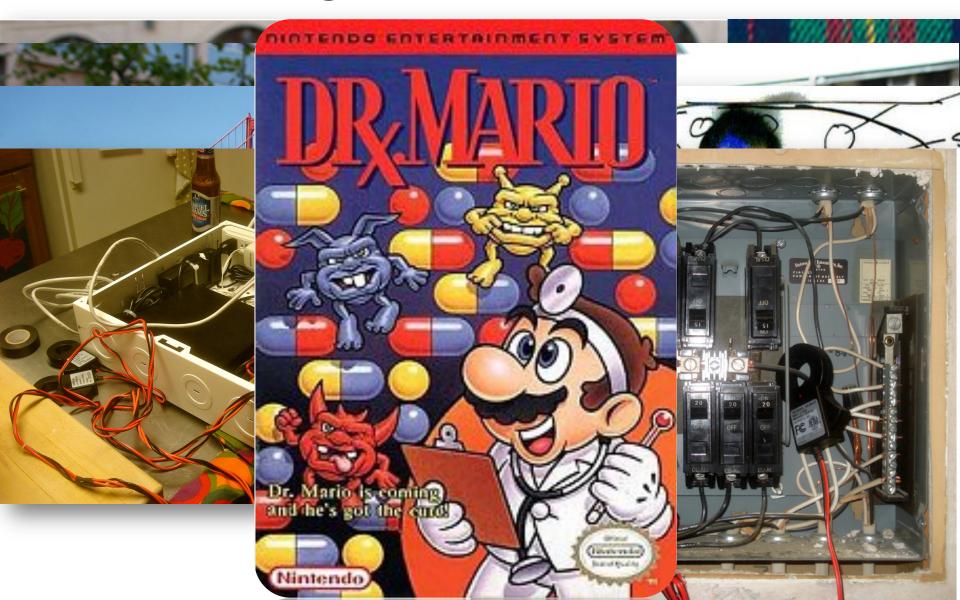
Lecture #1: Introduction

Mario Bergés Professor

12-778: Sensors, Circuits and Data Interpretation/Mgmt. for CEE

Civil & Environmental ENGINEERING Carnegie Mellon

Who Google thinks I Am



About me

Born and raised in



Became a Civil Engineer there







, to merge my degree with



Now doing research on smart infrastructure



About the 7 of you

- CEE
- Architecture
- ___ Professionals?
- ___ Students?



6

1

Student Introduction

- Good opportunity to get to know your classmates
- Your program and year (undergraduate or graduate)
- Interest for this class (why you are taking this course)
- Future plans (going to industry or continuing for Ph.D.)
- Other information that you like to share with others



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WHY ARE YOU HERE?

Let's read the syllabus

Objectives

- Understand:
 - Physical Principles of Sensors
 - Fundamentals of Data Acquisition
 - Basic Signal Processing
 - Measurement Uncertainty and Errors
 - Data Management

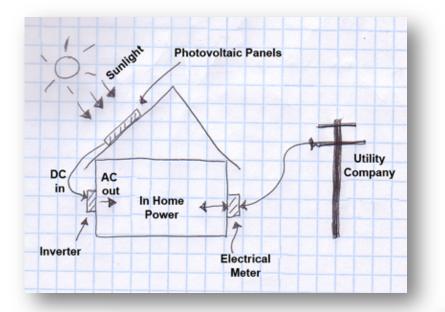


Objectives

- Acquire skills to develop:
 - Prototype sensing systems using Raspberry
 Pis and Python
 - Simple "Internet of Things" applications



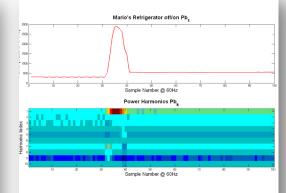
Objectives





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TUESDAY		THURSDAY	
Aug 29th	1	Aug 31st	2
THEORY: Introduction		THEORY: Principles of Sensors	
Sep 5th	3	Sep 7th	4
PRACTICE: Preparing Your Instrumentation Environments HW1: Out		PRACTICE: Sensing with the Rapsberry Pi Pico	
Sep 12th	5	Sep 14th	6
THEORY: Electronic Devices and Circuit Theory		THEORY: Fundamentals of Data Acquisition I	

Sep 19th 7	Sep 21st 8
THEORY: Fundamentals of Data Acquisition II	THEORY: Resistive Sensors HW1: Due
•	HW2: Out
Sep 26th 9	Sep 28th 10
THEORY: Resistive and Capacitive Sensors	PRACTICE: Practical Signal Conditioning and Sampling
Oct 3rd 11	Oct 5th 12
PRACTICE: Practical Measurements with an MCU	THEORY: Fourier Analysis

Oct 10th	13	Oct 12th	14
THEORY: Fourier Analysis II		THEORY: Linear Systems Theory HW2: Due	
Oct 17th	15	Oct 19th	16
NO CLASS: Fall Break		NO CLASS: Fall Break	
Oct 24th	17	Oct 26th	18
THEORY: Measurement Noise and Errors HW3: Out		Project Proposal Presentations	

TUESDAY		THURSDAY	
Oct 31st	19	Nov 2nd	20
THEORY: RECORDED LECTURE Uncertainty Quantification and Propagation		THEORY: Networked Sensors: IoT a WSNs PRACTICE: Project Assistance	nd
Nov 7th	21	Nov 9th	22
NO CLASS: Democracy Day		THEORY: Time-series Data: Representation, Processing HW3: Due HW4: Out	
Nov 14th	23	Nov 16th	24
THEORY: Set Theory and Entity Relationship Diagrams		THEORY: Relational Databases	

Nov 21st 25	Nov 23rd 26	
THEORY: SQL	NO CLASS: Thanksgiving Break	
Nov 28th 27	Nov 30th 28	
THEORY: Database Design and Normalization HW4: Due	THEORY: Lessons Learned from Computer Systems Research: Planning Monitoring Campaigns	
Dec 5th 29	Dec 7th 30	
PRACTICE: No Lecture: Assistance for Project	PRACTICE: No Lecture: Assistance for Project	

Schedule of Assignments

- 4 homework assignments
 - HW1: out on 9/5, due on 9/21
 - HW2: out on 9/21, due on 10/12
 - HW3: out on 10/24, due on 11/9
 - HW4: out on 11/9, due on 11/28
- 1 Project Proposal due 10/26



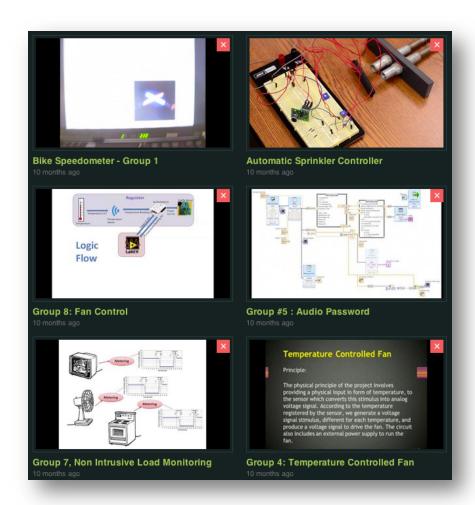
Projects

- Presentation (Video)
- Report (Github Pages)
- Themes:
 - M: Velocity of a Fan, S: Light Intensity Sensor
 - M: Audio Password, S: Microphone
 - M: Distance Traveled, S: Accelerometer
 - M: Temperature Vocoder, S: Temp. Sensor

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Past Projects

- Project Reports (2019)
- <u>Vimeo Group (2014)</u>
- <u>Vimeo Group (2013)</u>
- <u>Wiki Reports</u>



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Grading

Grading:

Assignments	
Final Project (Report)	$\dots \dots 20\%$
Final Project (Presentation)	
Final Exam	

Class Policy



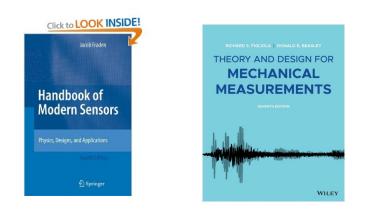




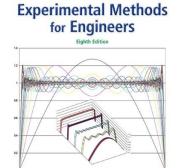
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Readings

Recommended books:



• Other references: - Will be listed on Canvas



0.5

6.2 0.3 0.4 J.P. Holman 0.8

Click to LOOK INSIDE Measurement nstrumentation Theory and Application Alan S. Morris H



Discussion

• What is a measurement?



Measurement Science



Measurement: the estimation of the magnitude of some attribute of an entity.



Usually involves a measuring instrument: ruler, scale, thermometer, etc.



Measurements are not perfect, thus they have three attributes to describe this imperfection: Estimate

Error bound

Confidence level



The history of measurement science is tightly linked to the history of science itself.



Measuring Instruments

Measuring instruments allow us to assess the relationship between:

- an attribute of an item under study (e.g. height of a tree)
- a referenced unit of measurement (e.g. a meter)

Traditional instruments:

- Hardware based
- Fixed, limited versatility
- Specific to a stimulus
- Example: oscilloscopes

Virtual Instruments:

- Hardware/software
- Versatile
- Less expensive

Sensors

- Broad definition: "devices that receive and respond to a signal or stimulus"
 - Human eye
 - A trigger in a pistol
- For our scope: "devices that receive a signal or stimulus and responds with an electrical signal"
 - Where electrical signal means: can be amplified, modified and channeled by electronic devices.
 - Examples:
 - Video camera
 - Microphone
 - Accelerometer on your iPhone



Actuators

- For our purpose:
 - They are the inverse of sensors
 - They convert an electrical signal into a physical stimulus
- Example:
 - Sound speakers
 - Electric Motors
 - Relay



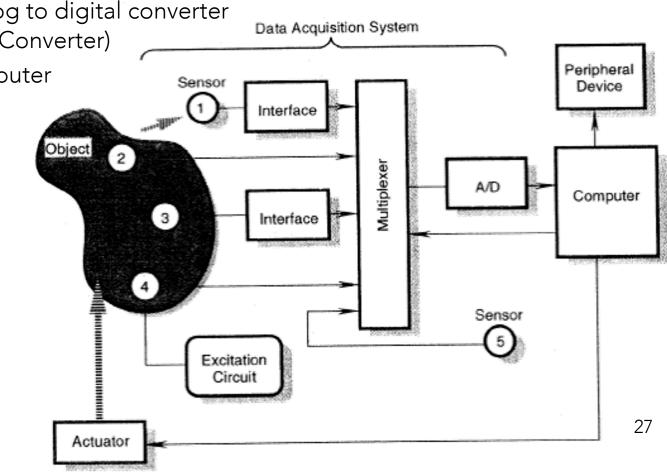
Transducers

- In broad terms:
 - A converter of one type of energy into another.
- For our purposes:
 - A term that encompasses both sensors and actuators.
- Examples:
 - The sound speaker is a transducer and can be used both as an actuator (a speaker) or a sensor (a microphone).

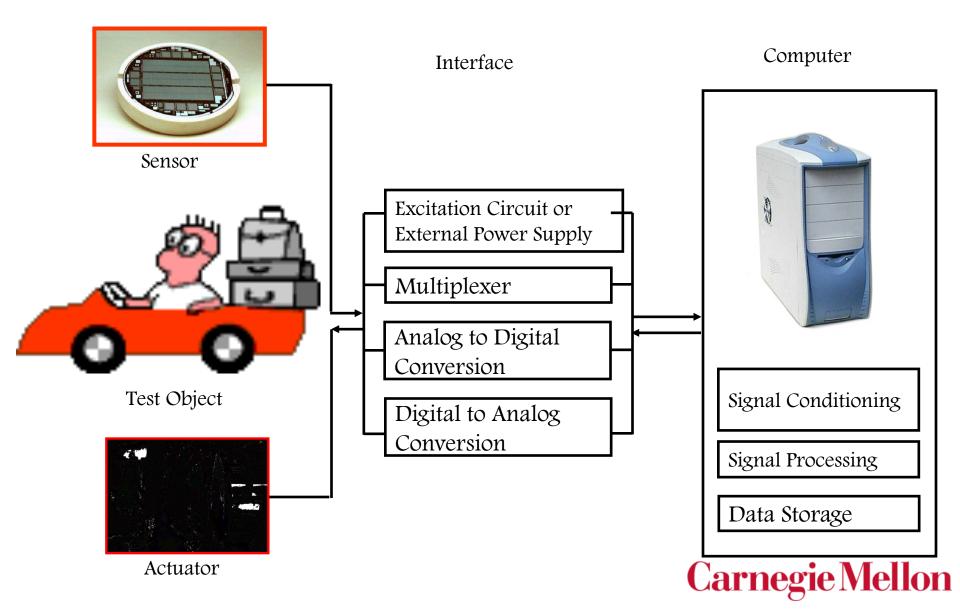
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Data Acquisition Systems

- Components:
 - Sensors
 - Actuators
 - Multiplexer (MUX)
 - Analog to digital converter (A/D Converter)
 - Computer



Overview of Data Acquisition Course



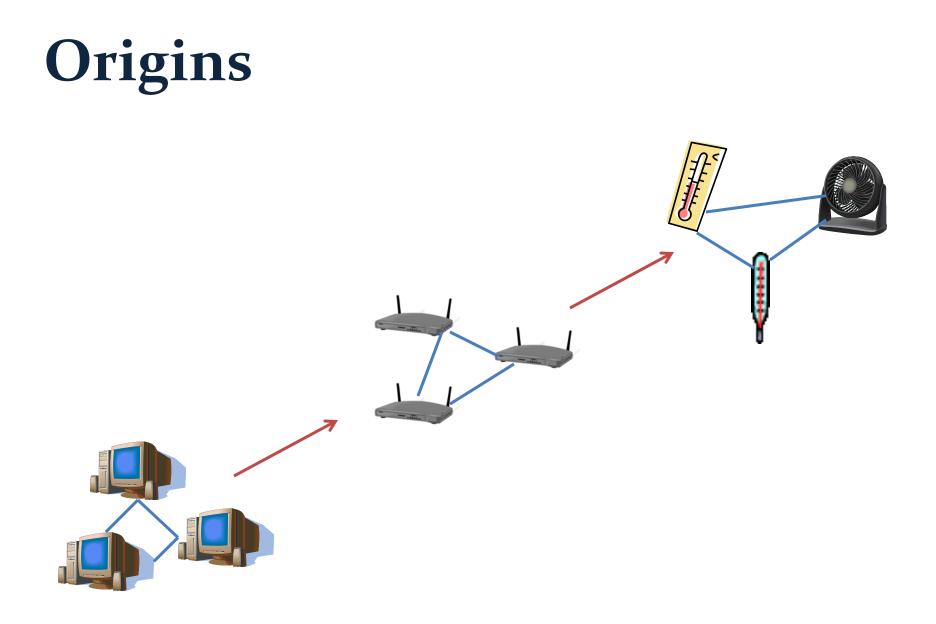
Data Acquisition System

- Illustration Example:
 - Imagine a a car door monitoring arrangement.
 - Every door has a sensor detecting open or closed position.
 - Signals from all doors are in a digital format (ones or zeros).
 - A microprocessor identifies which door is open, and sends an indicating signal to the peripheral devices (dashboard display, sound alarm).
 - The driver (actuator) gets the message and acts on the object (closes the door).
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Sensor Andrew

A Living Laboratory for Infrastructure Sensing Technologies

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More and more *Physical* data is becoming available...







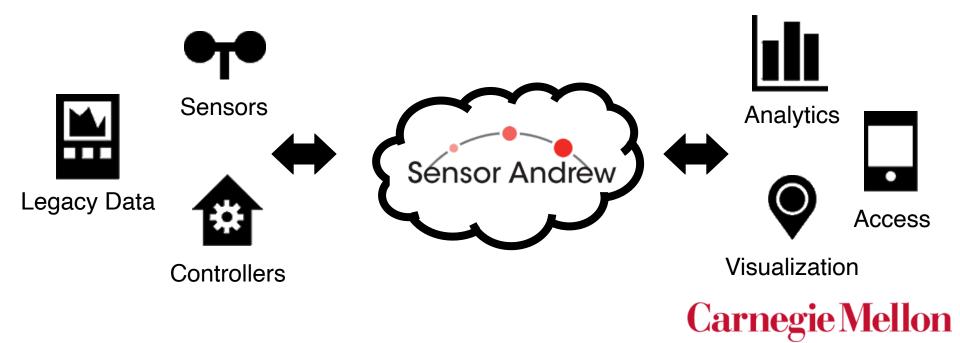




How do we make high-quantity data *high-quality* data?

Sensor Andrew

- Infrastructure to help connect the *virtual* and *physical world*
- Access, store, control, describe and search sensor data while maintaining security and privacy
- Internet-scale performance and Extensibility



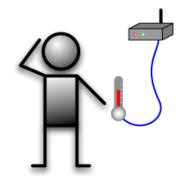
What is Sensor Andrew?







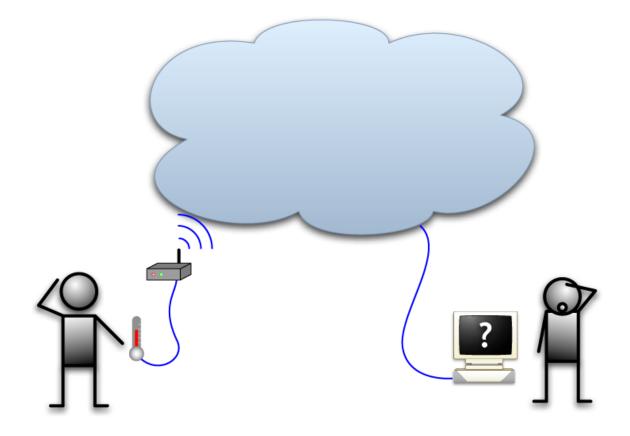
What is Sensor Andrew?





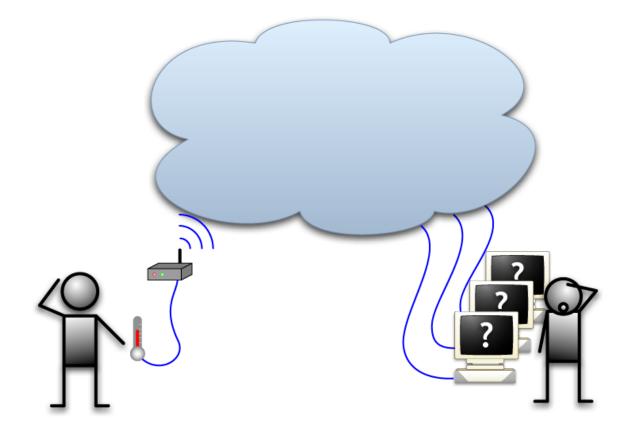


What is Sensor Andrew?

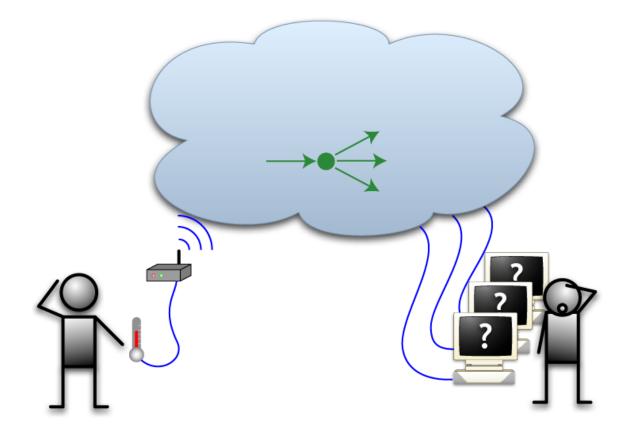


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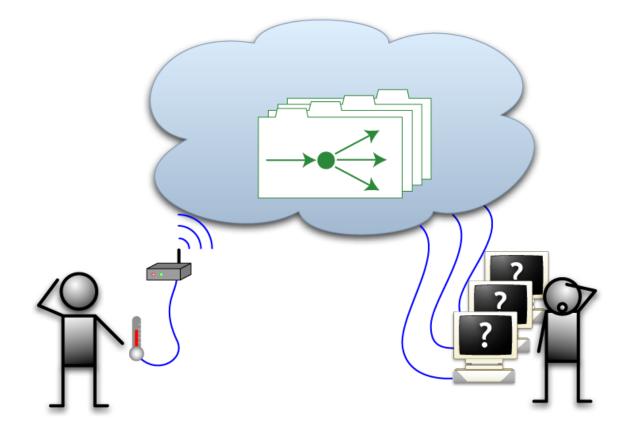
What is Sensor Andrew?



What is Sensor Andrew?



What is Sensor Andrew?



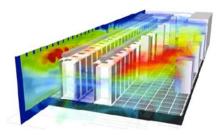
Motivation



Sensor Andrew Projects



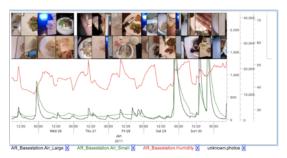
Smart Home



Data Center Energy



Building Automation Systems



Body Track



Wireless Protocols



Campus Facilities



Water Quality Monitoring



People Tracking



Air Quality Tracking

Human-in-the-loop Sensing and Control for Commercial Building Energy Efficiency and Occupant Comfort

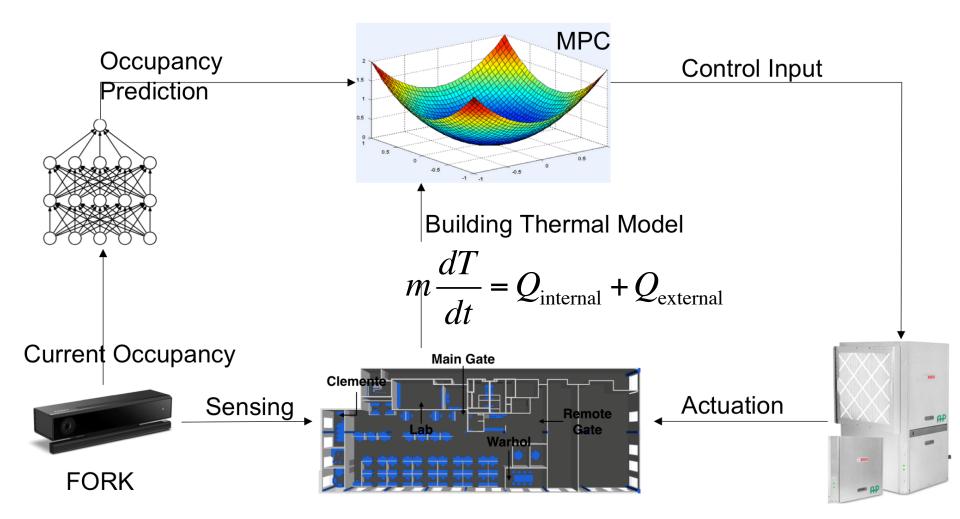
What if we could incorporate room-level occupancy counts and thermal comfort preferences into our building controls?

What if we could do it at a low cost and with high accuracy?

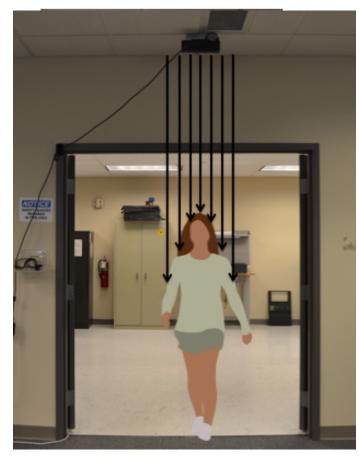


and a state of the state of the

Human-in-the-loop Sensing and Control for Commercial Building Energy Efficiency and Occupant Comfort



Original Platform





(b)



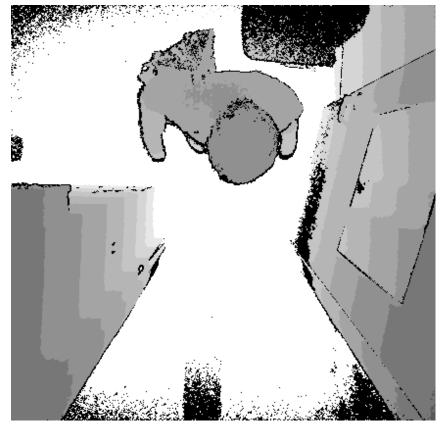
(c)

(a) Placement of a Kinect sensor on ceiling tile.

(b) Kinect sensor

(c) Embedded computer: Odroid-XU4

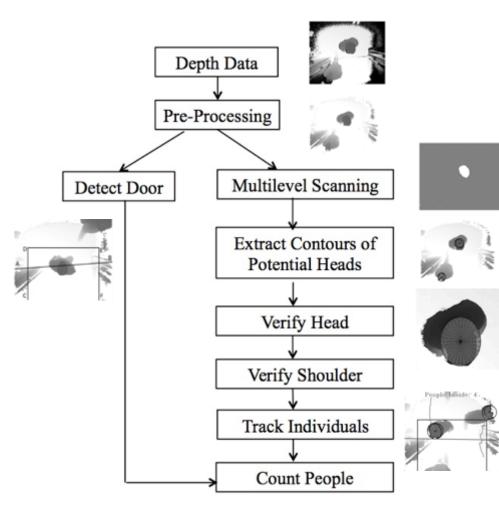
Depth Map

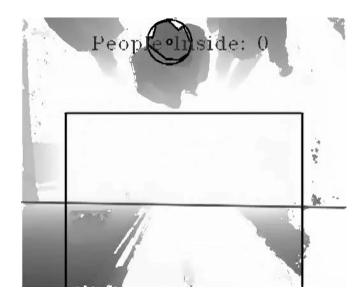


RGB Data



Algorithm (v1) Overview

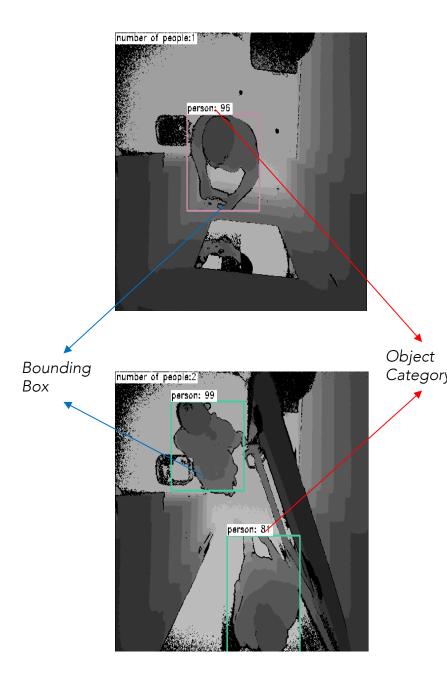


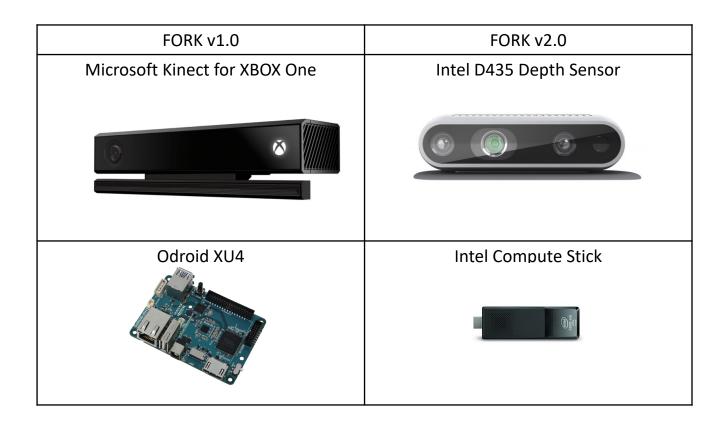


Algorithm (v2)

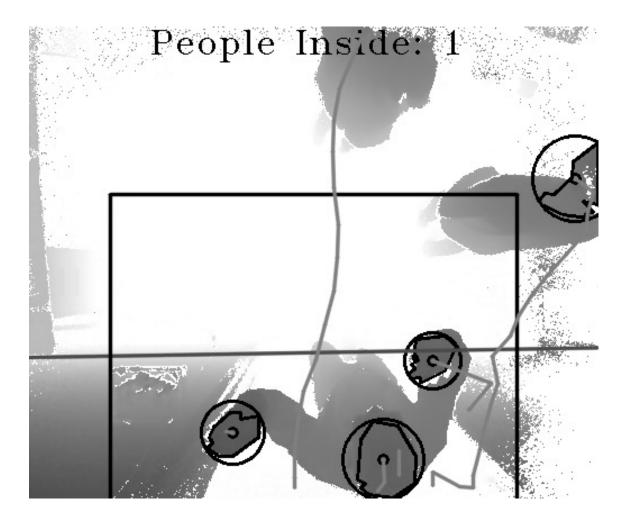
Approach: Faster RCNN (Region Based Convolutional Networks)

- Training stage
 - Inputs: Depth Image data
 - Targets: Bounding boxes and object category
 - (Person Background)
- Testing stage
 - Input: Depth Image data (unseen)
 - Output: Bounding box and object category
 - Number of people in frame
- Features are automatically learned.
- People are first detected and then classified.
- Network can be trained to detect additional objects like backpacks, computers, etc.





Hands detected as heads





INFERLab

Intelligent Infrastructure Research Laboratory

Rethinking and redesigning our built environment to:

- improve its operational efficiency
- increase its resilience, adaptiveness and autonomy.

http://inferlab.org/

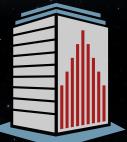


The End OUESTIONS?



@bergesmario







Intelligent Infrastructure Research Laboratory

