

# CONTEXT AWARENESS AND LOCALIZATION



# Context-aware computing



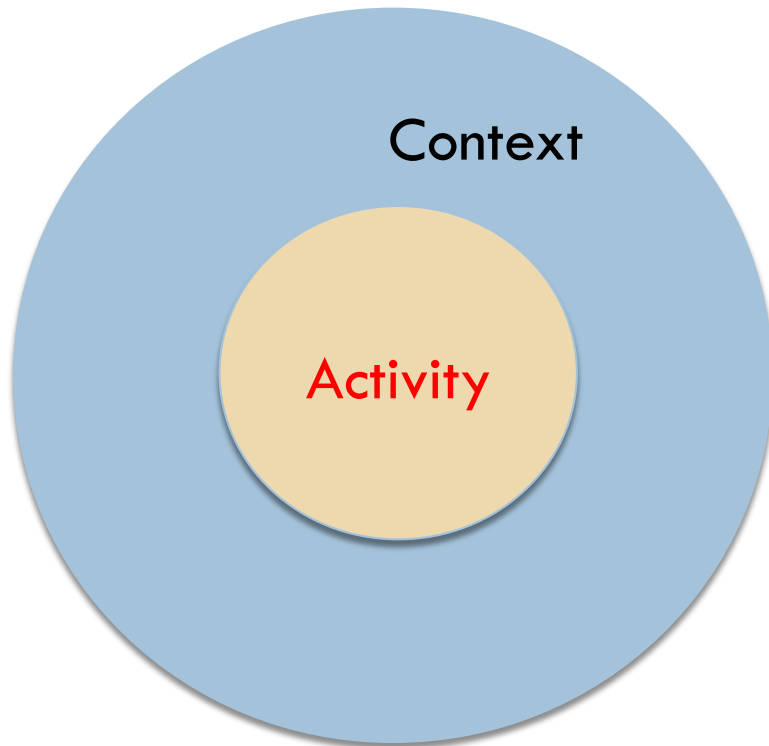
- How computation can be made sensitive and responsive to its context
  - ▣ What context?
  - ▣ How to represent/evaluate/detect?
  - ▣ How to respond?

# What is Context?



- Dictionary definition: “the interrelated conditions in which something exists or occurs”
- The representational view
  - ▣ “Context encompasses more than just the user’s location, because other things of interest are also mobile and changing. Context includes lighting, noise level, network connectivity, communication costs, communication bandwidth and even the social situation (e.g., who you are with)

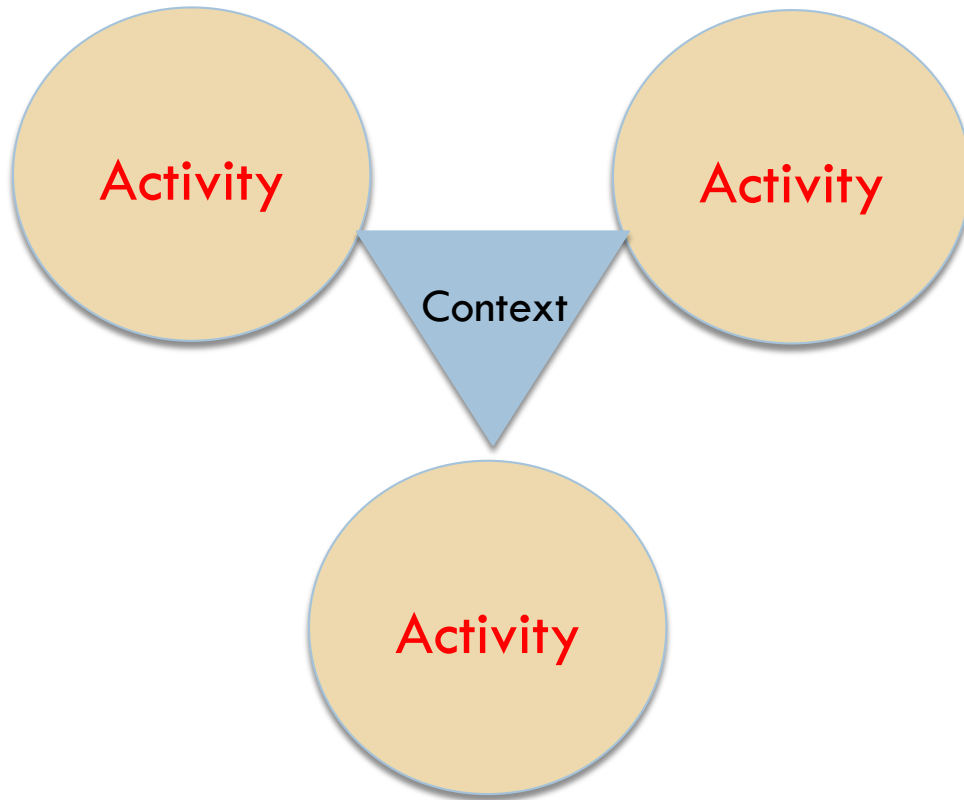
# The representational view



- Example: adaptive ring tone
  - Activity: ring
  - Context: noise level of the environment, location
  - Relation: noise level of the environment decides the volume of the ring tone

The key question is thus how to encode and represent *relevant* context

# The interactional view



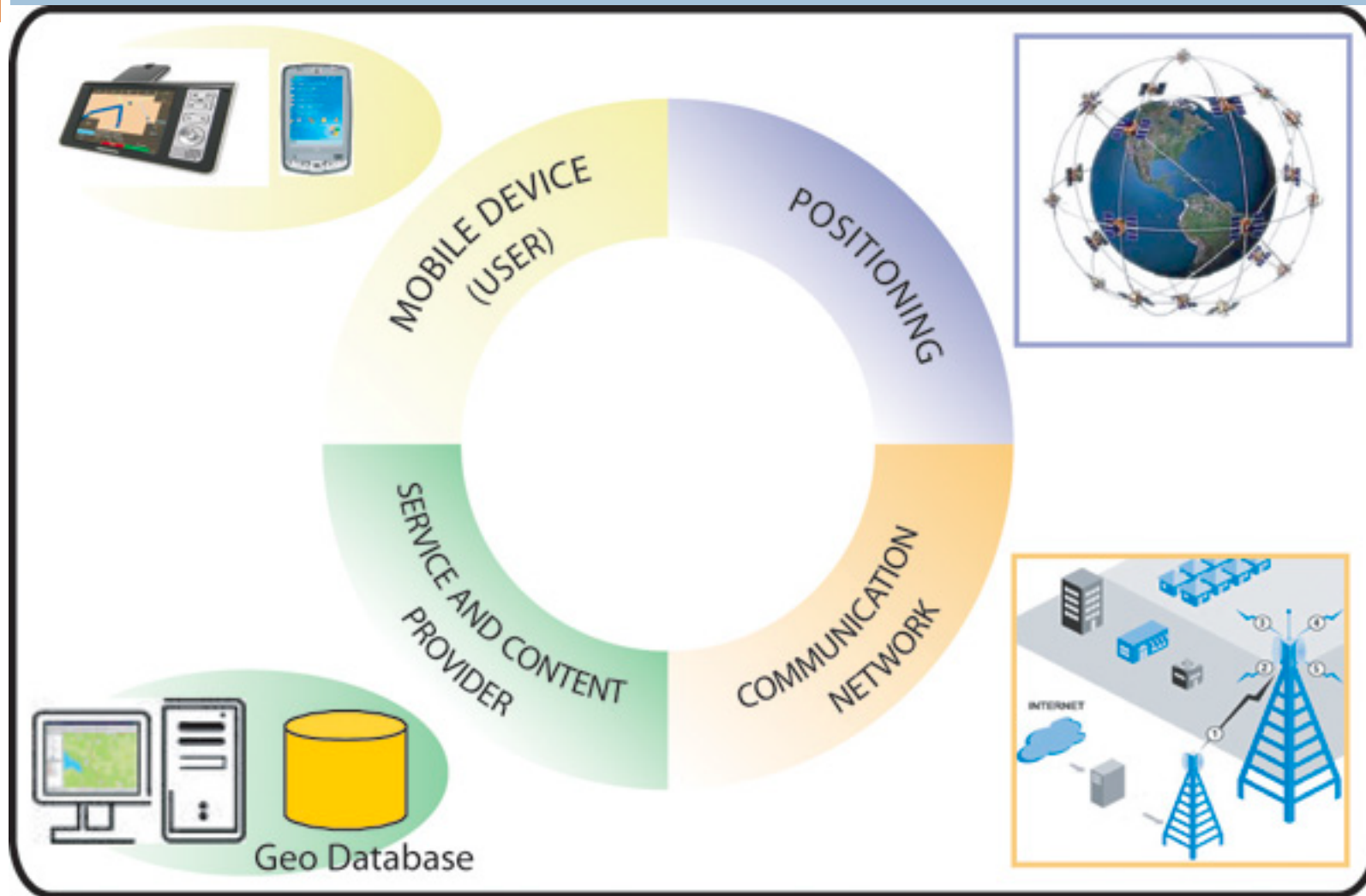
- *contextuality is a relational property that holds between objects or activities*
- Dynamic and evolving through the interactions
- Example: adaptive ring tone
  - ▣ Activity: people's activity, mobile ring
  - ▣ Context: The interaction among activities determine "the norm" – keep quiet

how and why, in the course of their interactions, do people achieve and maintain a mutual understanding of the context for their actions



Location, location, location

# Components of LBS



Steiniger et al. "Foundation of Location Based Services"

# Usage of LBS

Action	Questions	Operations
<b>orientation &amp; localisation</b> locating	where am I? where is {person   object}?	positioning, geocoding, geodecoding
<b>navigation</b> navigating through space, planning a route	how do I get to {place name   address   xy}?	positioning, geocoding, geodecoding routing
<b>search</b> searching for people and objects	where is the {nearest   most relevant   &}{person   object}?	positioning, geocoding, calculating distance and area, finding relationships
<b>identification</b> identifying and recognising persons or objects	{what   who   how much} is {here   there}?	directory, selection, thematic/ spatial, search
<b>event check</b> checking for events; determining the state of objects	what happens {here   there}?	



# A Taxonomy of Localization Techniques

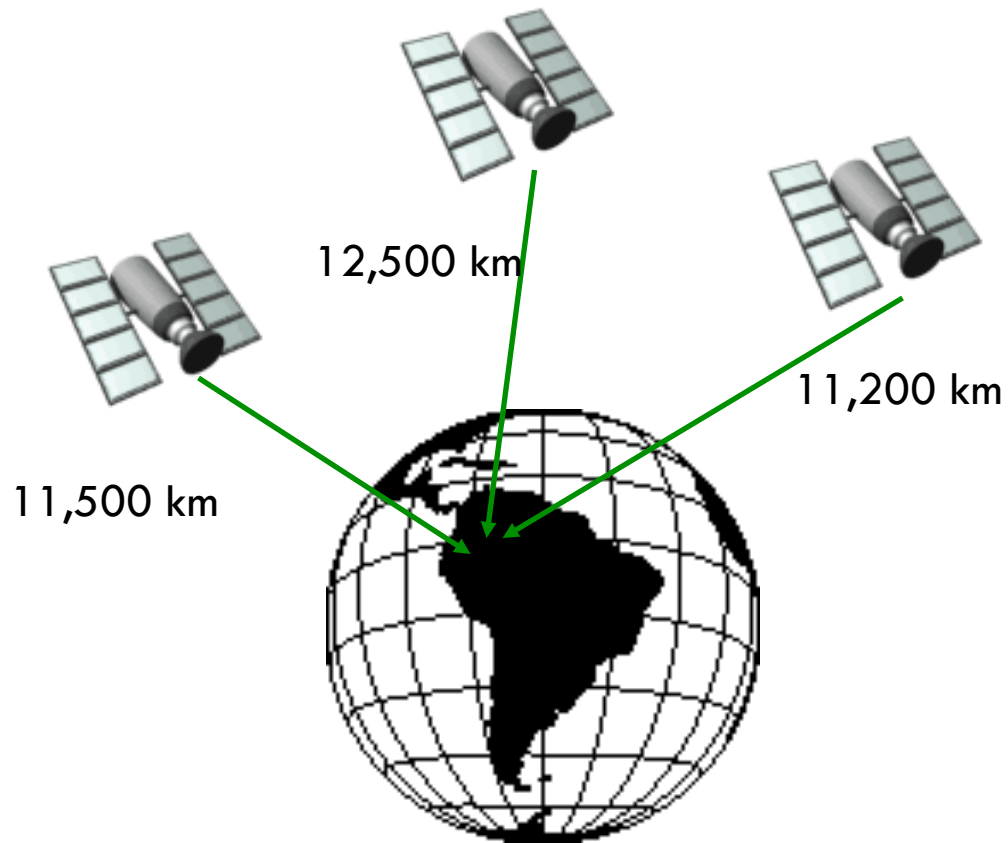
- Types of location (physical, symbolic, relative)
- Granularity of location
- How is infrastructure involved
  - ▣ Infrastructure provides the location
  - ▣ Mobile devices determine the location
- Indoor vs outdoor
- Signal used
  - ▣ **Wireless**
  - ▣ **Inertial**
  - ▣ Optical
  - ▣ Acoustic
  - ▣ ...

# Some localization techniques



- GPS
- WiFi-based indoor localization
- Inertial navigation

# How does GPS work?



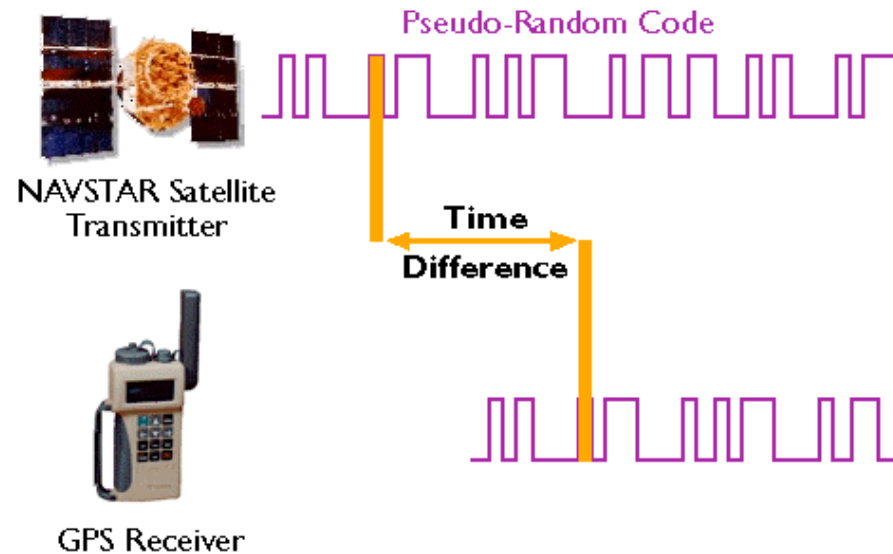
# How to measure the distance

## □ Solution 1

- Generate the same copy of the signal at the exactly the same time on the satellites and the ground unit
- Measure the time difference

delayed: "I can't fight this feeling any more,"

Local: "I can't fight this feeling any more,"

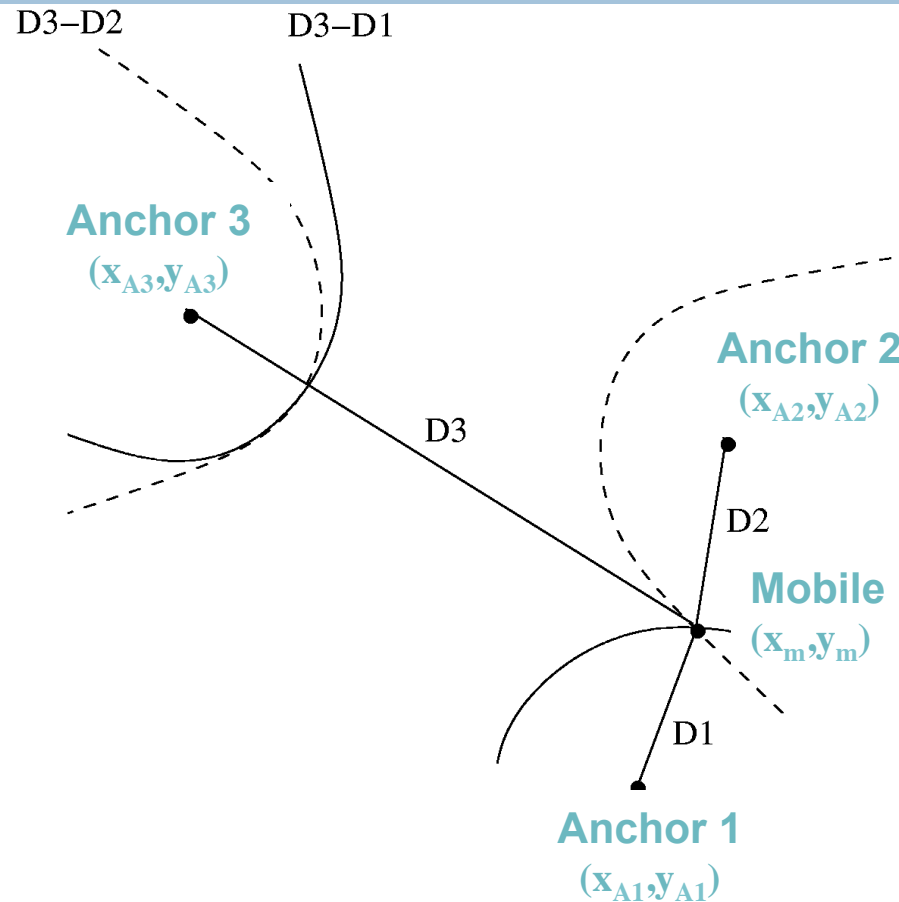


$$\text{Distance} = \text{Speed of Light} \cdot \text{Time Difference}$$

# Time Difference of Arrival (TDOA)

3 anchors with known positions (at least) are required to find a 2D-position from a couple of TDOAs

In 3D, needs the 4th satellite!

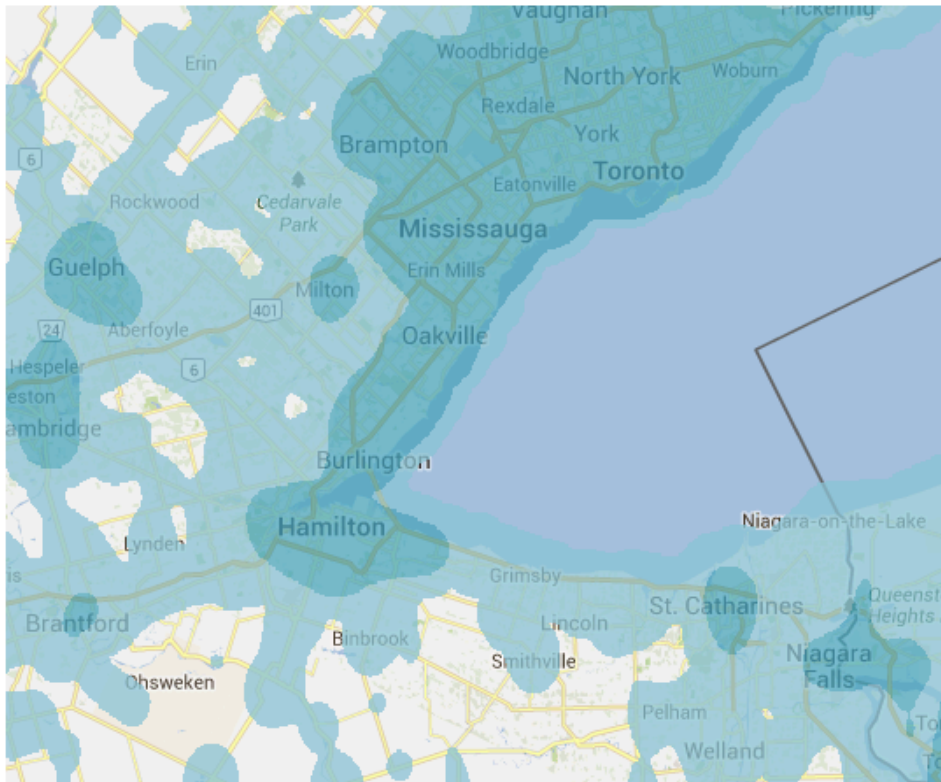


4 unknowns ( $x, y, z, \text{time}$ ) and 4 knowns

Have the added benefit of synchronizing the clock on the ground unit

# WiFi-based Indoor Localization

- Weaker signal and rich multipath indoor make GPS highly inaccurate or inaccessible
- WiFi infrastructure abundant

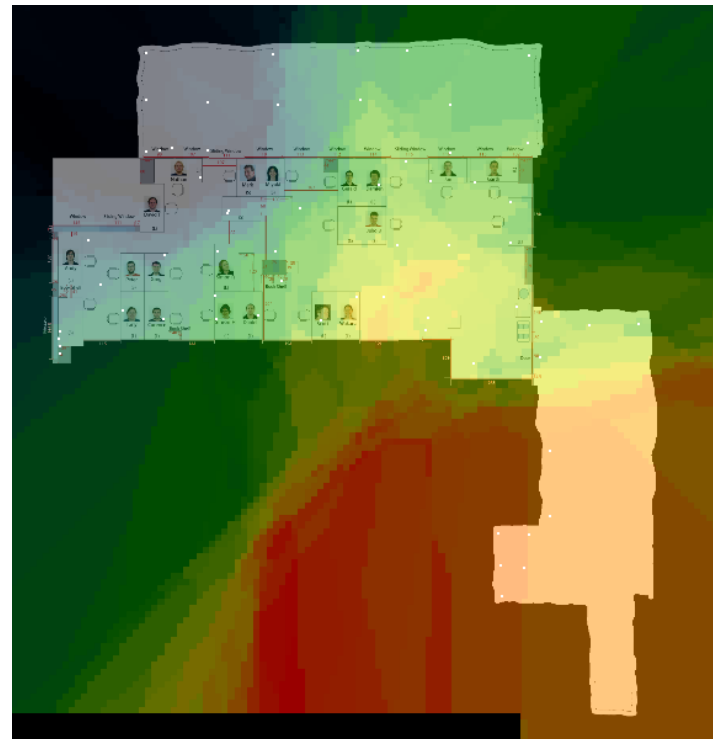
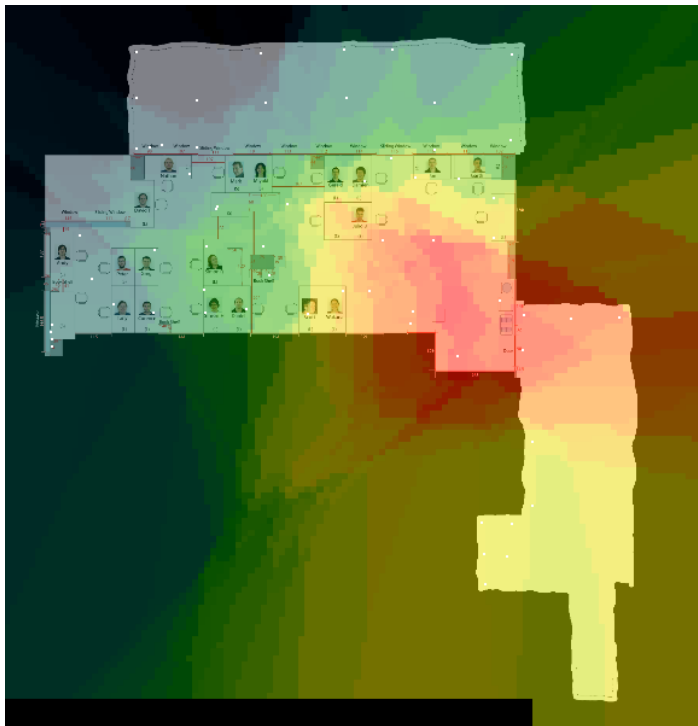


Skyhook has 275 employees, 240 of whom are drivers recording Wi-Fi signals (2008)

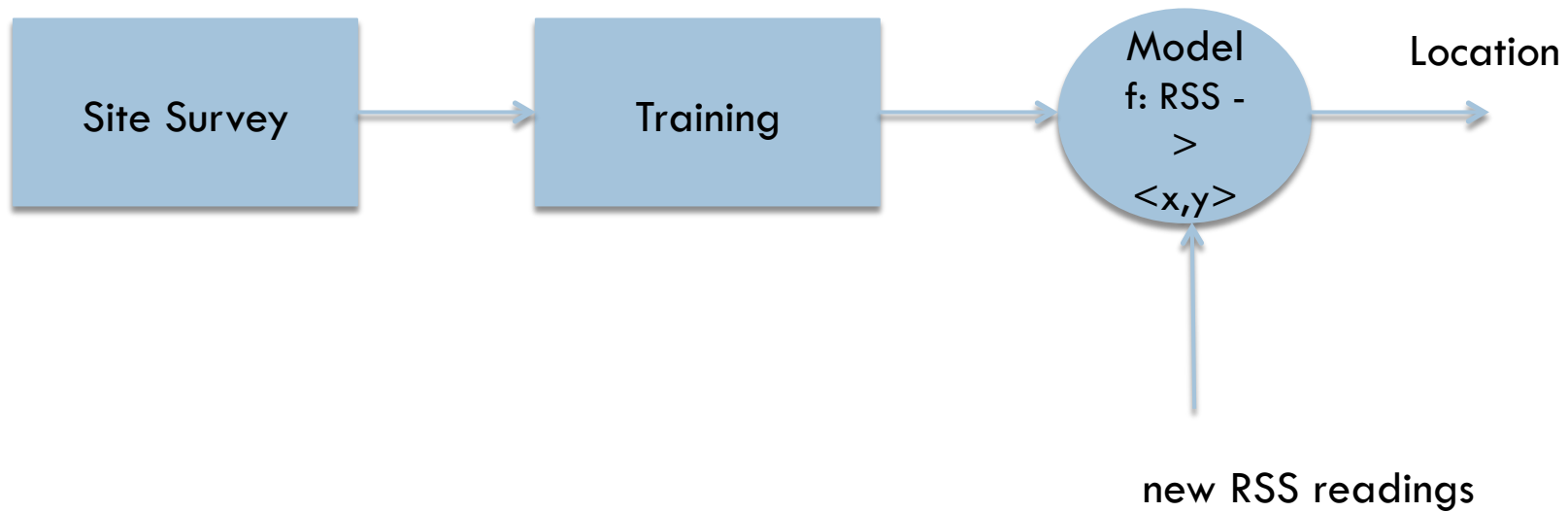
(why not yet killed by Google and Apple?)

# WiFi Fingerprinting

- ❑ TOA, TODA, AOA are generally difficult to be estimated accurately with WiFi devices
- ❑ Small-scale fading leads to large variations of received WiFi signal even when the device is stationary



# Solution approach



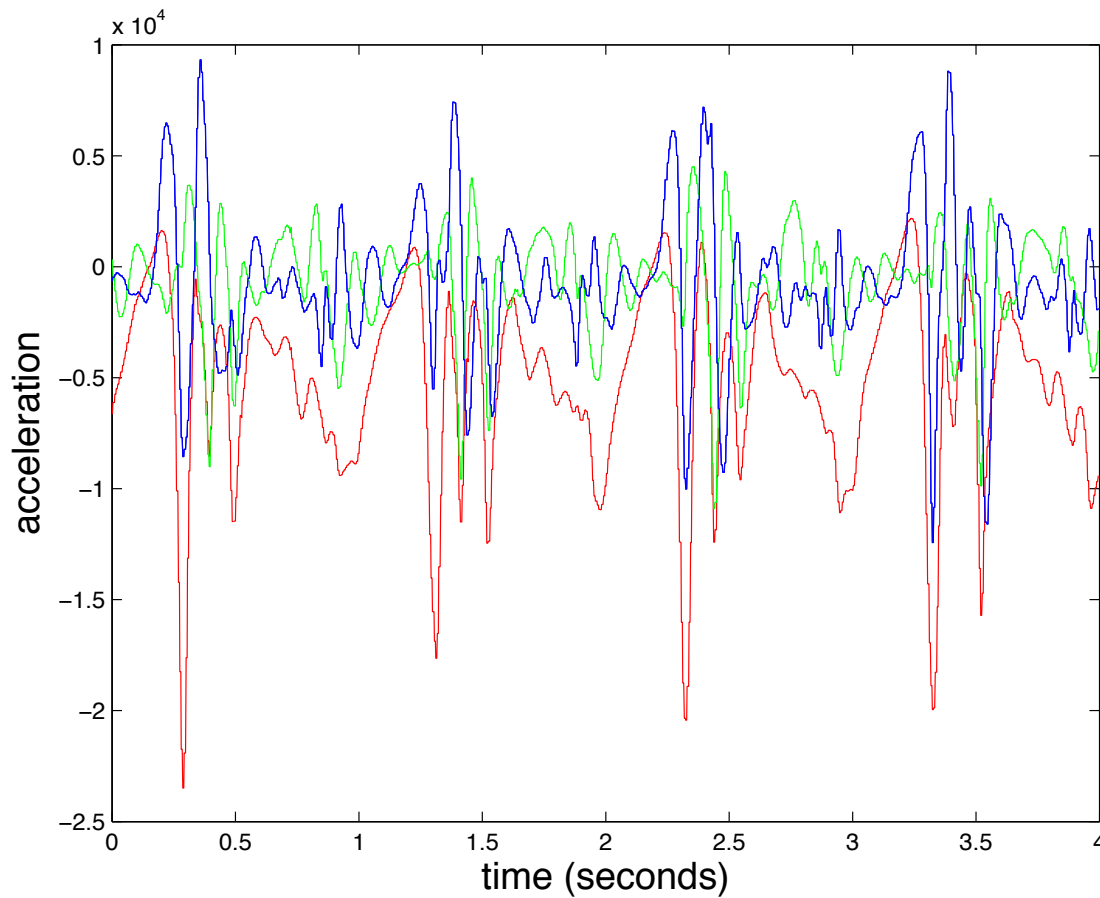
Other signatures can be used, e.g., CSI



# Challenges with FP-based Approaches

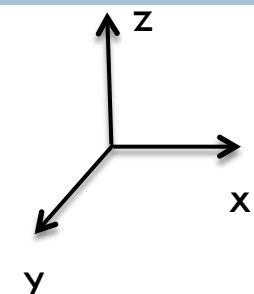
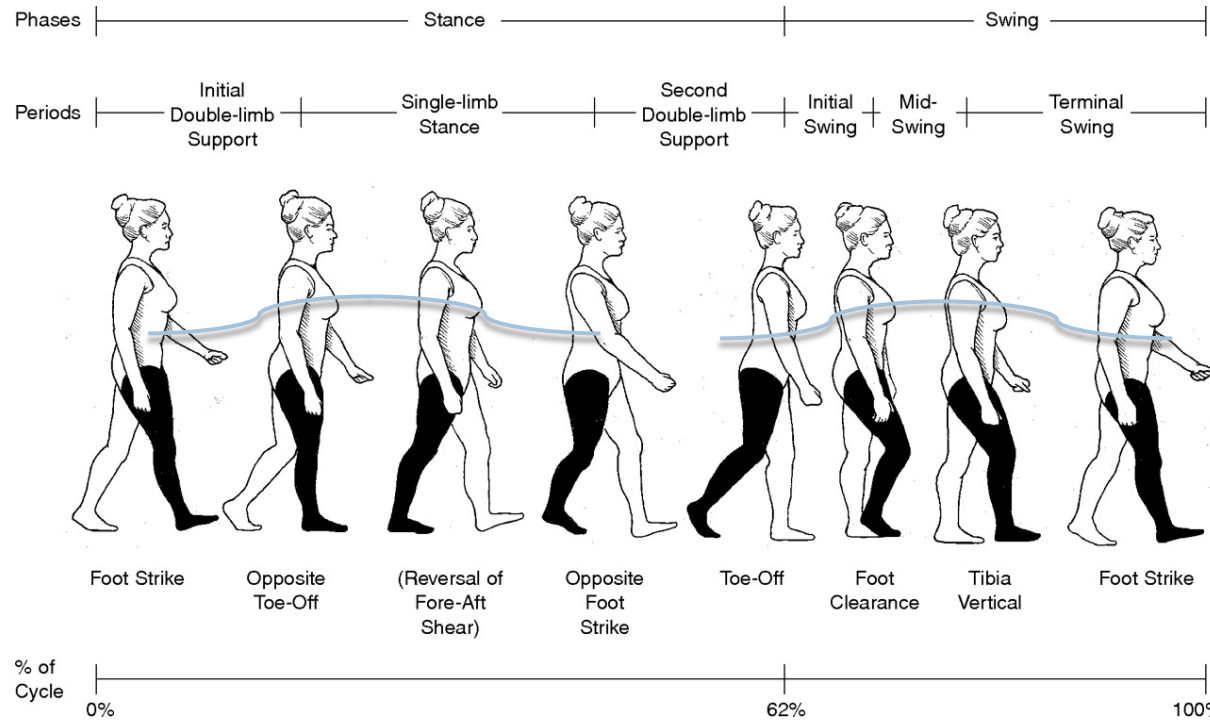
- Time-varying
- Boils down to a supervised clustering approach
- Device heterogeneity
- Needs site survey
  - ▣ Subject to changes
- Room-level accuracy
- Map required to determine the symbolic locations
- Solution: Other sensing modalities
  - ▣ Inertial sensors: accelerometers, gyro sensor, magnetometer/compass
  - ▣ Ranging sensors: acoustic, infrared, ultra-wide band RF, laser

# Accelerometer readings while walking



□ Potential energy  
to/from kinetic  
energy

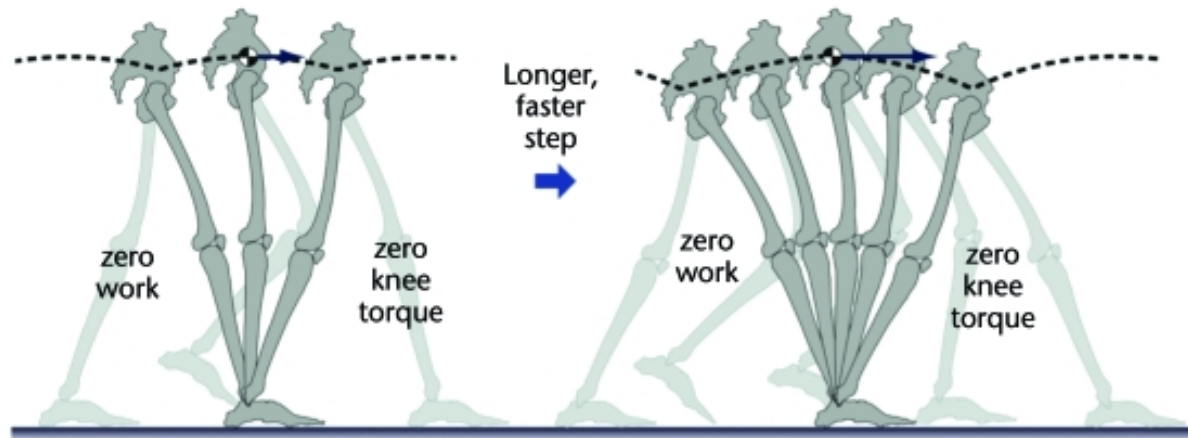
# Gait cycle



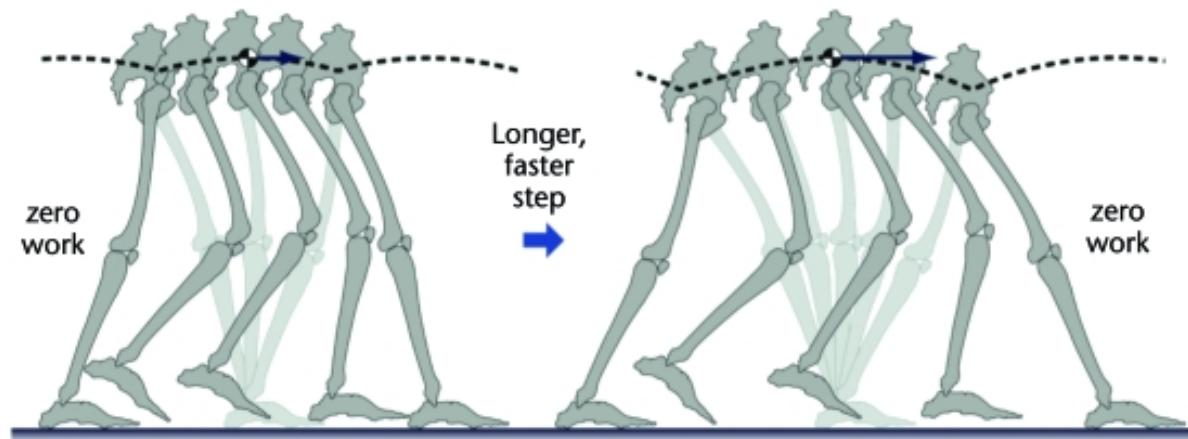
X	-	+	+	-	-	+	+	0
Z	+	-	-	+	+	-	-	+
Y	+	0	-	-	0	+	+	+

# Inverted pendulum model

Inverted Pendulum Stance Leg

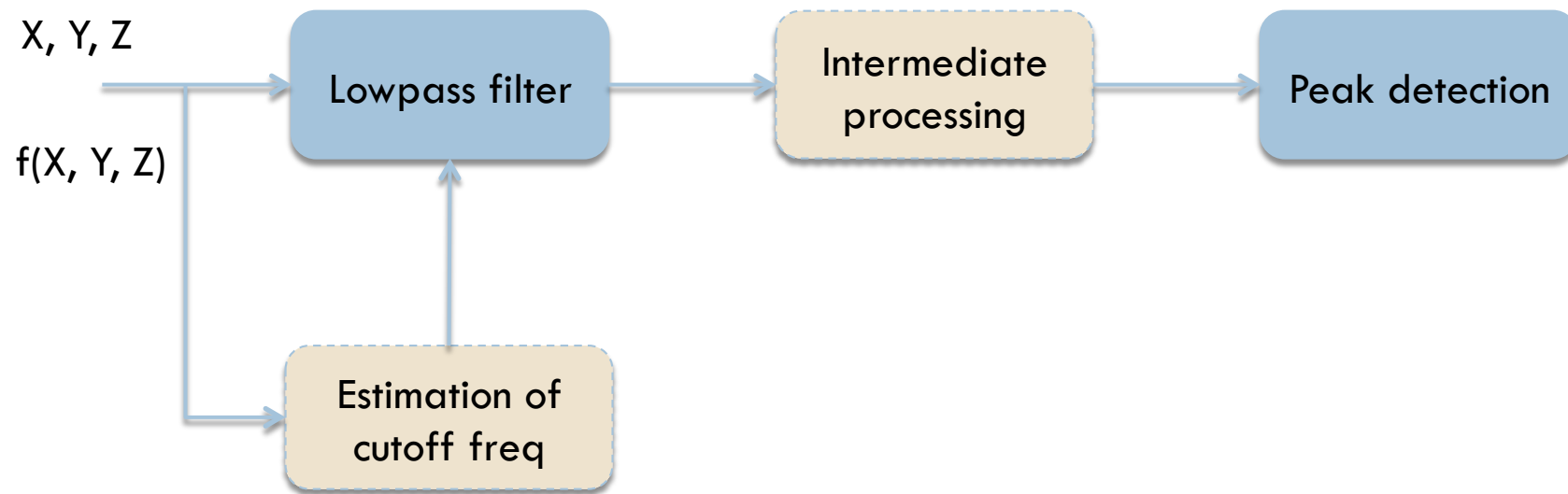


Pendulum Swing Leg



\*AD Kuo, JM Donelan, Dynamic Principles of Gait and Their Clinical Implications

# Step counting



# Stride length estimation

- Height based
  - ▣ Height x .413 (female)
  - ▣ Height x .415 (male)
- Speed related
  - ▣  $S = av^b$ , where  $a = 1.22 \pm 0.11$ ,  $b = 0.54 \pm 0.10^*$
- Estimated online
  - ▣ From height, length of leg, acceleration\*\*

\*Steven H. Collins and Arthur D. Kuo, Two independent contributions to step variability during over-ground human walking

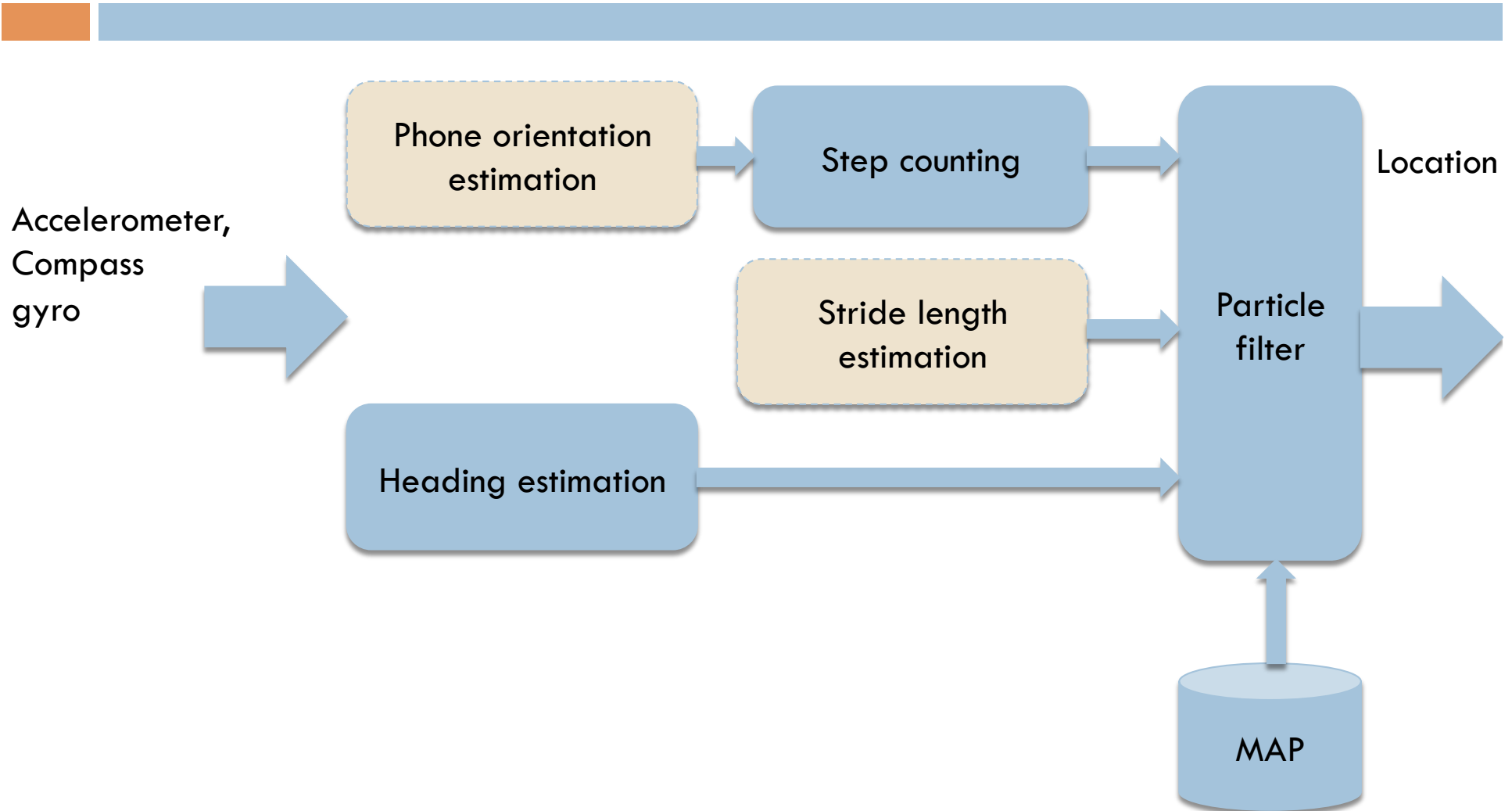
\*\*Valérie Renaudin\*, Melania Susi and Gérard Lachapelle, Step Length Estimation Using Handheld Inertial Sensors

# Issues



- ❑ Miscount (over/under-estimation) occurs
- ❑ Sensor placement (on the body) matters
- ❑ Stride length estimation may be inaccurate
- ❑ Healthy vs unhealthy subject
- ❑ Age and gender matters

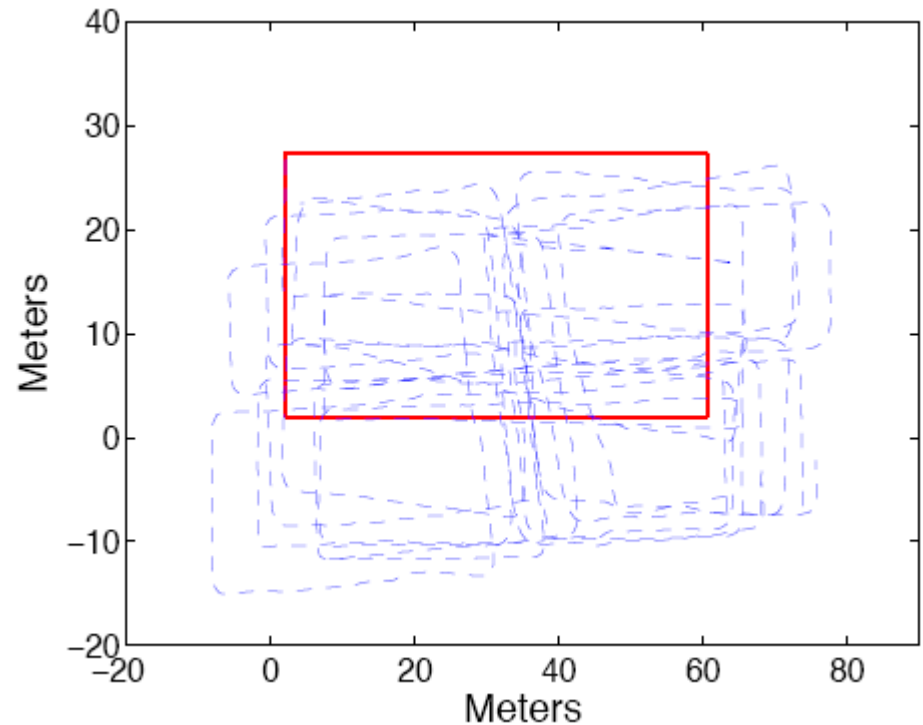
# Location estimation using inertial sensors





# Challenges with inertial sensing

- Noise is cumulative
- Need to start from a known location



# Hybrid approaches

