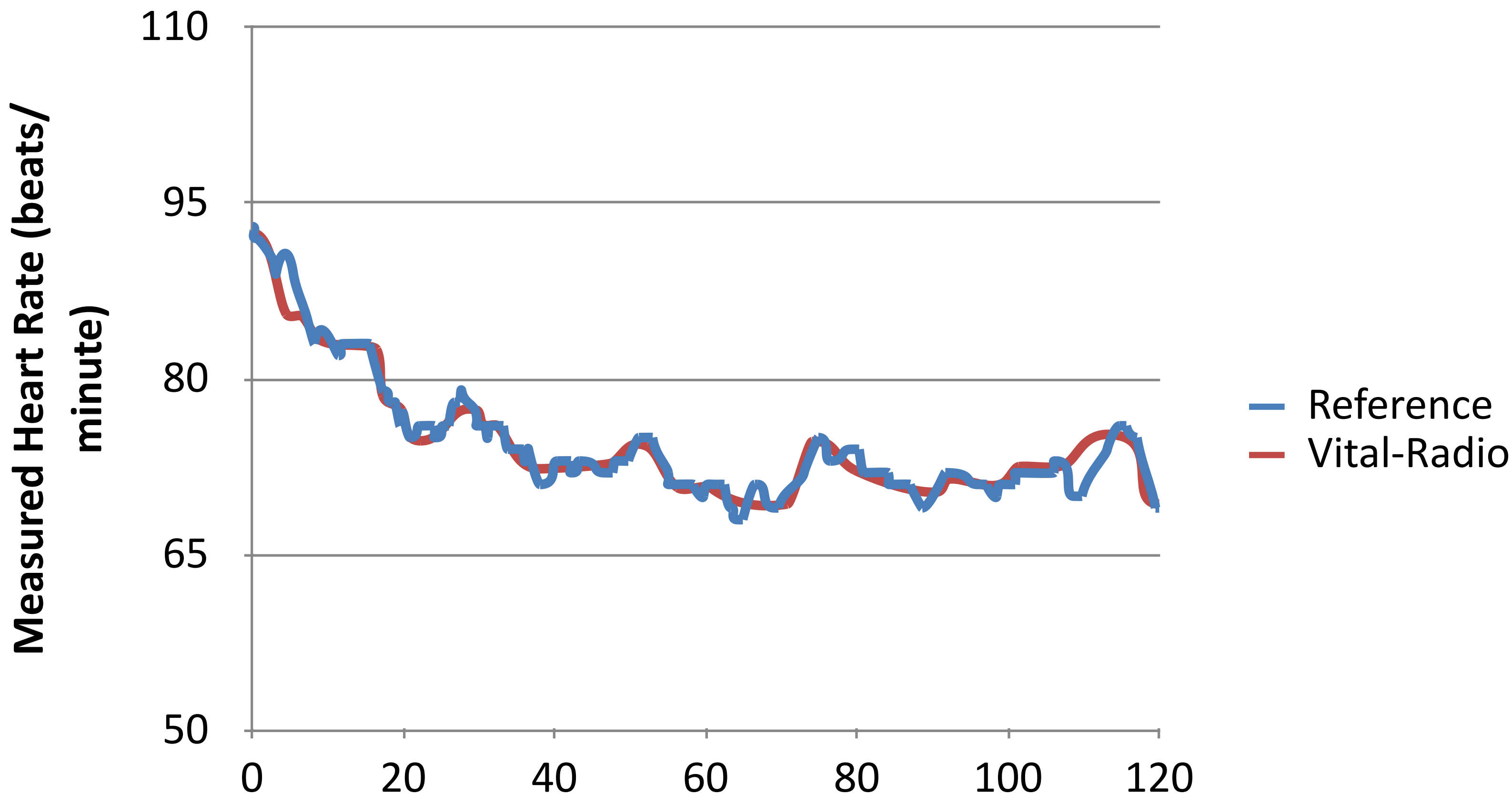


Nearest (at 2m) Middle (at 4m) Furthest (at 6m)



Accuracy for Tracking Heart Rate

Measure user's heart rate after exercising



Vital-Radio accurately tracks changes in vital signs

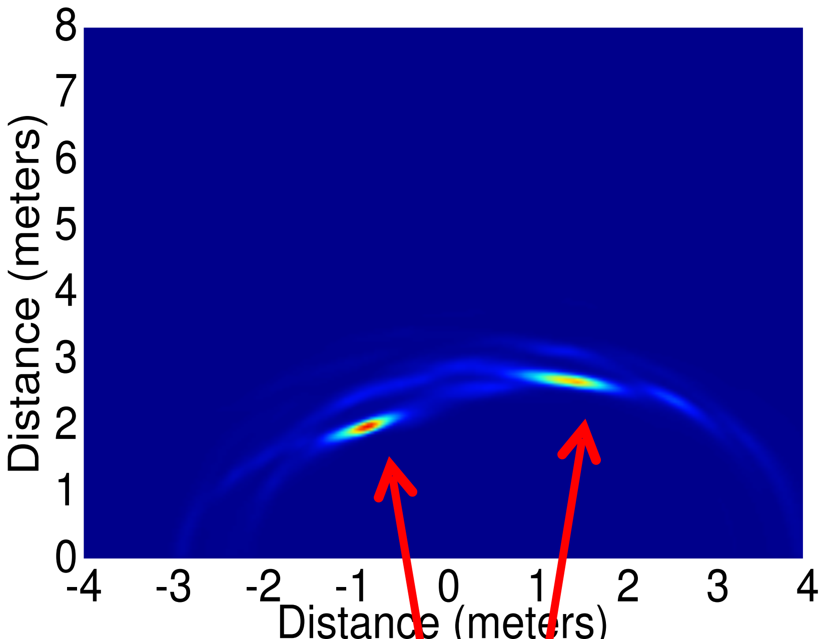
Vital-Radio Limitations

- Minimum separation between users: 1-2m
- Monitoring range: 8m
- Collects measurements when users are quasi-static

Baby Monitoring

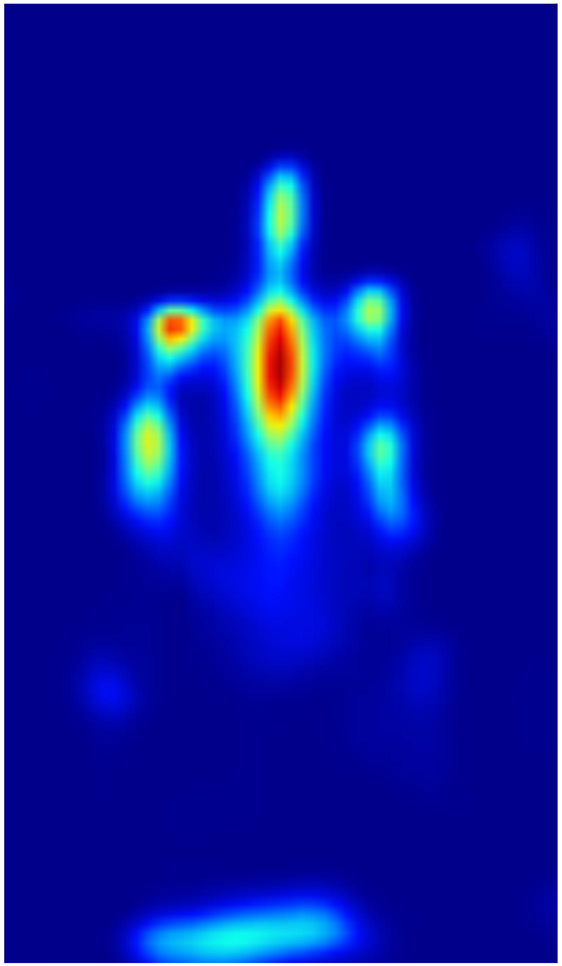
RF Imaging

People are points

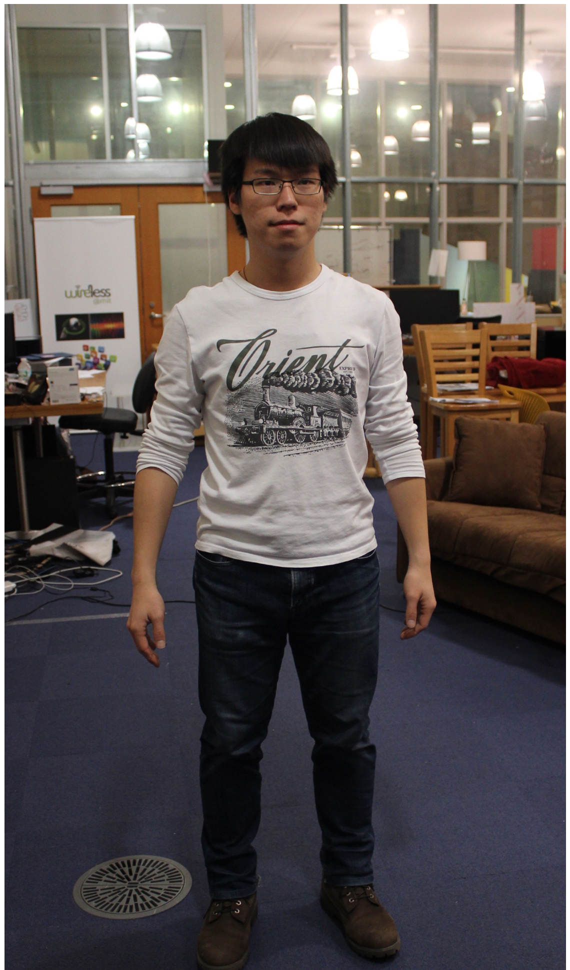


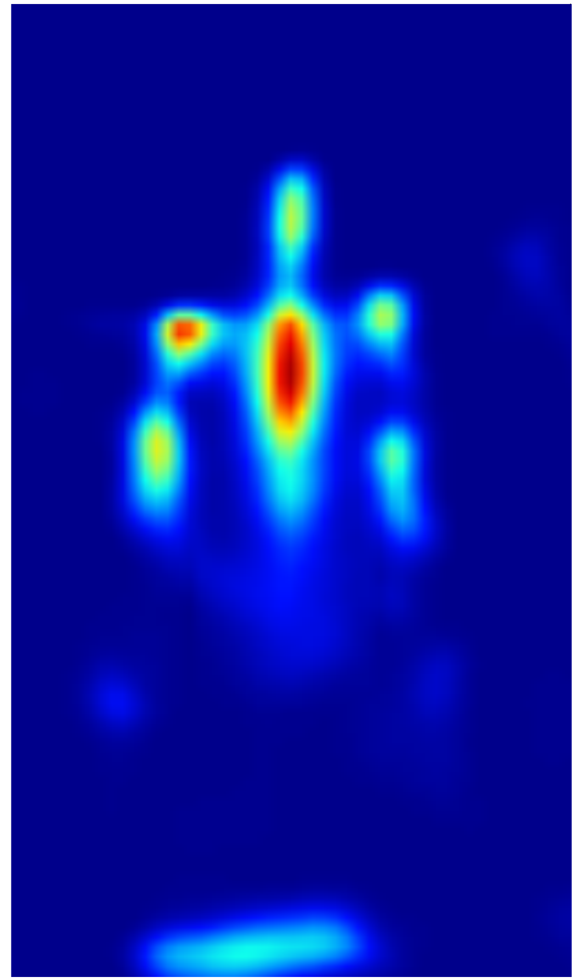
Localize the two users

Want a silhouette

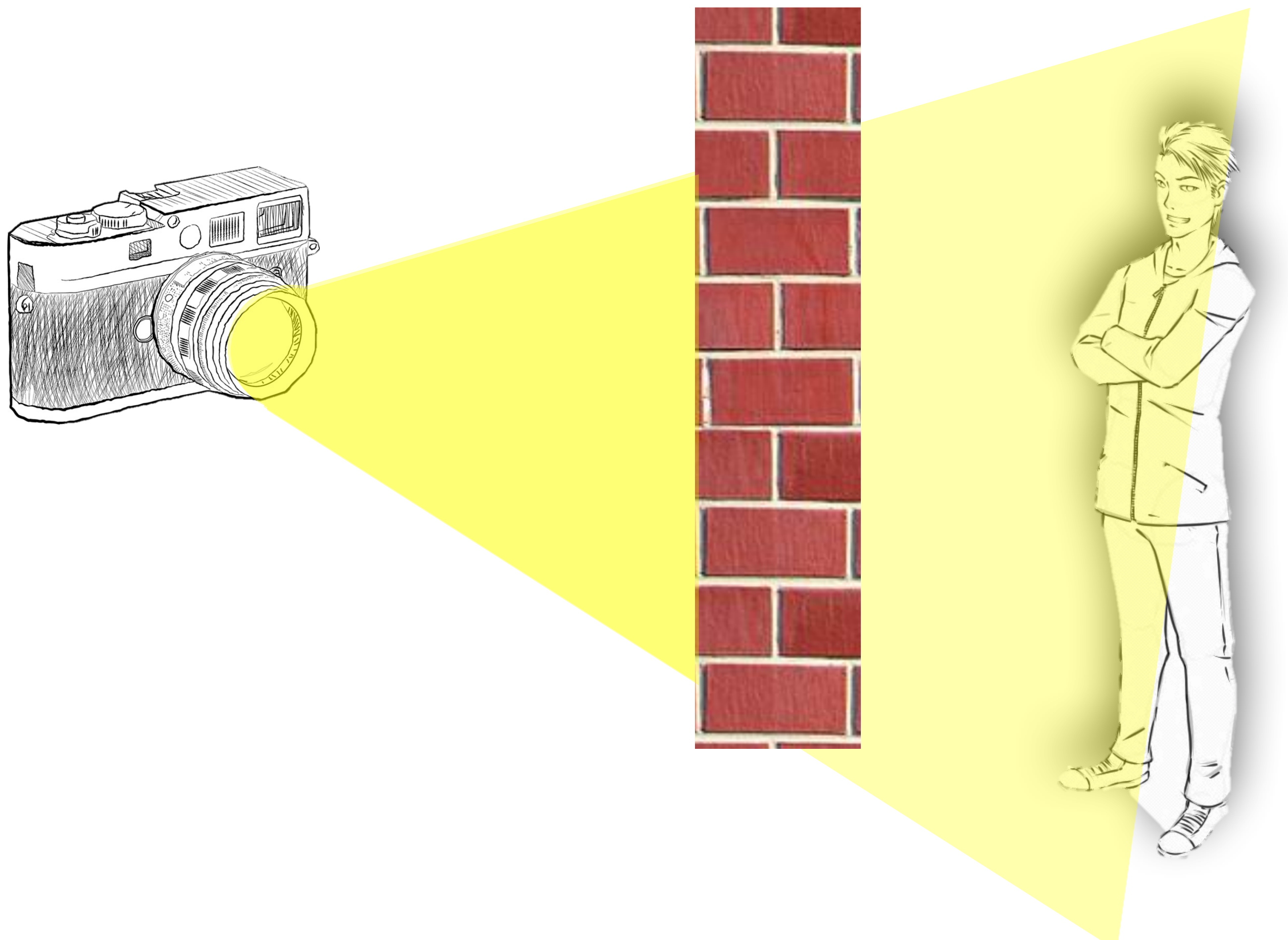


Capturing a Coarse Human Silhouette

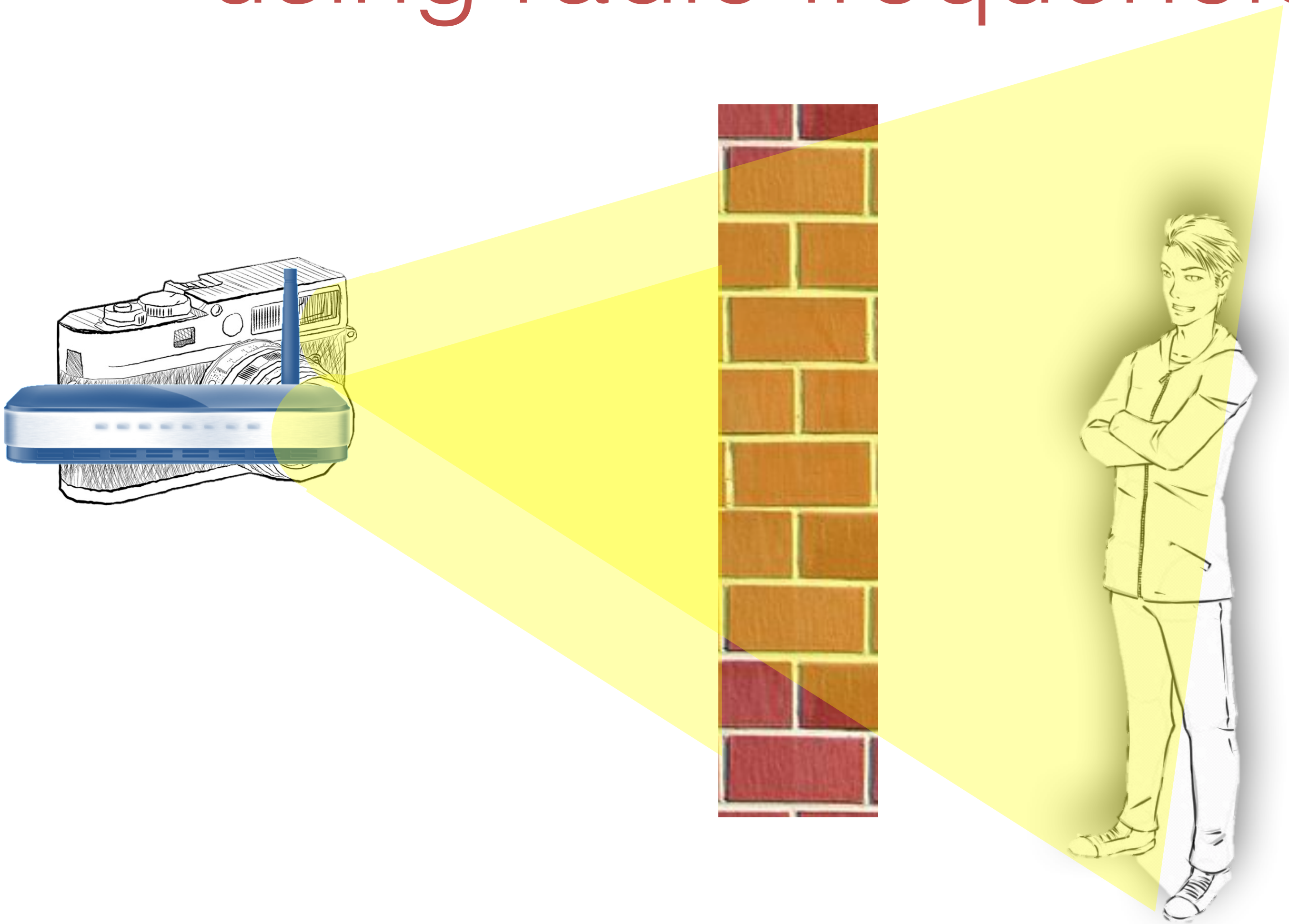




Imaging through occlusions



Imaging through occlusions using radio frequencies



Traditional Imaging

Cannot image through occlusions like walls

Form 2D images using lenses

Get a reflection from all points: can image all the body

RF Imaging

+ Walls are transparent and can image through them

− No lenses at these frequencies

− No reflections from most points: all reflections are specular

RF Imaging

+ Walls are transparent and can image through them

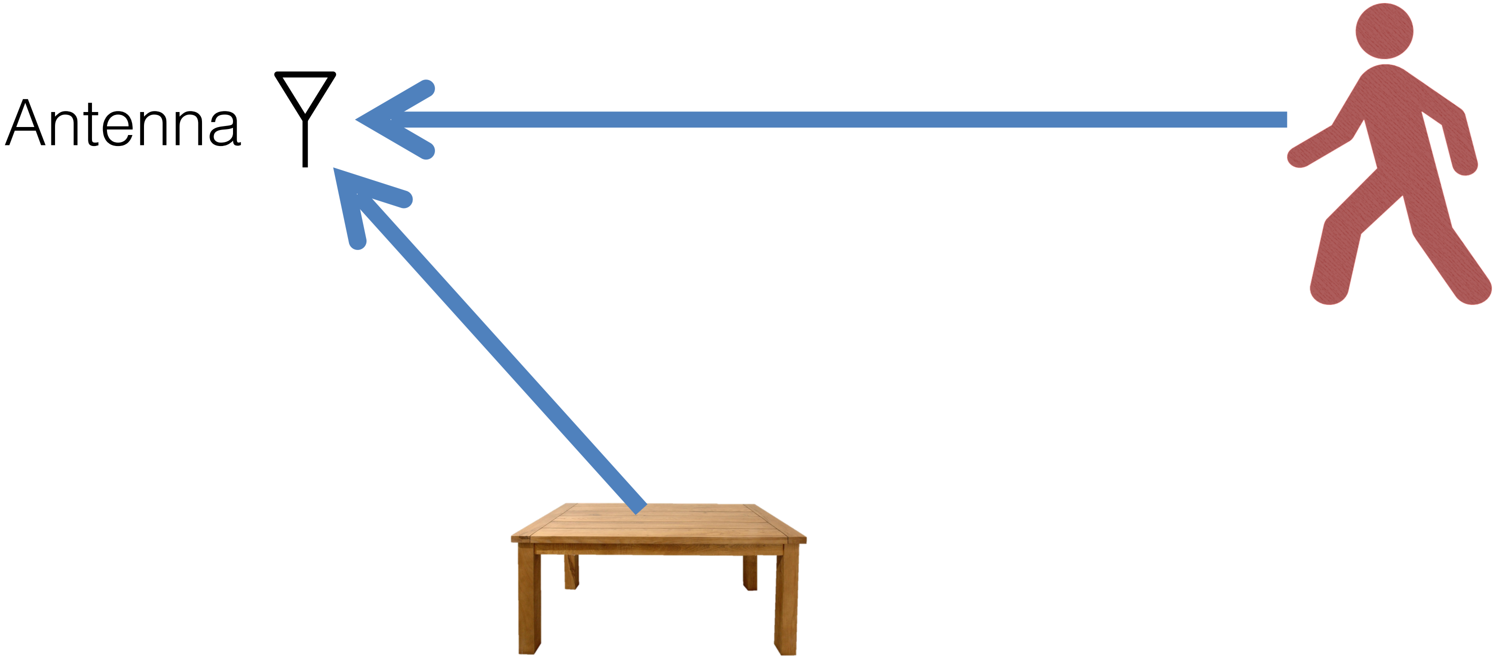
− No lenses at these frequencies → Our Solution: A component that scans 3D space with RF and outputs reflection snapshots at every point in time

− No reflections from most points: all reflections are specular → ?

Imaging with RF

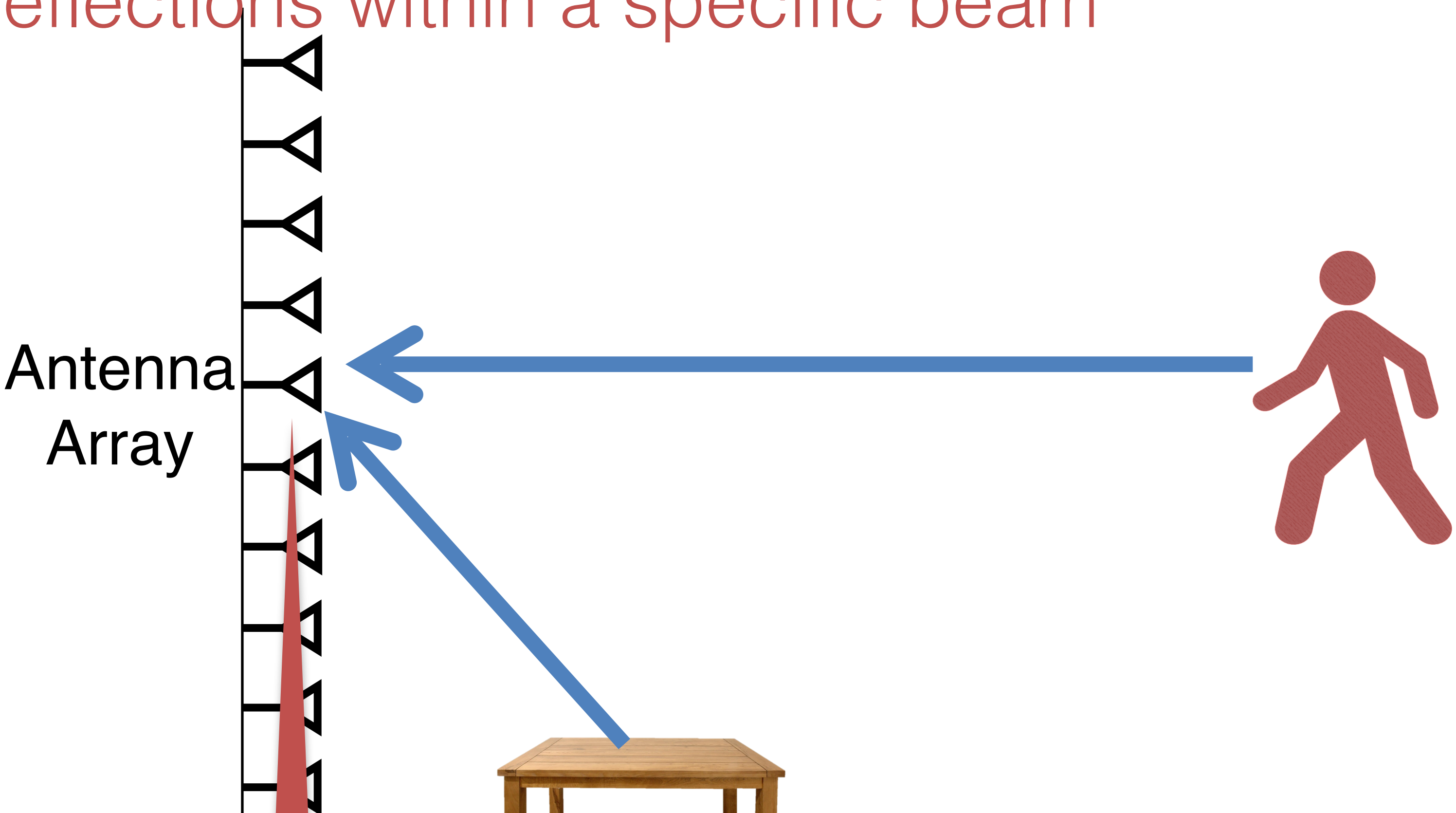
No lens at these frequencies

Antenna cannot distinguish bounces from different directions



Imaging with RF

Beamforming: Use multiple antennas to scan reflections within a specific beam



Extend to 3D with time-of-flight measurements by repeating this at every depth

Scanning every direction is slow

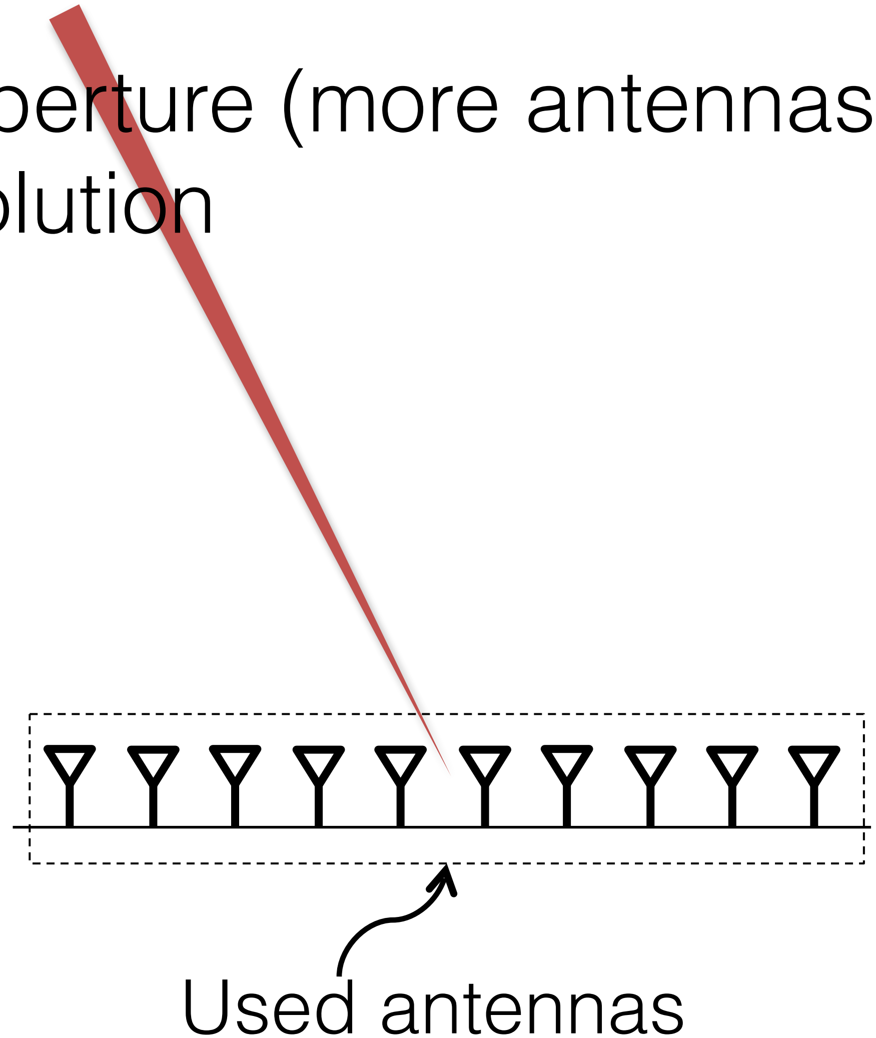
- Each angle/depth needs to be processed separately

- Most of the 3D scene is empty

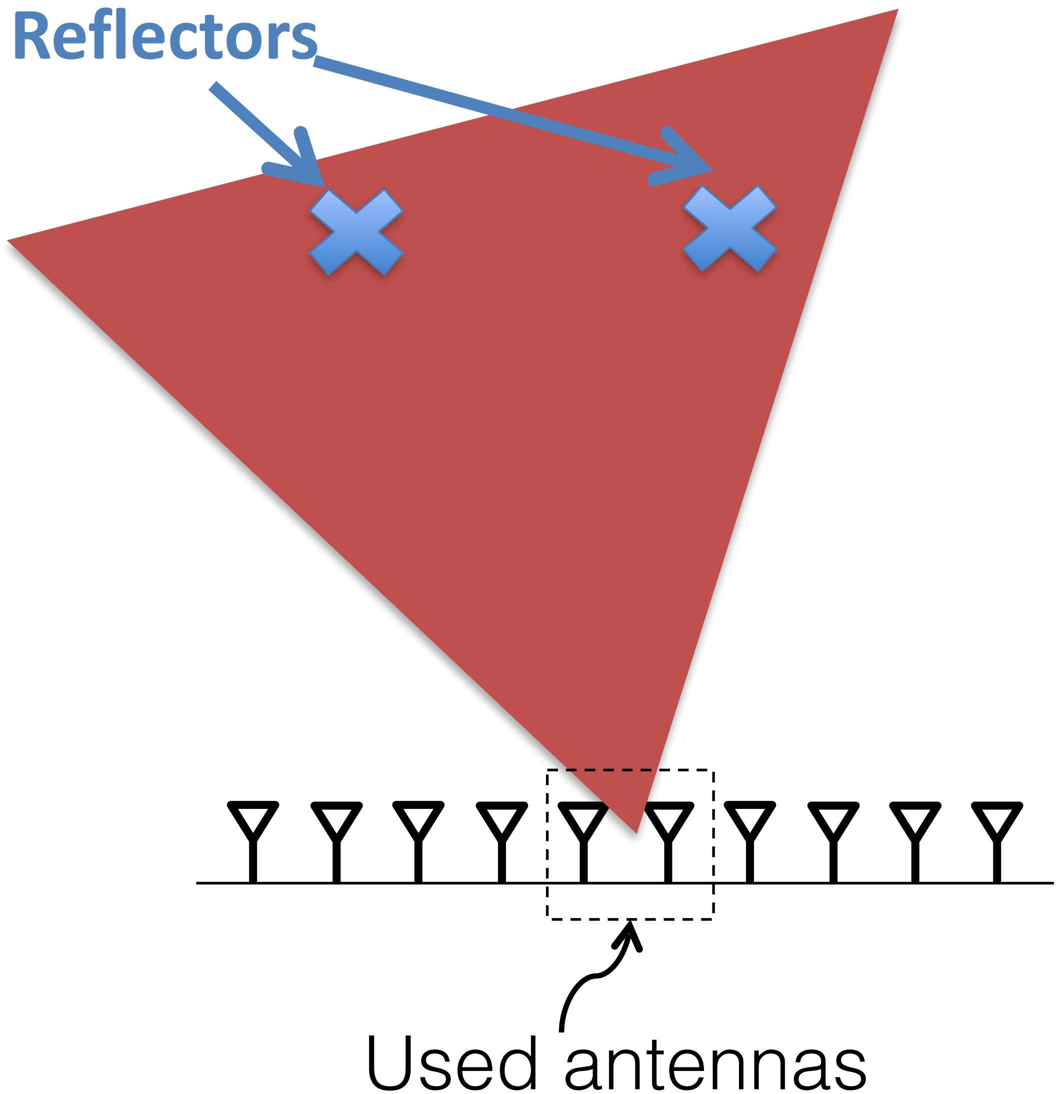
- <u>Our solution:</u> Coarse-to-fine scan that iteratively refines the resolution

Coarse-to-fine Scan

- Larger aperture (more antennas) means finer resolution



Coarse-to-fine Scan



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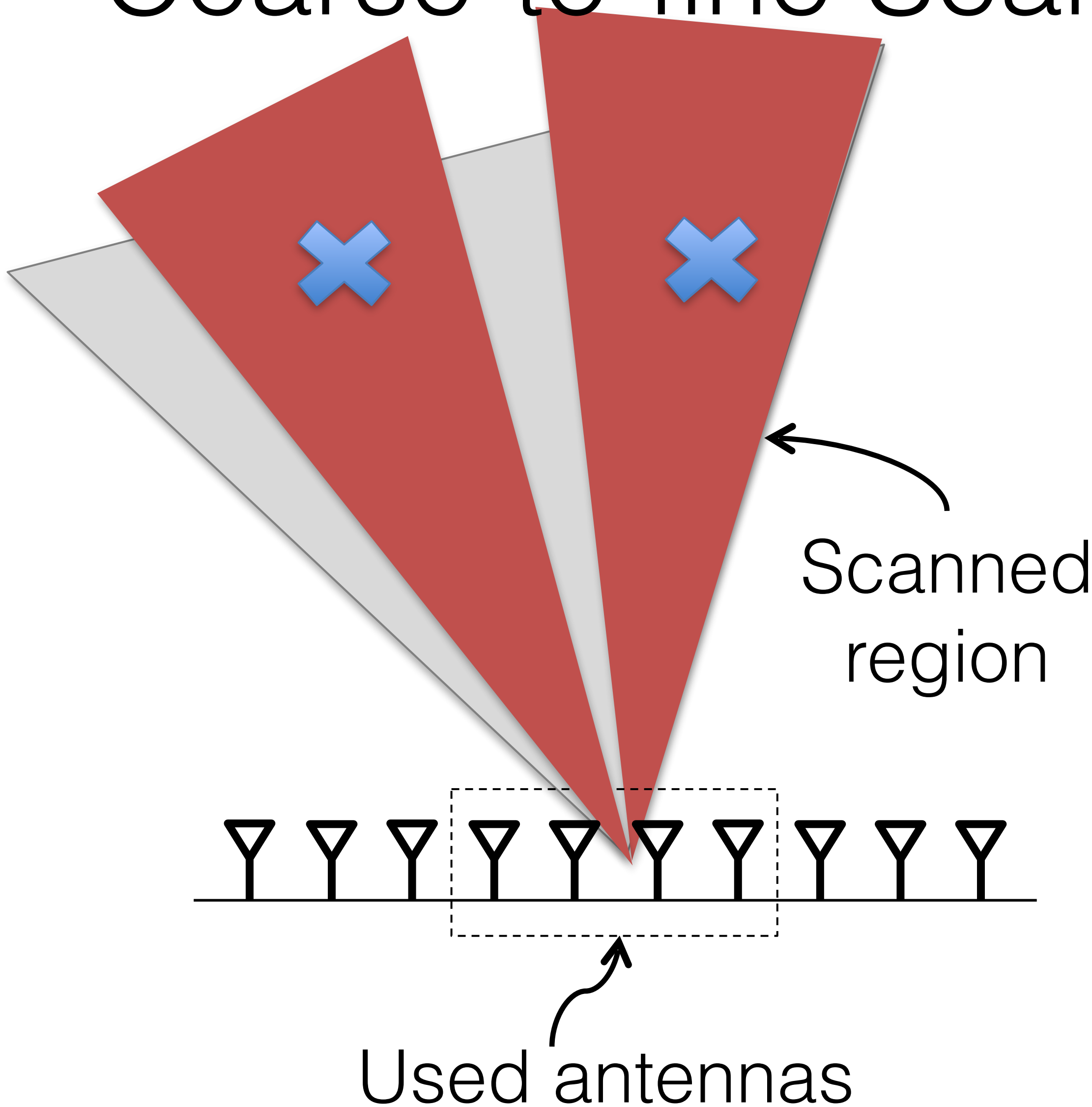
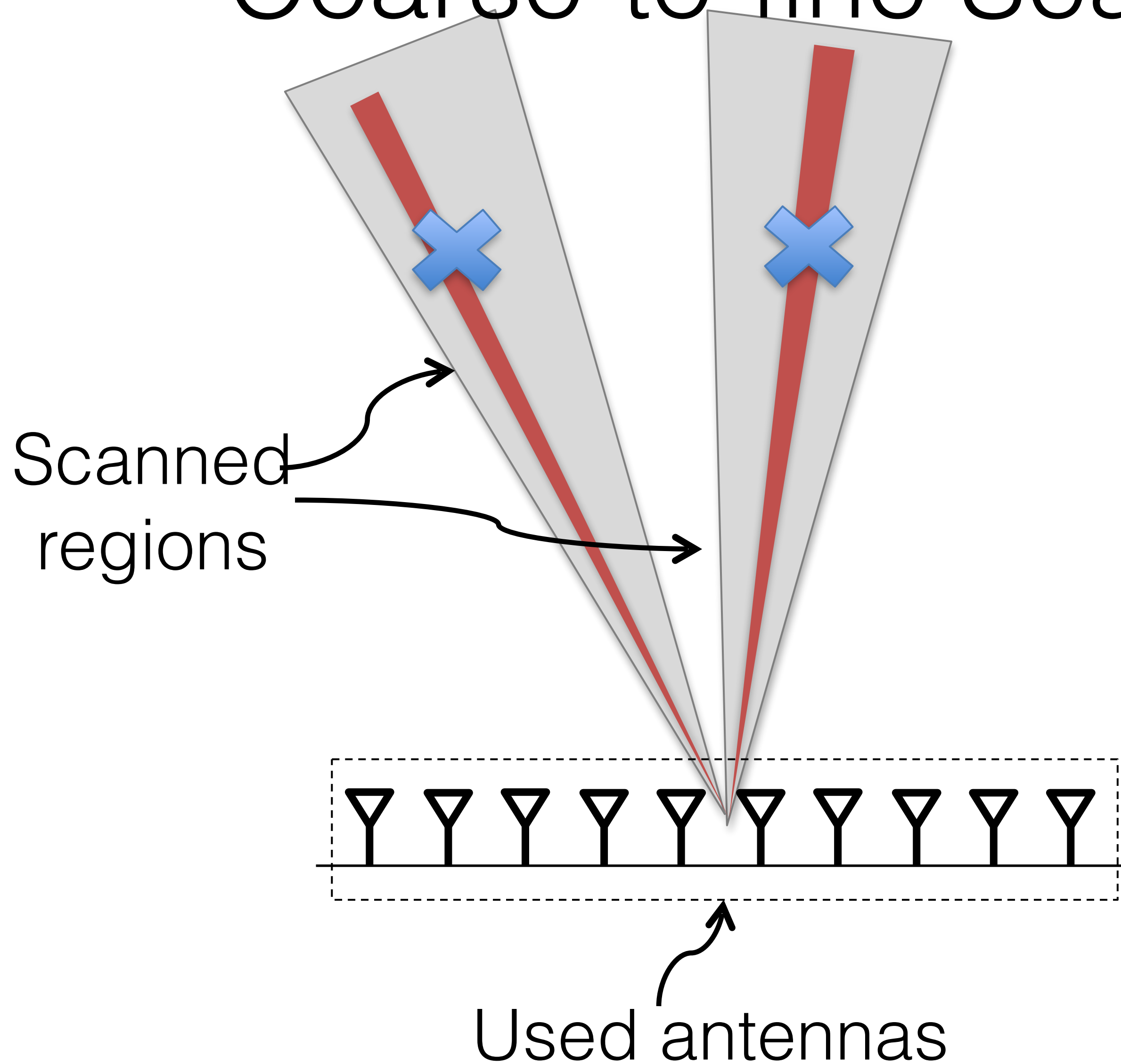
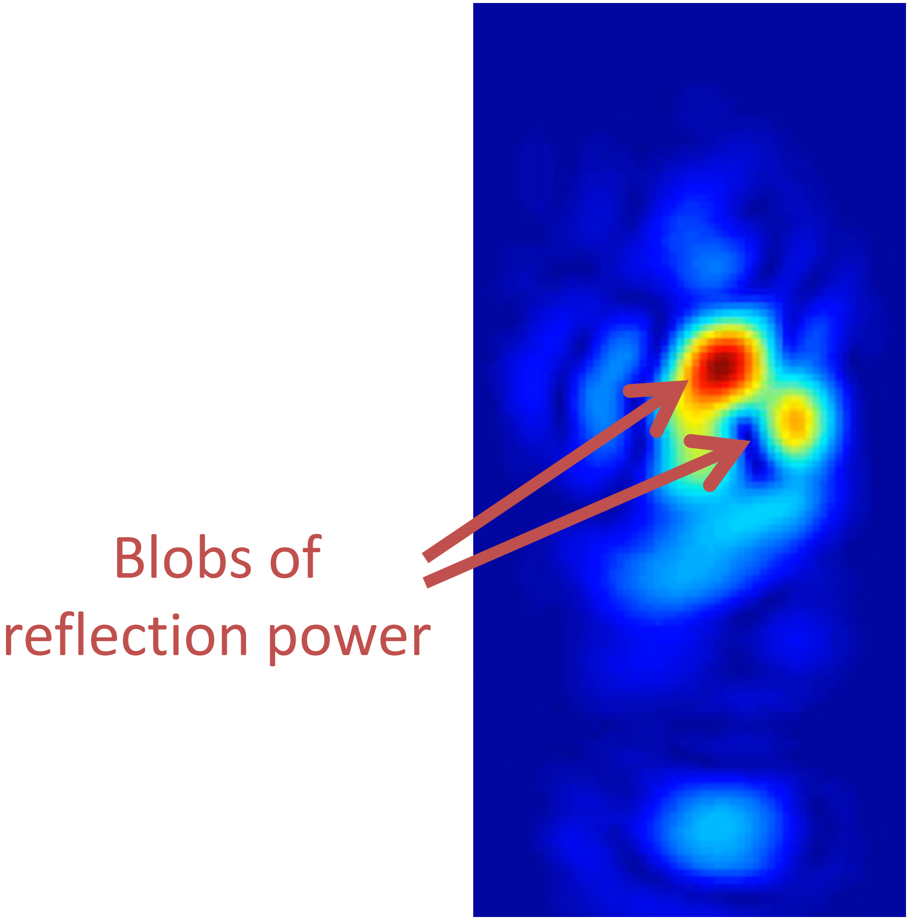


Coarse-to-fine Scan

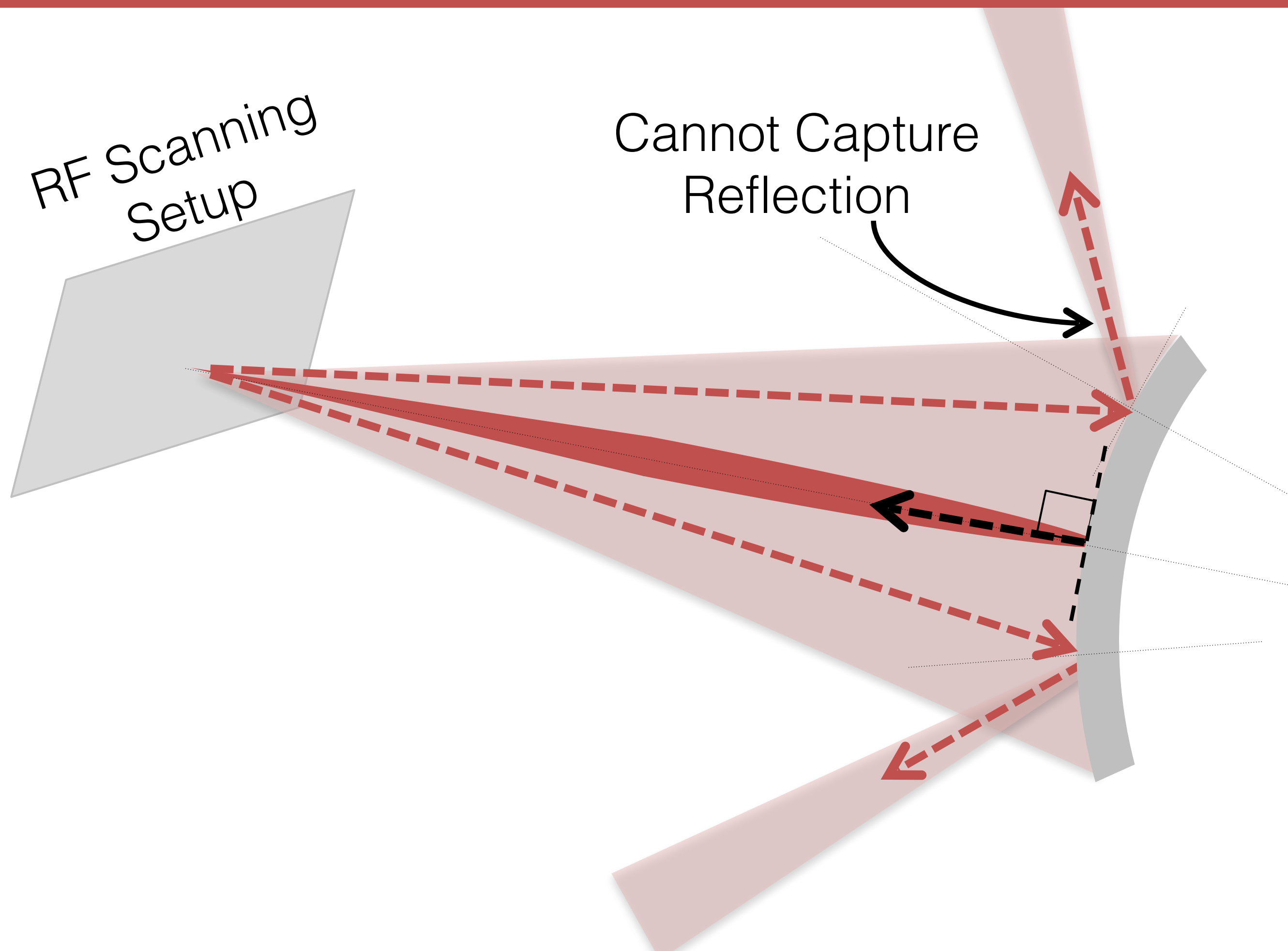


<u>Challenge:</u> We only obtain blobs in space

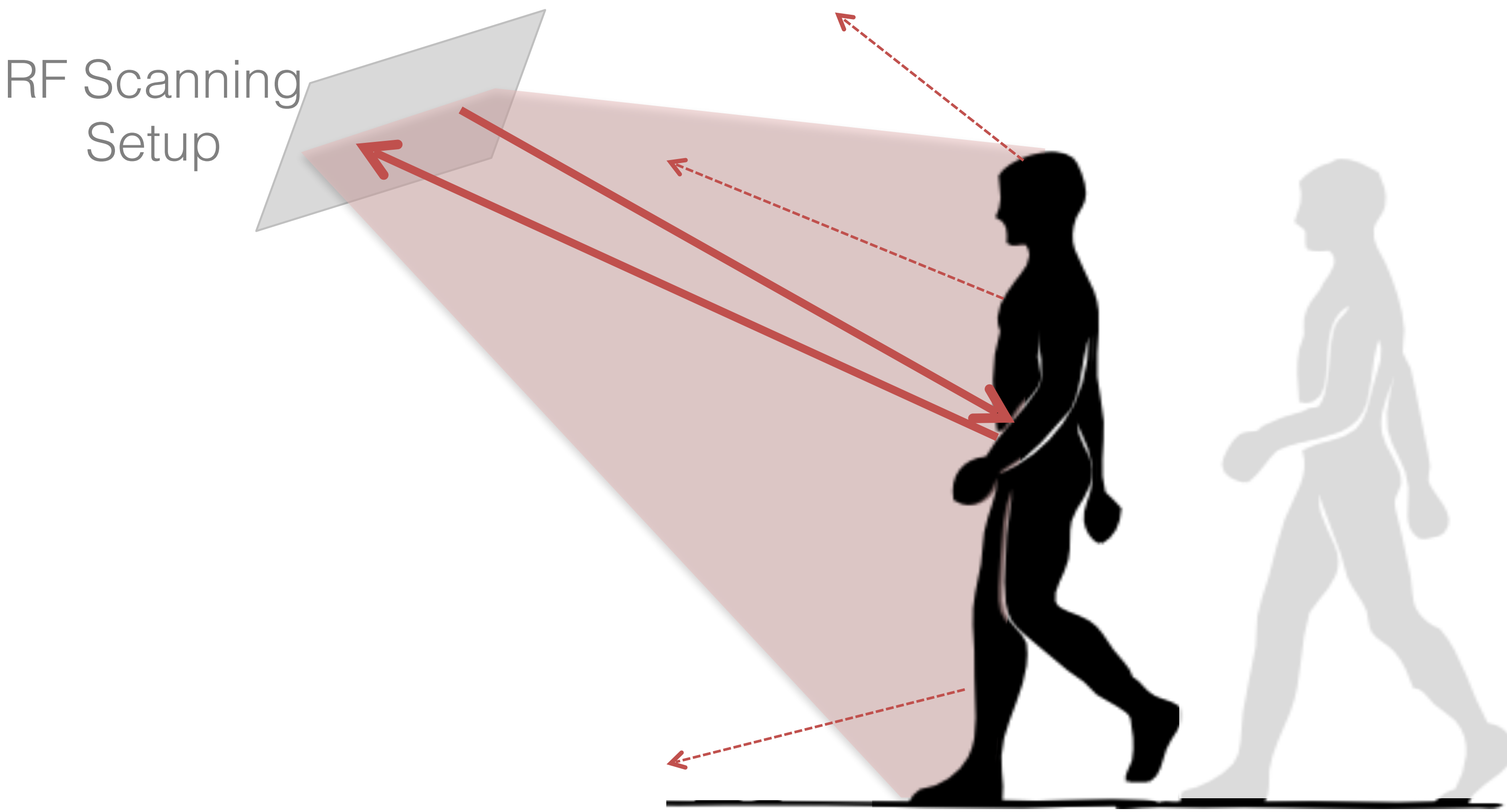
Output of 3D RF Scan



At every point in time, we get reflections from only a subset of body parts.



Solution Idea: Exploit Human Motion and Aggregate over Time



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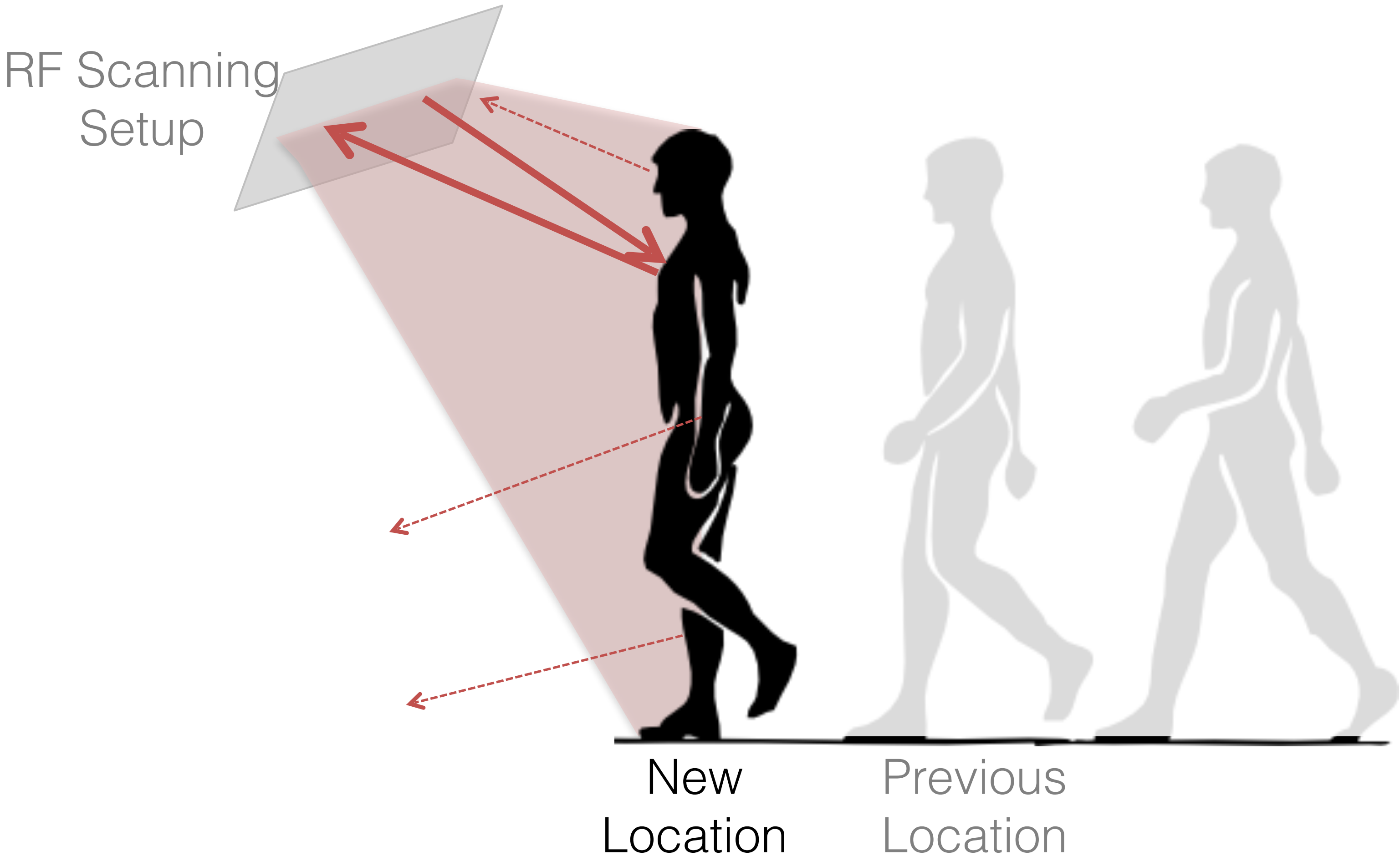
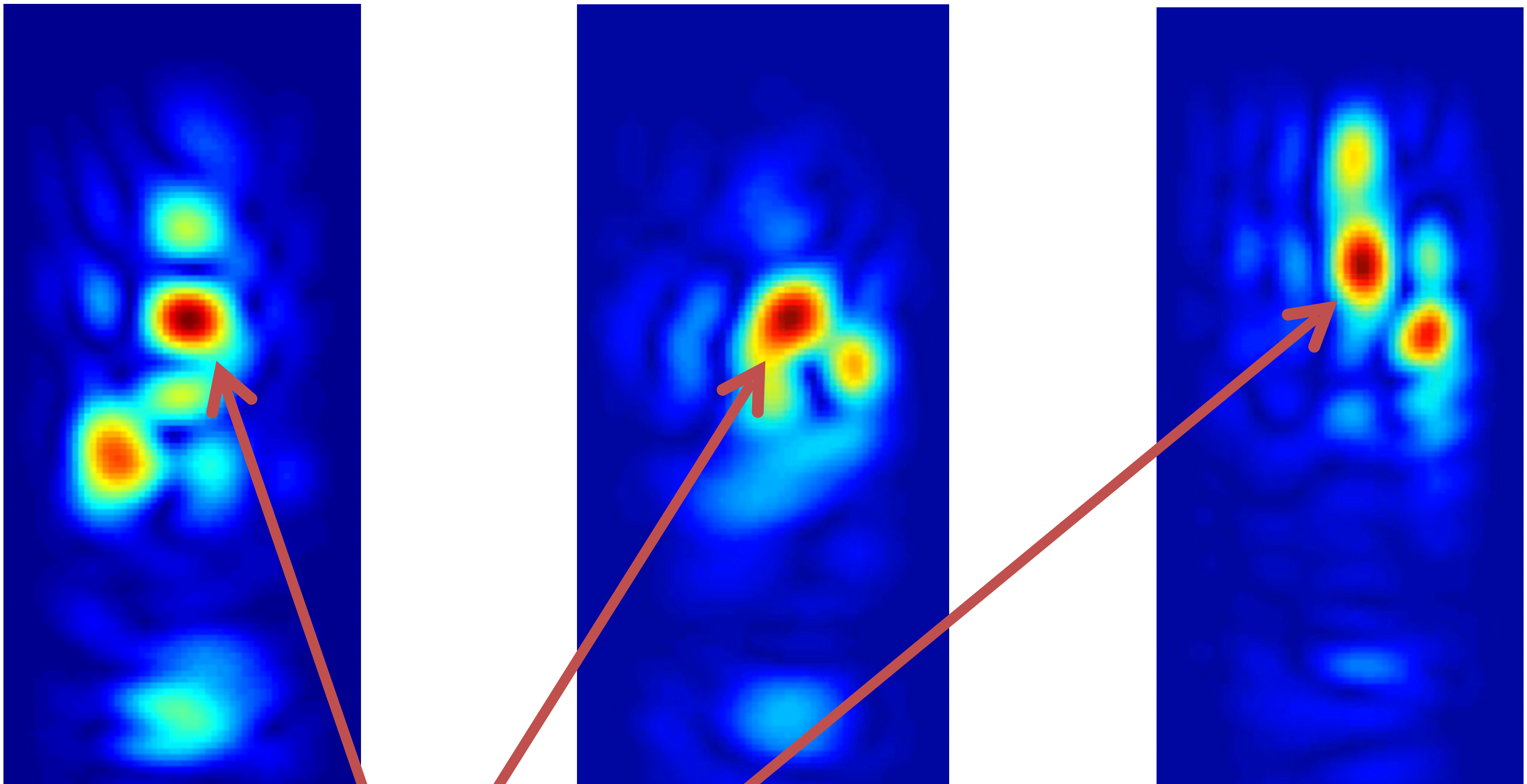


Combine the various snapshots

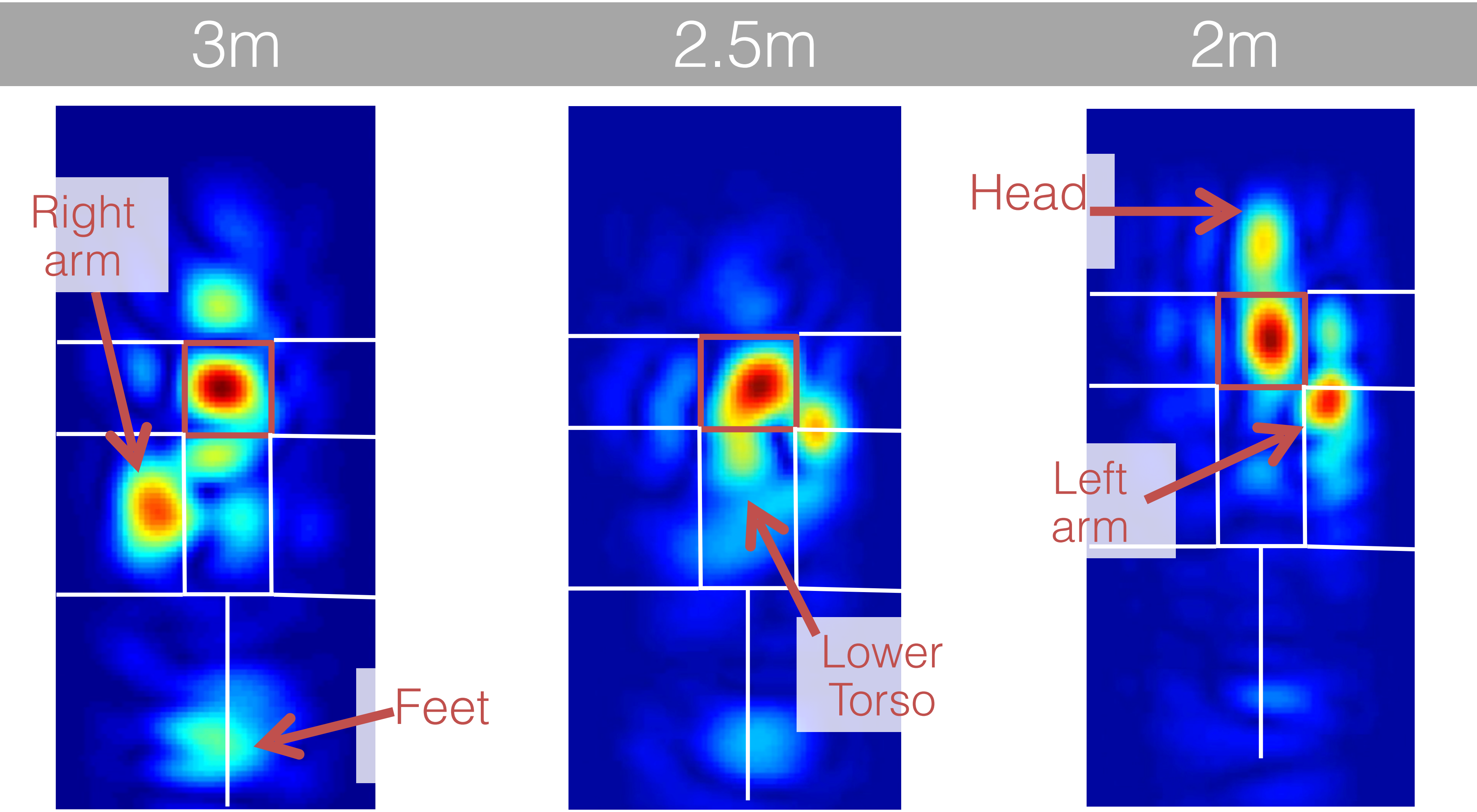
Human Walks toward Sensor

3m	2.5m	2m



Chest (Largest Convex Reflector) → Use it as a pivot: for motion compensation and segmentation

Human Walks toward Sensor

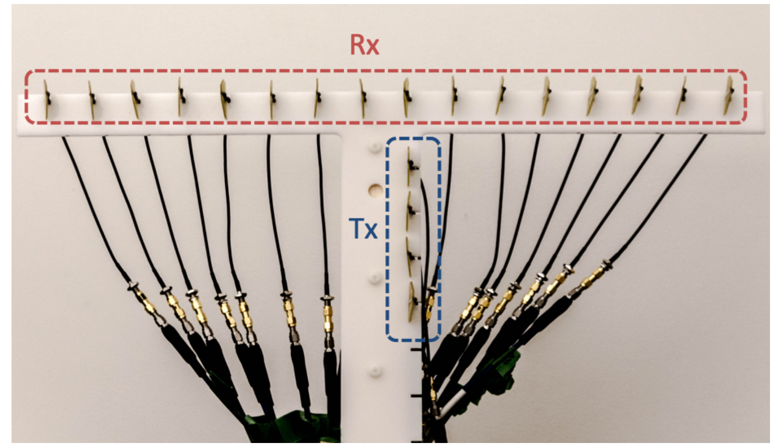


Combine the various snapshots

Human Walks toward Sensor

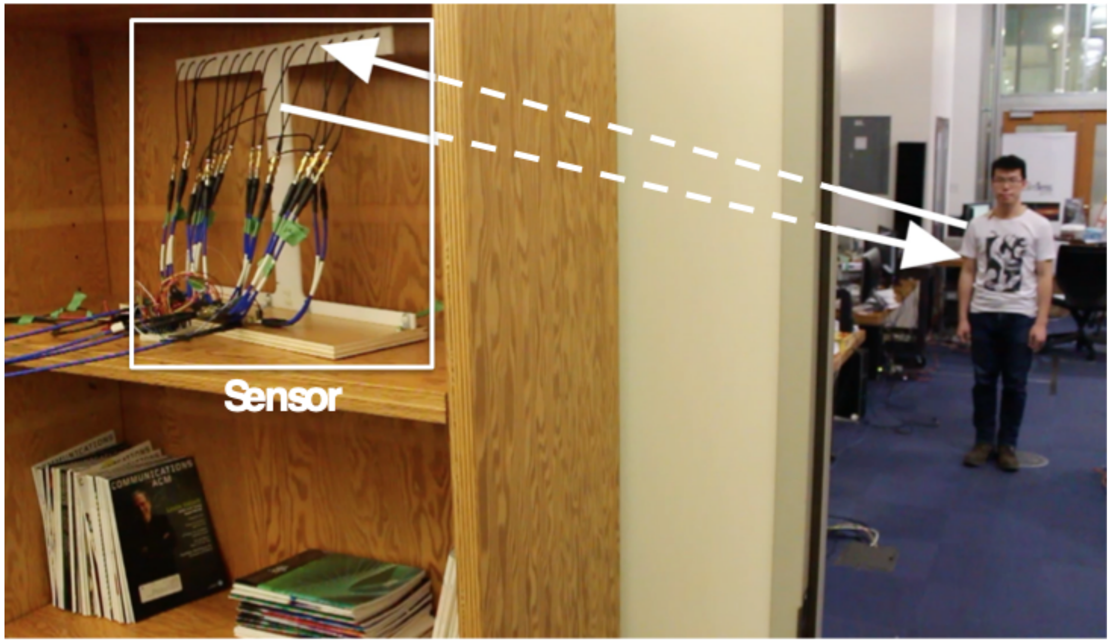
Implementation

- Hardware
  - 2D Antenna Array
  - Built RF circuit
    - 1/1,000 power of WiFi
    - USB connection to PC



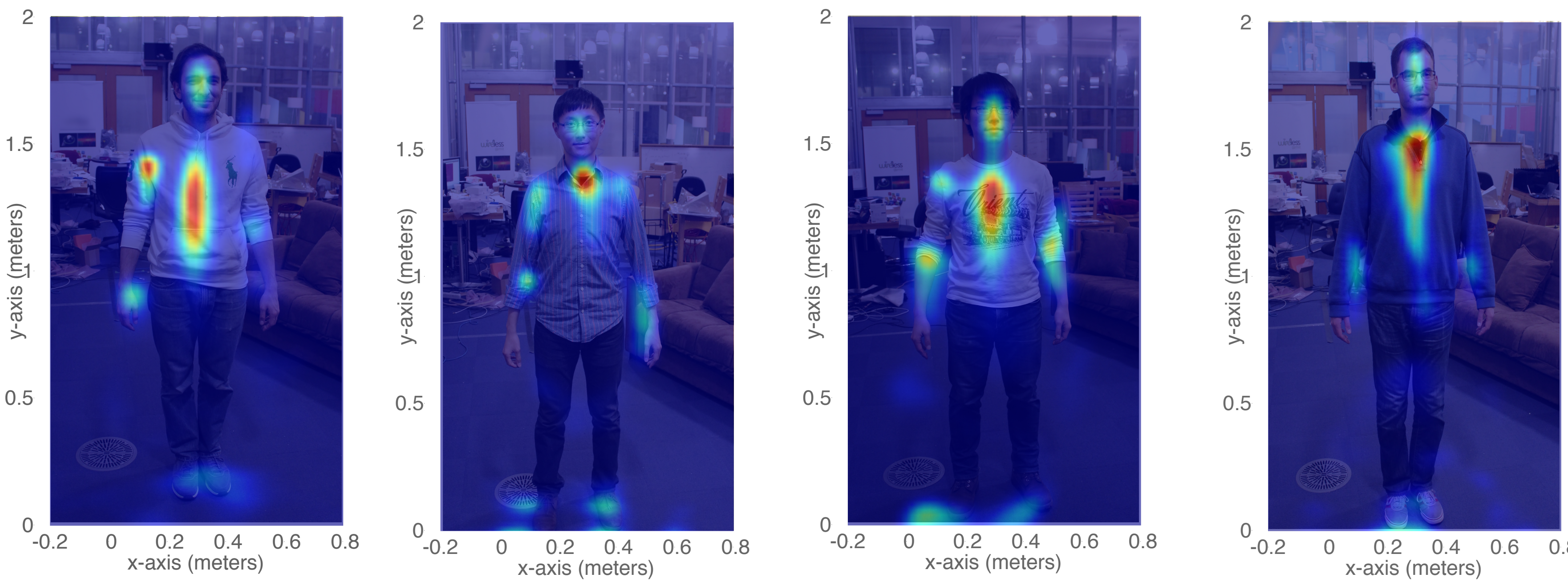
- Software
  - Coarse-to-fine algorithm implemented in GPU to generate reflection snapshots in real-time

Evaluation

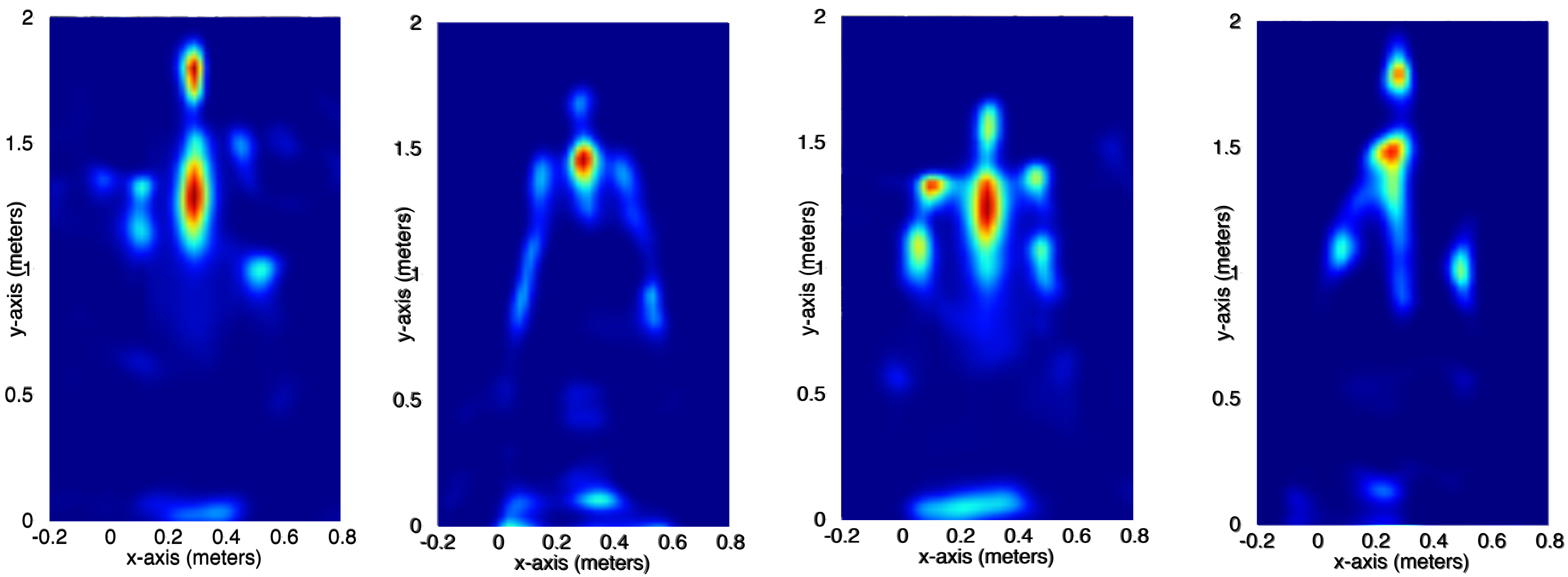


- RF-Capture sensor placed behind the wall
- 15 participants
- Use Kinect as baseline when needed

Sample Captured Figures through Walls



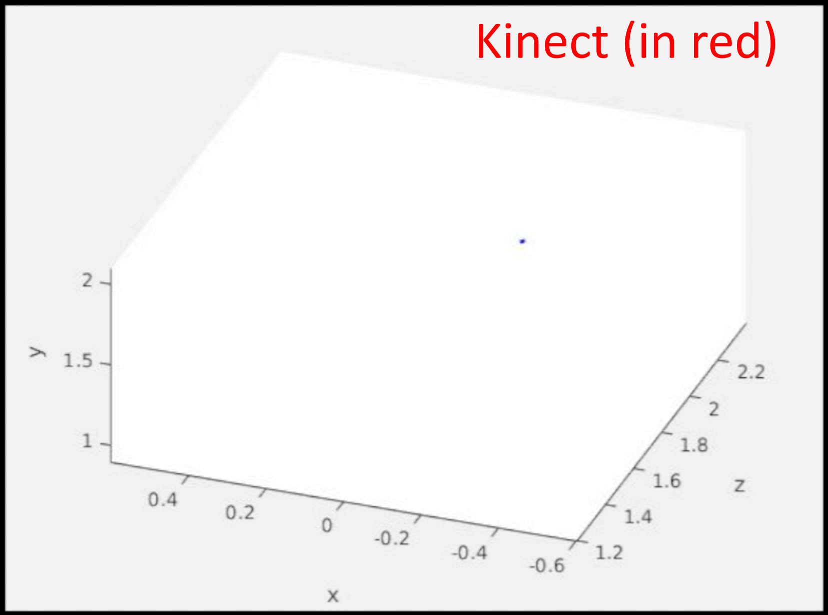
Through-wall classification accuracy of 90% among 13 users



Writing in the air

Device

Our Tracking Result



Median Accuracy is 2cm