

# **Smartphone Sensors**

- Typical smartphone sensors today
  - accelerometer, compass, GPS, microphone, camera, proximity
- Use machine learning to classify sensor data



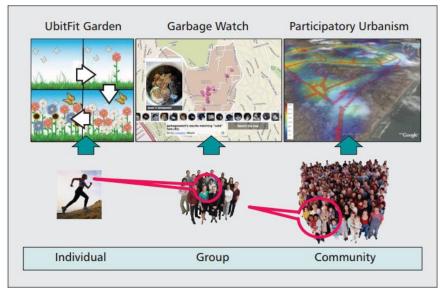
Future sensors?

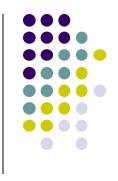
- Heart rate monitor,
- Activity sensor,
- Pollution sensor,
- etc



## **Mobile CrowdSensing**

- Mobile CrowdSensing: Sense collectively
- Personal sensing: phenomena pertain to individual
  - E.g: activity detection and logging for health monitoring
- Group: friends, co-workers, neighborhood
  - E.g. GarbageWatch recycling reports, neighborhood surveillance

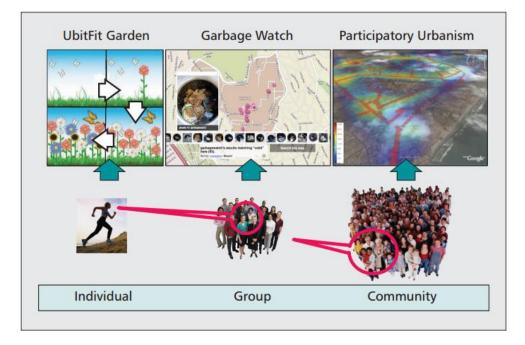




## **Mobile CrowdSensing**

#### • Community sensing (mobile crowdsensing):

- Large-scale phenomena monitoring
- Many people contribute their individual readings
- Examples: Traffic congestion, air pollution, spread of disease, migration pattern of birds, city noise maps





# **Mobile Crowd Sensing Types**

- Many people cooperate, share sensed values
- 2 types:
  - 1. Participatory Sensing: User enters sensed values (active involvement)
    - E.g. Comparative shopping: Compare price of toothpaste at CVS vs Walmart
  - Opportunistic Sensing: Mobile device automatically senses values (passive involvement)
    - E.g. Waze crowdsourced traffic







## Sense What?

- Environmental: pollution, water levels in a creek
- **Transportation:** traffic conditions, road conditions, available parking
- **City infrastructure:** malfunctioning hydrants and traffic signs
- Social: photoblogging, share bike route quality, petrol price watch
- Health and well-being:
  - Share exercise data (amount, frequency, schedule),
  - share eating habits and pictures of food





# **Smartphone Sensing Examples**

# **Personal Sensing**



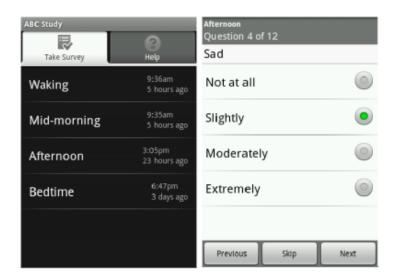
- Personal monitoring
- Focusing on user's daily life, physical activity (Khan et al. 404)

# AT NO WAY ----

# **Other Examples of Personal Participatory Sensing**

#### AndWellness

- "Personal data collection system"
- Active user-triggered experiences and surveys
- Passive recording using sensors
- UbiFit Garden
  - Uses smartphone sensors, real-time statistical modeling, and a personal, mobile display to encourage regular physical activity



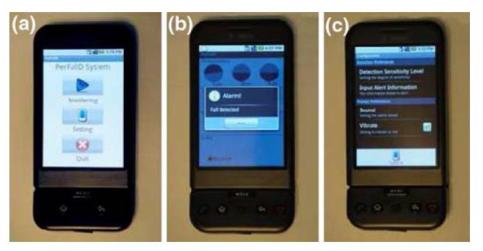






# **Personal Opportunistic Sensing**

- PerFalld
  - How It Works
    - Detects if someone falls using sensor
    - Starts a timer if it detects that someone fell
    - If individual does not stop timer before it ends, emergency contacts are called

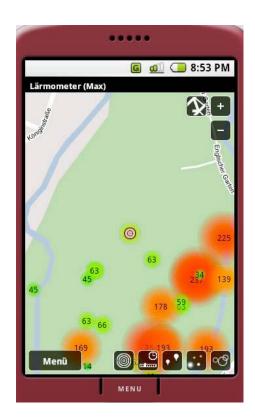


User interfaces in PerFallD: (a) bright, large virtual buttons on operating screen (b) clear alert window (c) simple, non-confusing preference screen

# **Public Sensing**

- Data is shared with everyone for public good
- Traffic
- Environmental
  - Noise levels
  - Air pollution





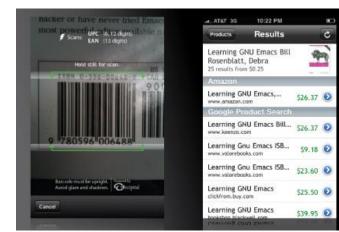




# **Public Participatory Sensing**

#### LiveCompare

- User-created database of UPCs and prices
- GPS and cell tower info used to find nearby stores
- PetrolWatch
  - Turns phone into fully automated dash-cam
  - Uses GPS to know when gas station is near







## **Public Participatory Sensing**

#### • Pothole Monitor

• Combines GPS and accelerometer

#### • Party Thermometer

• Asks you questions about parties



• Detects parties through GPS and microphone



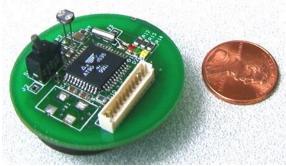
# Sensing with Smartphones vs Dedicated Sensors



- More resources: Smartphones have much more processing and communication power
- **Easy deployment:** Millions of smartphones already owned by people
  - Instead of installing sensors in road, we detect traffic congestion using smartphones carried by drivers
- Time-varying data: population of mobile devices, type of sensor data, accuracy changes often due to user mobility and differences between smartphones



# Smartphone Sensing vs Dedicated Sensors



# Sensing with Smartphones vs Dedicated Sensors



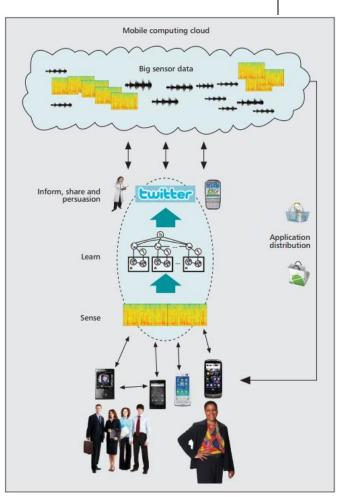
- Reuse of few general-purpose sensors: While sensor networks use dedicated sensors, smartphones reuse relatively few sensors for widerange of applications
  - E.g. Accelerometers used in transportation mode identification, pothole detection, human activity pattern recognition, etc
- Human involvement: humans who carry smartphones can be involved in data collection (e.g. taking pictures)
  - Human in the loop can collect complex data
  - Incentives must be given to humans



# **Smartphone Sensing Architecture**

## **Smartphone Sensing Architecture**

- Paradigm proposed by Lane *et al*
- Sense: Phones collect sensor data
- Learn: Information is extracted from sensor data by applying machine learning and data mining techniques
- Inform, share and persuasion: inform user of results, share with group/community or persuade them to change their behavior
  - Inform: Notify users of accidents (Waze)
  - Share: Notify friends of fitness goals (MyFitnessPal)
  - **Persuasion:** avoid speed traps (Waze)







# **Final Project Proposal**

# **Final Project Proposal**

- While working on projects 3 & 4, also brainstorm on final project
- Nov 2, Propose mobile/ubicomp app, solves WPI problem or Machine learning
- Proposals should include:
  - 1. Problem you intend to work on
    - Solve WPI/societal problem (e.g. walking safe at night)
    - Use at least 3 mobile/ubicomp components (e.g. location, sensor or camera)
    - If games, must gamify solution to real world problem

#### 2. Why this problem is important

- E.g. 37% of WPI students feel unsafe walking home
- Related Work: What prior solutions have been proposed for this problem

#### 4. Summary of envisioned mobile app (?) solution

1. E.g. Mobile app automatically texts users friends when they get home at night

# **Final Project Proposal**

- Can also do Machine learning project that classifies/detects analyzes a dataset of builds a real-time app to classify some human sensor data. E.g. Classifies
  - A speaker's voice to determine if nervous, sad, etc
  - A user's accelerometer data and recognizes their walk from 5-10 other people
  - A picture of a person's face and determines their mood
  - Data from a person's phone to measure their sleep duration or/and quality
  - Video of a person's face to detects their heart rate
  - A person's communication/phone usage patterns to detect their mood
- Also propose evaluation plan
  - E.g. Small user study to evaluate app
  - Machine learning performance metrics (e.g. classification accuracy, cross validation, etc)
- Can bounce ideas of me (email, or in person)
- Can change idea any time



## **Rubric: Grading Considerations**

### • Problem (10/100)

- How much is the problem a real problem (e.g. not contrived)
- Is this really a good problem that is a good fit to solve with mobile/ubiquitous computing? (e.g. are there better approaches?)
- How useful would it be if this problem is solved?
- What is the potential impact on the community (e.g. WPI students) (e.g. how much money? Time? Productivity.. Would be saved?)
- What is the evidence of the importance? (E.g. quote a statistic)

### Related Work (10/100)

- What else as been done to solve this problem previously
- Proposed Solution/Classification (10/100)
  - How good/clever/interesting is the solution?
  - How sophisticated and how many are the mobile/ubiquitous computing components (high level) proposed? (e.g. location, geofencing, activity recognition, face recognition, machine learning, etc)



## **Rubric: Grading Considerations**

- Implementation Plan + Timeline (10/100)
  - Clear plans to realize your design/methodology
  - Android modules/3<sup>rd</sup> party software used
  - Software architecture,
  - Screenshots (or sketches of UI), or study design + timeline

#### Evaluation Plan (10/100)

- How will you evaluate your project.
- E.g. small user studies for apps
- Machine learning cross validation, etc
- 50 more points allotted for your slides + presentation





# **The Rest of the Class**

## The Rest of this class

#### • Part 1: Course and Android Introduction

- Introduce mobile computing, ubiquitous Computing, Android,
- Basics of Android programming, UI, Android Lifecycle
- Part 2: Mobile and ubicomp Android programming
  - mobile Android components (location, Google Places, maps, geofencing)
  - Ubicomp Android components (camera, face detection, activity recognition, etc)

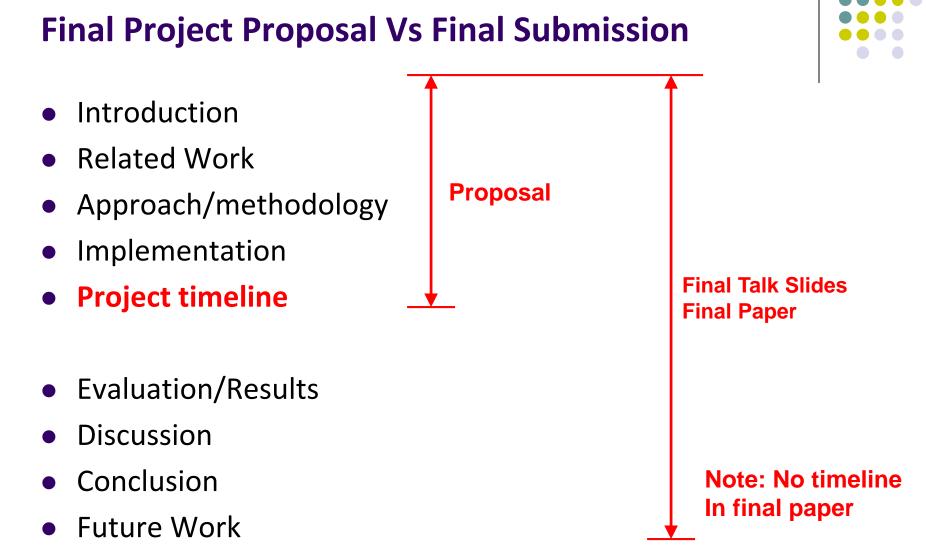
#### • Part 3: Mobile Computing/Ubicomp Research

- Machine learning (classification) in ubicomp
- Ubicomp research (smartphone sensing examples, human mood detection, etc) using machine learning
- Mobile computing research (app usage studies, energy consumption, etc)





## **Final Project: Proposal Vs Final Submission**



### References



- A Survey of Mobile Phone Sensing. Nicholas D. Lane, Emiliano Miluzzo, Hong Lu, Daniel Peebles, Tanzeem Choudhury, Andrew T. Campbell, In IEEE Communications Magazine, September 2010
- Mobile Phone Sensing Systems: A Survey, Khan, W.; Xiang, Y.; Aalsalem, M.; Arshad, Q.; , Communications Surveys & Tutorials, IEEE , vol.PP, no.99, pp.1-26