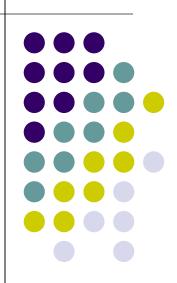
CS 528 Mobile and Ubiquitous Computing Lecture 7a: Other Android UbiComp Components, Tech Talk, Final Project Proposal & Smartphone Sensing

### **Emmanuel Agu**





# What other Android APIs may be useful for Mobile/ubicomp?

### **Speaking to Android**

http://developer.android.com/reference/android/speech/SpeechRecognizer.html https://developers.google.com/voice-actions/

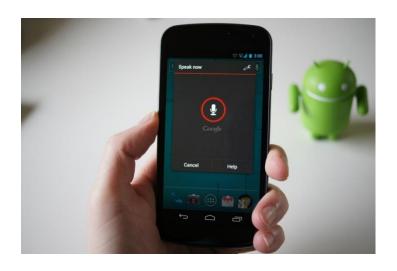
#### • Speech recognition:

Accept inputs as speech (instead of typing) e.g. dragon dictate app?

Speech

to text

- Note: Requires internet access
- Two forms
  - 1. Speech-to-text
    - Convert user's speech to text. E.g. display voicemails in text
  - 2. Voice Actions: Voice commands to smartphone (e.g. set alarm)



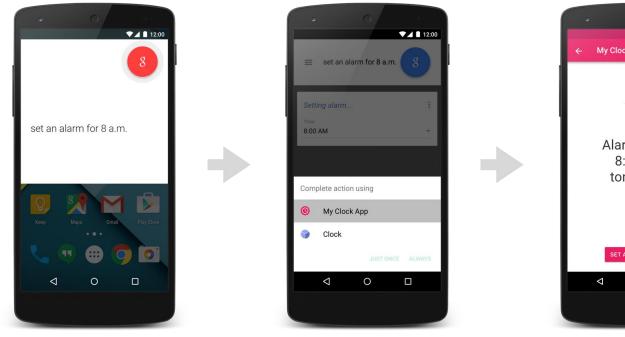


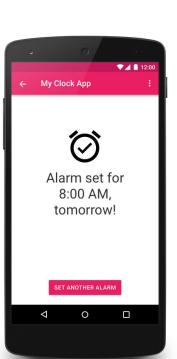


# **Google Voice Actions**

https://developers.google.com/voice-actions/

• E.g. Tell Google to set an alarm





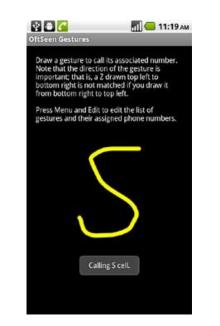


### Gestures

https://developer.android.com/training/gestures/index.html http://www.computerworld.com/article/2469024/web-apps/android-gestures--3-cool-ways-to-control-yourphone.html

- Gesture: Hand-drawn shape on the screen
- Example uses:
  - Search your phone, contacts, etc by handwriting onto screen
  - Speed dial by handwriting first letters of contact's name
  - Multi-touch, pinching







### **More MediaPlayer & RenderScript**

http://developer.android.com/guide/topics/renderscript/compute.html https://developer.android.com/reference/android/media/MediaRecorder

- MediaRecorder is used to record audio
  - Manipulate raw audio from microphone/audio hardware, PCM buffers
    - E.g. if you want to do audio signal processing, speaker recognition, etc
    - **Example:** process user's speech, detect emotion, nervousness?
  - Can playback recorded audio using MediaPlayer

#### RenderScript

- High level language for computationally intensive tasks/GPGPU,
- Can be used to program phone CPU, GPU in a few lines of code
- Use Phone's Graphics Processing Unit (GPU) for computational tasks
- Useful for heavy duty tasks. E.g. image processing, computational photography, or computer vision



# **Wireless Communication**

http://developer.android.com/guide/topics/connectivity/bluetooth.html http://developer.android.com/reference/android/net/wifi/package-summary.html

#### Bluetooth

- Discover, connect to nearby bluetooth devices
- Communicating over Bluetooth
- Exchange data with other devices

#### • WiFi

- Scan for WiFi hotspots
- Monitor WiFi connectivity, Signal Strength (RSSI)
- Do peer-to-peer (mobile device to mobile device) data transfers







# **Wireless Communication**

http://developer.android.com/guide/topics/connectivity/nfc/index.html

#### • NFC:

- Contactless, transfer small amounts of data over short distances
- Applications: Share spotify playlists, Google wallet
- Android Pay
  - Store debit, credit card on phone
  - Pay by tapping terminal









# **Telephony and SMS**

http://developer.android.com/reference/android/telephony/package-summary.html http://developer.android.com/reference/android/telephony/SmsManager.html

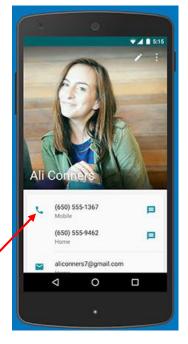
#### • Telephony:

- Initiate phone calls from within app
- Access dialer app, etc

#### • SMS:

- Send/Receive SMS/MMS from app
- Handle incoming SMS/MMS in app

Dialer





SMS



### **Google Play Services: Nearby Connections API**

https://developers.google.com/nearby/connections/overview

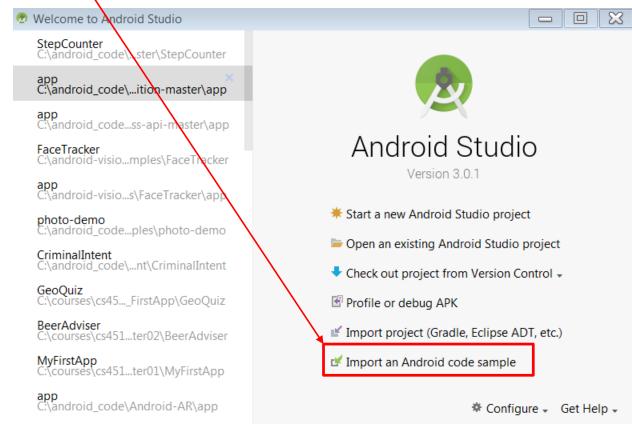
- Peer-to-peer networking API, allows devices communicate over a LAN
- Allows one device to serve as host, advertise
- Other devices can discover host, connect, disconnect
- Use case: Multiplayer gaming, shared virtual whiteboard
- Good tutorial by Paul Trebilcox-Ruiz

https://code.tutsplus.com/tutorials/google-play-services-using-the-nearby-connections-api--cms-24534?\_ga=2.245472388.1231785259.1517367257-742912955.1516999489



### **Google Android Samples**

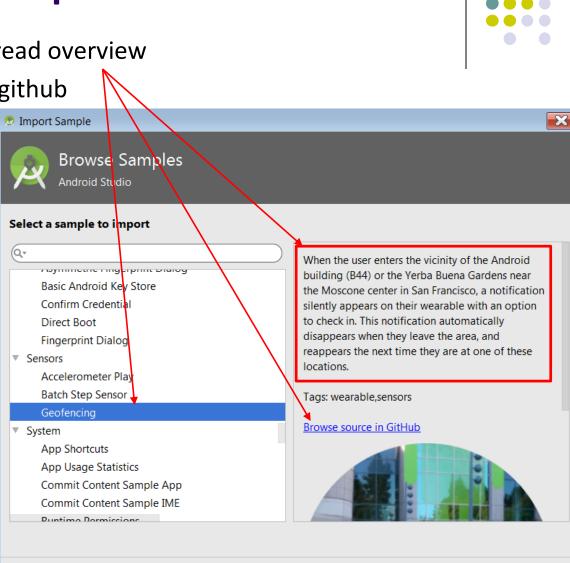
- Android Studio comes with many sample programs
- Just need to import them





### **Google Android Samples**

- Can click on any sample, read overview
- Source code available on github
- Tested, already working



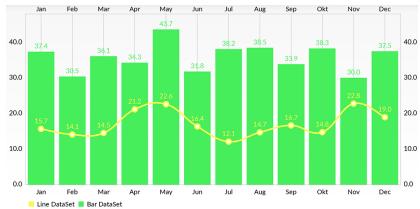
Next

Cancel

# **Other 3rd Party Stuff**

http://web.cs.wpi.edu/~emmanuel/courses/ubicomp\_projects\_links.html https://developer.qualcomm.com/software/trepn-power-profiler

• MPAndroid: Add charts to your app



- Trepn: Profile power usage and utilization of your app (CPU, GPU, WiFi, etc)
  - By Qualcomm

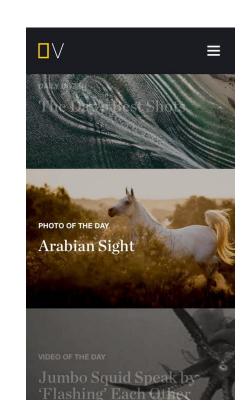




# **Other 3rd Party Stuff**

http://web.cs.wpi.edu/~emmanuel/courses/ubicomp\_projects\_links.html

- **Programmable Web APIs:** 3<sup>rd</sup> party web content (e.g RESTful APIs) you can pull into your app with few lines of code
  - Weather: Weather channel, yahoo weather
  - Shared interests: Pinterest
  - Events: Evently, Eventful, Events.com
  - Photos: flickr, Tumblr
  - Videos: Youtube
  - Traffic info: Mapquest traffic, Yahoo traffic
- E.g. National Geographic: picture of the day







# **Student Presentation: Mobile Technologies**

# **Talk: Mobile Technology**



- GROUP to research, master and present on any TWO mobile technologies.
- Your talk should cover:
  - Background on the technology (tell a story about its history, etc)
  - Specific problems it's designed to solve
  - Typical example use case: When is it typically used?
  - Real world examples of where it is being used. E.g. by XYZ company for ABC
  - Overview of how it works?
  - Code snippet: Walk through a simple program that uses the technology including how to compile it and how to run it.

# **Talk on Mobile Technology**



- Submit talk slides + working code
- To avoid duplicate presentations, each group email me their TWO topics by October 31, 2019
- This talk is 15% of your grade!
- The idea is to become expert, help any groups that need your help on that technology

# Talk on Mobile Technology

- Mobile programming/develpment:
  - Kotlin
  - iPhone development
  - 3rd part libraries: E.g. Xamarin
  - Mobile web programming
  - PhoneGap
  - AppInventor
  - Mobile game development tools: Unity,
- Machine/Deep Learning:
  - Deep Learning/machine learning in Android: Tensorflow, etc
  - Mobile machine/deep learning support in MATLAB
  - Keras support for Android Deep learning
  - Neural Networks API (NNAPI)



# **Talk on Mobile Technology**

- More Google APIs (that could be used by mobile devices):
  - Analytics
  - Google Drive
  - Google Fit
  - Google Cast
  - Advertising: E.g. Adwords, Admobs
- More Android APIs:
  - Firebase (database, messaging, authentication, analytics, etc)
  - Speaking to Android (Speech recognition, Voice Actions)
  - Renderscript
  - Media Recorder
  - Wireless Communication: Bluetooth, WiFi, NFC, etc
  - Android Pay
  - Telephone/SMS
  - Nearby Connections API
  - Depth Sensing: Project Tango
  - Augmented Reality: ARtoolkit, vuforia, EasyAR





# **Final Project Proposal**

# **Final Project Proposal**

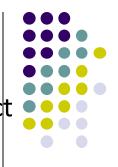
- While working on projects 3 & 4, also brainstorm on final project
- Oct 31, Propose mobile/ubicomp app, solves WPI problem or Machine learning
- Proposals should include:
  - 1. Problem you intend to work on
    - Solve WPI/societal problem (e.g. walking safe at night)
    - Use at least 3 mobile/ubicomp 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
- Can use existing smartphone datasets online
- See project difficulty points rubric
- Also propose evaluation plan
  - E.g. Small user study to evaluate app.
  - Can trade with another team: you review our app, we review yours
  - Machine learning performance metrics (e.g. classification accuracy, cross validation, etc)
- Can bounce ideas off 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 are the mobile/ubiquitous computing components (high level) used? (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

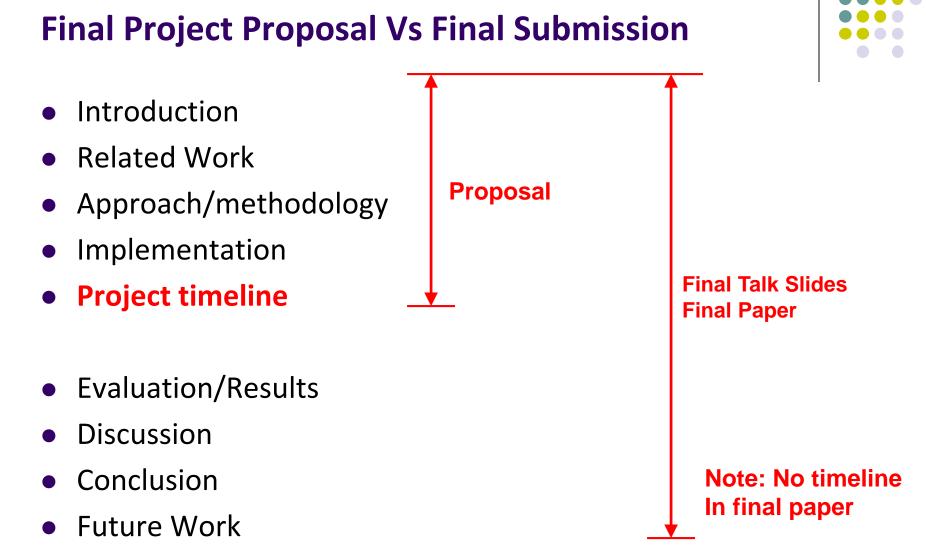
#### Evaluation Plan (10/100)

- 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





# Final Project: Proposal Vs Final Submission (Presentation + Paper)





# **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)





# **Smartphone Sensing**

# **Smartphone Sensors**

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



Future sensors?

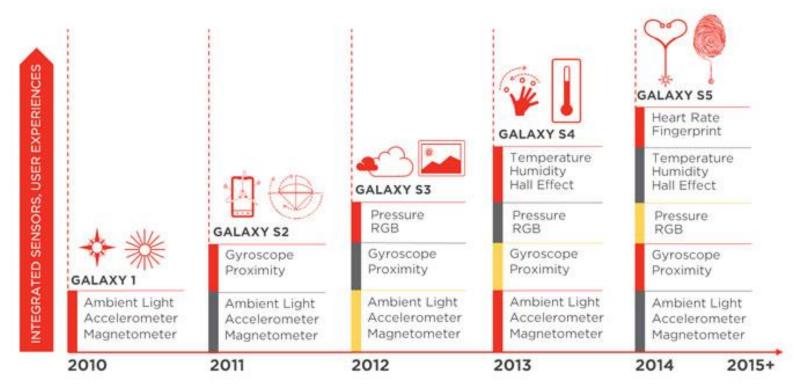
- Heart rate monitor,
- Activity sensor,
- Pollution sensor,
- etc



#### **Growth of Smartphone Sensors**

• Every generation of smartphone has more and more sensors!!







#### Image Credit: Qualcomm

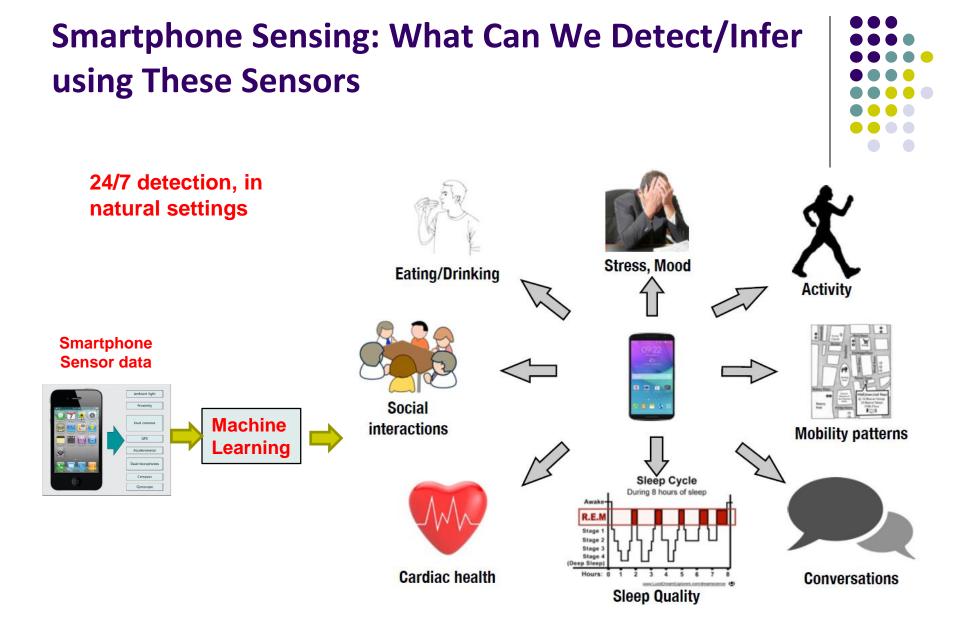


Image Credit: Deepak Ganesan, UMass

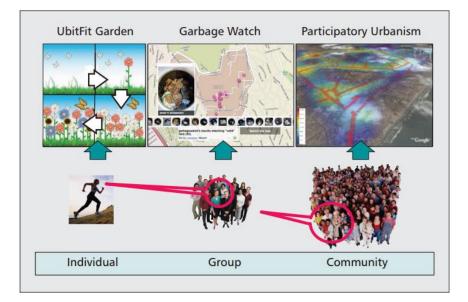
### Sense What?

- 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



### **Mobile CrowdSensing**

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

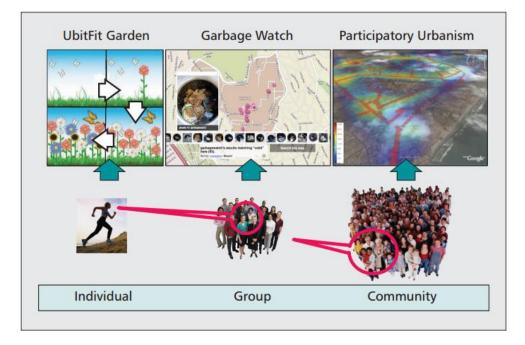




### **Mobile CrowdSensing**

#### • 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:
  - Participatory Sensing: User manually 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









## **Smartphone Sensing Examples**

## **Personal Sensing**

- Personal monitoring
- Focusing on user's daily life, physical activity (Khan et al.)
- Basically like Fitbit on your phone

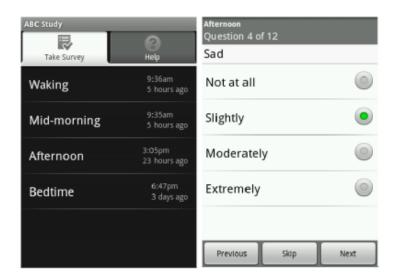
# AT NO WAY ----



## **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 tracking, statistical modeling, and a personal, mobile display to encourage regular physical activity



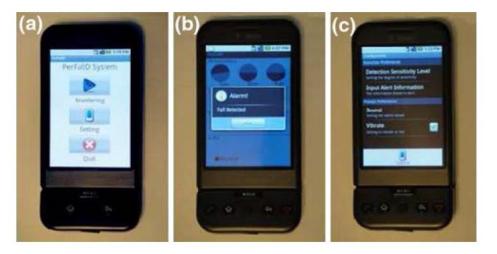






## **Personal Opportunistic Sensing**

- PerFalld
  - Detects if user 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

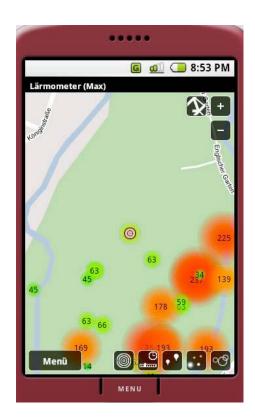


User interfaces in PerFallD: (a) bright, large virtual buttons on operating screen (b) clear alert window (c) simple, non-confusing preference screen

## **Public Sensing**

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





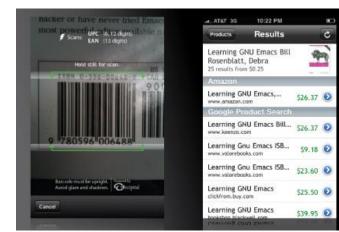




## **Public Participatory Sensing**

#### 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





## Smartphone Sensing vs Dedicated Sensors

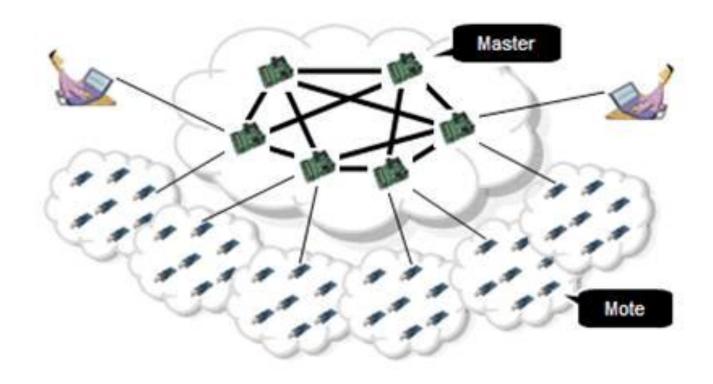


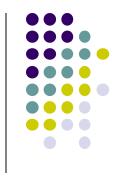
VS



## Background: Wireless Sensors for Environment Monitoring

- Embedded in room/environment
- Many sensors cooperate/communicate to perform task
- Monitors conditions (temperature, humidity, etc)
- User can query sensor (What is temp at sensor location?)





## 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
  - Makes maintance easier. E.g. owner will charge their phone promptly
- Time-varying data: population of mobile devices, type of sensor data, accuracy changes often due to user mobility and differences between smartphones

## Sensing with Smartphones vs Dedicated Sensors



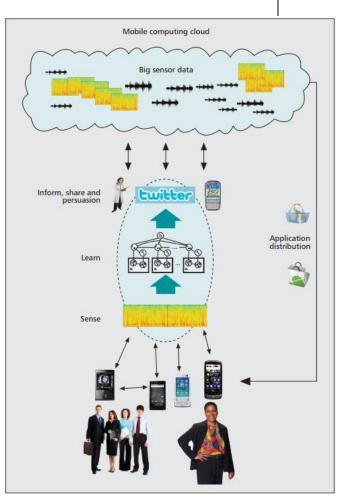
- Reuse of few general-purpose sensors: While sensor networks use dedicated sensors, smartphones reuse relatively few sensors for widerange of applications
  - E.g. Accelerometers used in transportation mode identification, pothole detection, human activity pattern recognition, etc
- Human involvement: 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)





#### 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