# 6.S062: Mobile and Sensor Computing 

Lecture 4: Device-Free Localization

## So Far: Device-based Localization



This Lecture: Using radio signals to track humans without any sensors on their bodies

This Lecture: Using redidsignals to track humans without any senbors on their bodies


## - Location <br> - Vital Signs <br> - Gestures

Operates through occlusions

## Example: WiTrack



## Applications

## Smart Homes



Energy Saving
Gaming \& Virtual Reality


## Measuring Distances



Distance $=$ Reflection time $x$ speed of light

## Measuring Reflection Time

## Option1: Transmit short pulse and listen for echo



## Measuring Reflection Time

## Option1: Transmit short pulse and listen for echo



Capturing the pulse needs sub-nanosecond sampling

Why?

Capturing the pulse needs subnanosecond sampling Why?

Multi-GHz samplers are expensive, have high noise, and create large I/O problem

Why was this not a problem for Cricket?


Distance $=$ time $\times$ speed

$10 \mathrm{~cm}=\Delta t^{2} \times c^{133 \times 10^{2}}$

$$
\Delta t=\frac{1}{3} \times 10^{-9}=0.3 \sin 5
$$

$$
\begin{aligned}
& \text { Creket" } \Rightarrow 300 \mathrm{~m} / 5 \text { vs } 311^{3} \\
& \text { ultiasoont } \quad \frac{1}{30^{2}} \quad \Rightarrow 10^{6} \text { slaver } \\
& \text { BKsps }
\end{aligned}
$$

## FMCW: Measure time by measuring frequency

Transmitted


* Wires Signal eff frequency


* FMCW (chirp)



FMCW: Measure time by measuring frequency
Transmitted


How do we measure $\triangle F$ ?

## Measuring $\Delta F$

- Subtracting frequencies is easy (e.g., removing carrier in WiFi)
- Done using a mixer (low-power; cheap)

Power


Signal whose frequency is $\Delta F$

Basics of Boric transform. $\begin{aligned} & \text { time signal } \\ & i \\ & \vdots \\ & i\end{aligned}+$ ค


## Measuring $\Delta F$

- Subtracting frequencies is easy (e.g., removing carrier in WiFi)
- Done using a mixer (low-power; cheap)


Signal whose frequency is $\Delta F$

## $\Delta F \rightarrow$ Reflection Time $\rightarrow$ Distance

# Challenge: Multipath $\rightarrow$ Many Reflections 



## Static objects don't move

$\rightarrow$ Eliminate by subtracting consecutive measurements


Why 2 peaks when we only have one moving person?


Distance

The direct reflection arrives before dynamic multipath!


## Mapping Distance to Location

Person can be anywhere on an ellipse whose foci are ( $T x, R x$ )


By adding another antenna and intersecting the ellipses, we can localize the person

From Location to tracking


Fails for multiple people in the environment, and we need a more comprehensive solution

Tx
Rx


How can we deal with multi-path reflections when there are multiple persons in the environment?

Idea: Person is consistent across different vantage points while multi-path is different from different vantage points

Combining across Multiple Vantage Points
Experiment: Two users walking
Setup


Single Vantage Point


Mathematically: each round-trip distance can be
mapped to an ellipse whose foci are the transmitter and the receiver

## Combining across Multiple Vantage Points

Experiment: Two users walking
Setup



# Combining across Multiple Vantage Points 

Experiment: Two users walking

Setup

16 Vantage Points


Localize the two users

## Multi-User Localization

## Experiment: Four persons walking

## Setup

All Vantage Points

first person
other people or noise?

## Near-Far Problem: Nearby persons have more

 power than distance reflectors and can mask themSetup

four persons

All Vantage Points

first person other people or noise?

## Successive Silhouette Cancellation:

 a new algorithm that localizes multiple persons in the scene by addressing the near-far problem
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a new algorithm that localizes multiple persons in the scene by addressing the near-far problem

Goal: Recover human reflections

Decode human location

> Model human and reconstruct reflection patterns

First localize the user with the strongest reflection


## Cancel the impact of the person's whole body



After reconstructing and cancelling the first user's reflections


Iteratively localize the remaining users in the scene


Iteratively localize the remaining users in the scene


## How can we localize static users?

## Dealing with multi-path when there is one moving user



We eliminated direct table reflections by subtracting consecutive measurements

## Needs User to Move

## Dealing with multi-path when there is one moving user



We eliminated direct table reflections by subtracting consecutive measurements

## Needs User to Move

# Exploit breathing motion for localize static users 

- Breathing and walking happen at different time scales
-A user that is pacing moves at $1 \mathrm{~m} / \mathrm{s}$
-When you breathe, chest moves by few mm/s
- Cannot use the same subtraction window to eliminate multi-path


## User Walking at 1m/s

30ms subtraction window


Localize the person

3s subtraction window


Person appears in two locations

## User Sitting Still (Breathing)

30ms subtraction window


Cannot localize

3s subtraction window


Localize the person

## User Sitting Still (Breathing)

30ms subtraction window


3s subtraction window


Use multi-resolution subtraction window to eliminate multi-path while being able to localize both static and moving users


## Centimeter-scale localization without requiring the user to carry a wireless device



Localize the two users

## Want a silhouette

People are points



Approach: Combine antenna arrays with FMCW to get 3D image

- 2D Antenna array gives 2 angles
- FMCW gives depth (1D)

2D array


# Challenge: We only obtain blobs in space 



## 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

## 2.5 m



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



## Sample Captured Figures through Walls






## Through-wall classification accuracy of 90\% among 13 users



