



# Understanding Cities through Measurement and *Embedded Intelligence*: The Array of Things

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Argonne National Laboratory

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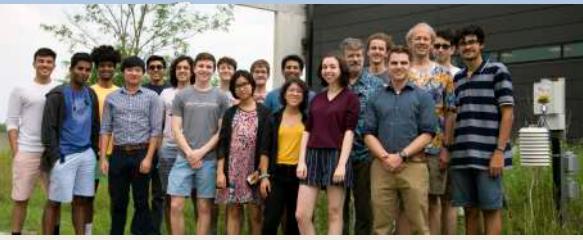
Collaborators:  
Charlie Catlett (ANL/UC)  
Pete Beckman (ANL/NLU)  
Rajesh Sankaran (ANL)  
Mike Papka (ANL/NIU)  
Kathleen Cagney (UC)  
Mark Potosnak (DePaul)  
Doug Pancoast (SAIC)  
Kate Kusiak Galvin (UC)  
Dan Work (Vanderbilt)

# Outline

- Machine Learning and Urban Science
- The Array of Things Project
- The Waggle Platform
- Survey of Machine Learning Opportunities

*This talk is a combination of several of mine and Pete Beckman's, covering work by dozens of collaborators.*

Summer 2018 Student Team



2018 Array of Things User Workshop



# Machine Learning and Urban Science

*The Array of Things and Waggle projects  
were developed in the context of scientific  
questions about cities and other  
environments.*

MacArthur  
Foundation

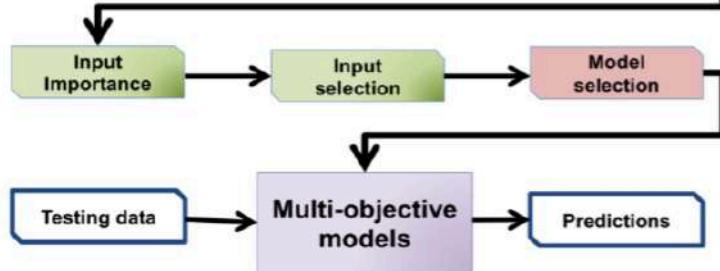
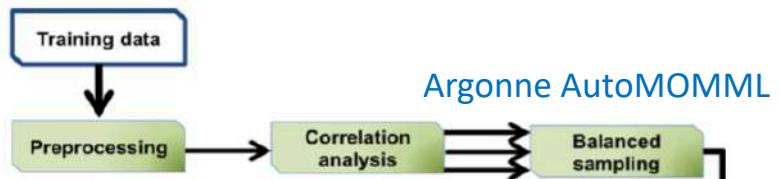
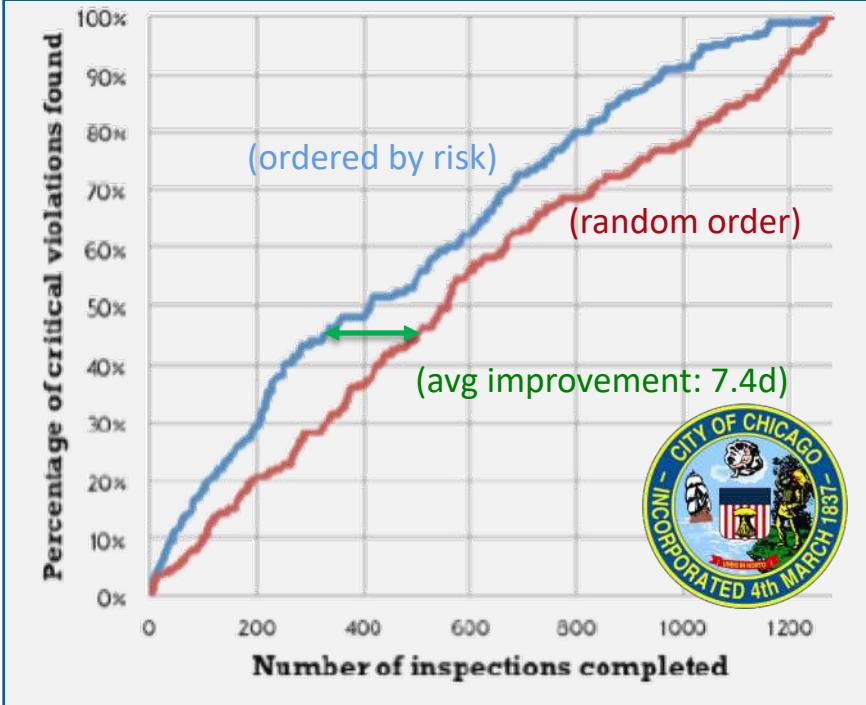


THE UNIVERSITY OF  
**CHICAGO**

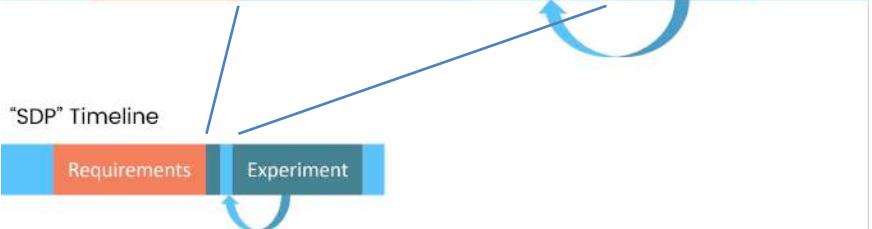
Argonne  
NATIONAL LABORATORY



# Improving Food Safety



Typical Research Project Timeline



P. Balaprakash, A. Tiwari, S. M. Wild, L. Carrington, and P. D. Hovland, "AutoMOMML: Automatic Multi-objective Modeling with Machine Learning," in International Conference on High Performance Computing, 2016, pp. 219-239: Springer.

Tom Schenk (City of Chicago); Allstate Insurance; Sven Leyffer, Stefan Wild, Prasanna Balaprakash (Argonne MCS)

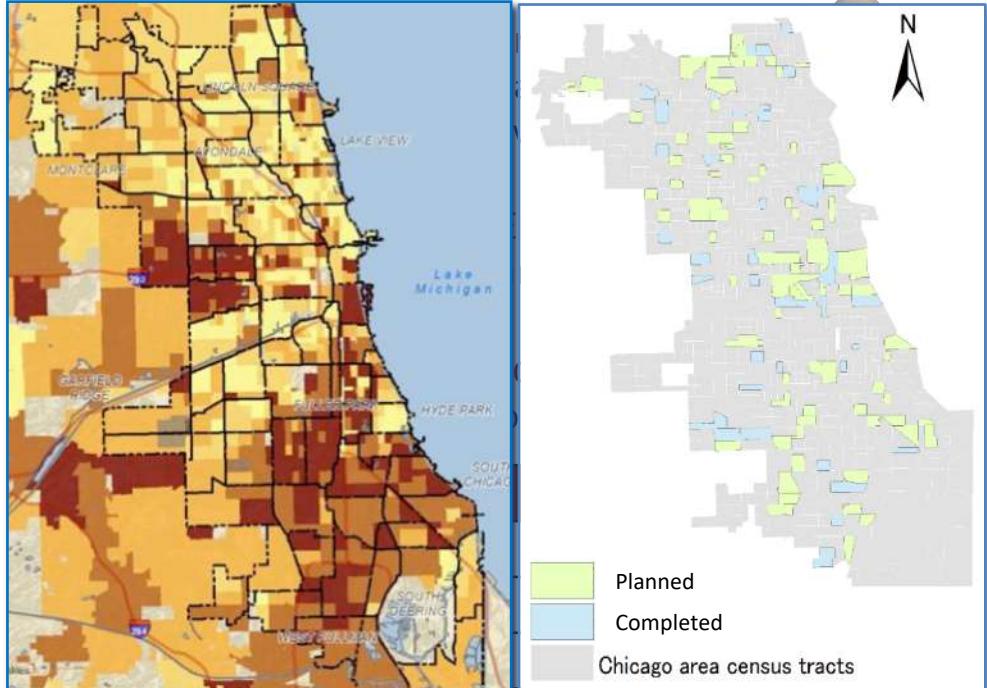
# Environmental Health (Risk Prediction)

## COMPASS Project

- What factors drive significant differences in life expectancy between neighborhoods?
- Cohort of >5,000 individuals from across Chicago

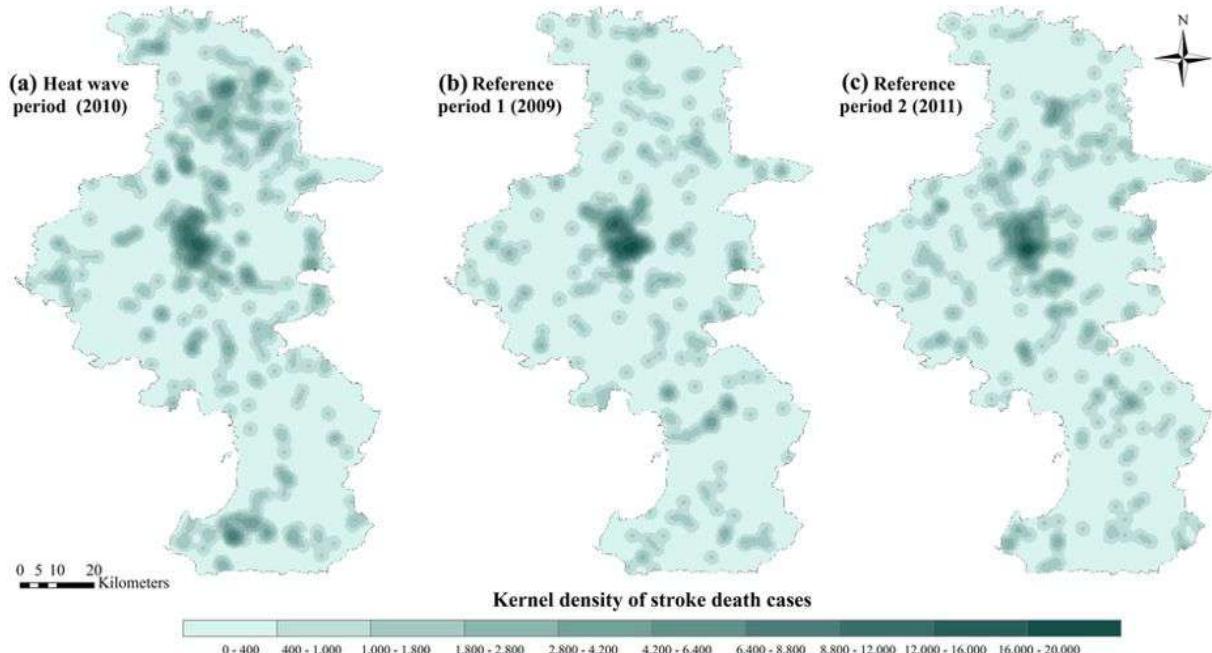
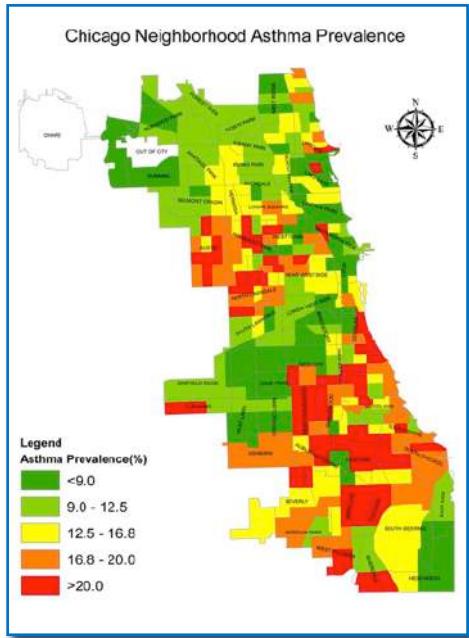
## COMPASS Data

- Standard medical (BP, BMI, immune status, medication, individual and family history)
- Lifestyle (smoking, alcohol, physical activity)
- Memory (recall tests)
- Psychological (depression, hostility, stress)
- Socioeconomic (education, income, occupation)
- Social Environment (sleep, stress, work hours)



# Health

## Spatial analysis of the effect of the 2010 heat wave on stroke mortality in Nanjing, China

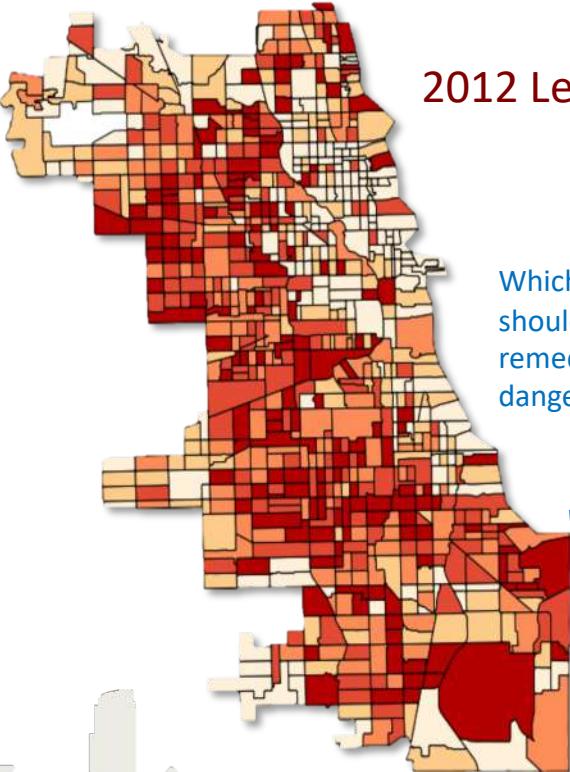


<http://www.mentalmunition.com/2011/09/south-side-children-have-greatest.html>

significantly increased or decreased adjusted odds ratio ( $p$ -value < 0.05).

Spatial analysis of the effect of the 2010 heat wave on stroke mortality in Nanjing, China  
Kai Chen, Lei Huang, Lian Zhou, Zongwei Ma, Jun Bi & Tiantian Li, **Nature**,  
*Scientific Reports* volume 5, Article number: 10816 (2015)

# Healthcare



API to integrate with electronic medical record clinical decision tools

**Lead Screening Risk Assessment**      DOB: 08/24/1964      Patient Age: 52 Years Old

**Chicago Assessment**  
The Chicago Assessment for lead risk is primarily geared towards patient in at-risk areas under the age of 6 years old. The tool can freely assess the risk at anytime. The tool will return ##### (number or text)

Previous Lead Risk Score: <5      >5  
**Run Risk Analysis**

Lead Risk 1      >5  
High Risk

Comments: Refer for Visual Home Inspection by CDPH

**Education**

**Lead Risk Education**      **Lead Reduction Strategies**      **Public Health Resources**

**Additional Questions:**

- Do you live in or regularly visit (once a week or more) any house or building built before 1978?  Yes  No  Don't Know
- Do you live in or regularly visit any house or building that has recently undergone renovation?  Yes  No  Don't Know
- Do you frequently come into contact with an adult whose job or hobby involves exposure to lead?  Yes  No  Don't Know
- Do you have contact with cosmetics, kohl, candies, spices, jewelry, ceramic dishware and/or home (or folk) remedies not made in the United States, and/or leaded crystal, imported ceramic, or pewter dishes?  Yes  No  Don't Know
- Do you play in loose soil, near a busy road or near any industrial sites such as a battery recycling plant, junk yard or lead smelter?  Yes  No  Don't Know
- Have you ever eaten dirt or put your mouth on painted surfaces, paint chips, toys, jewelry or vinyl mini blinds?  Yes  No  Don't Know
- Have you recently visited or lived in another country for an extended period of time?  Yes  No  Don't Know

R. Ghani (UChicago Center for Data Science and Public Policy); R. Mansour, Chicago Department of Public Health)

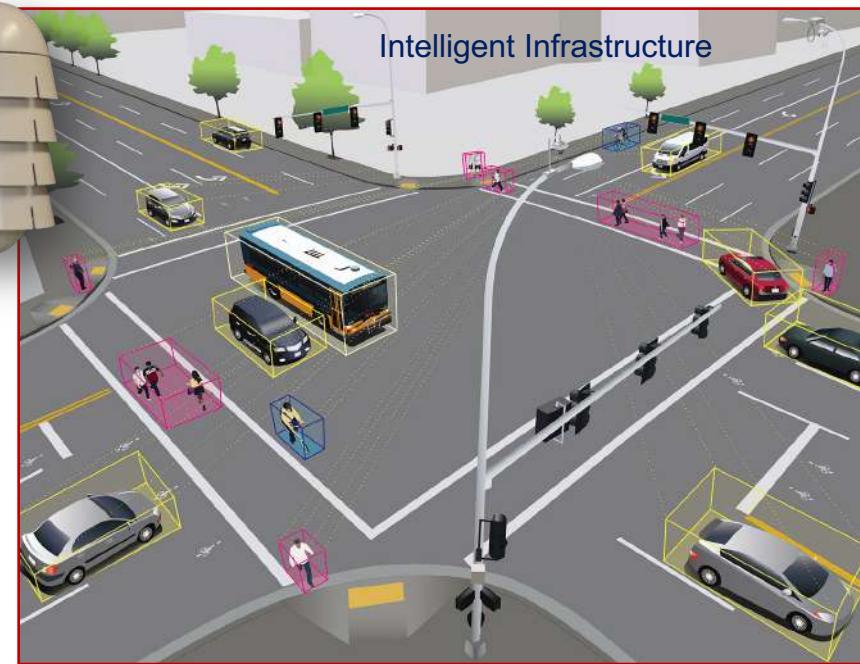
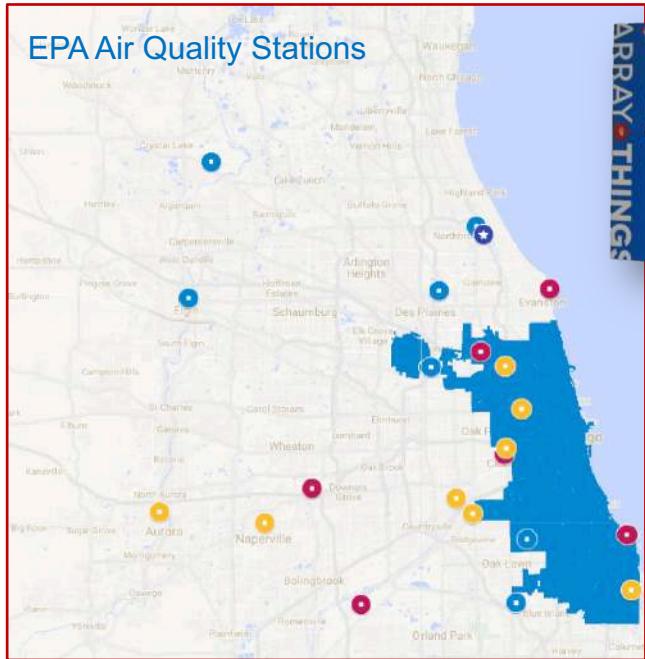
E. Potash, R. Ghani, R. Mansour, et. al., "Predictive Modeling for Public Health: Preventing Childhood Lead Poisoning," ACM KDD'15, August 2015, Sydney, Australia.

# The Array of Things Project

*For many questions, much greater  
resolution is necessary.*



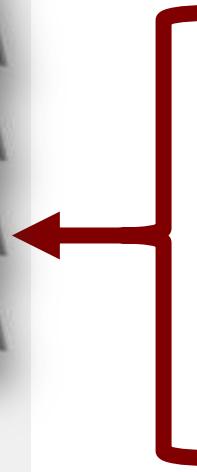
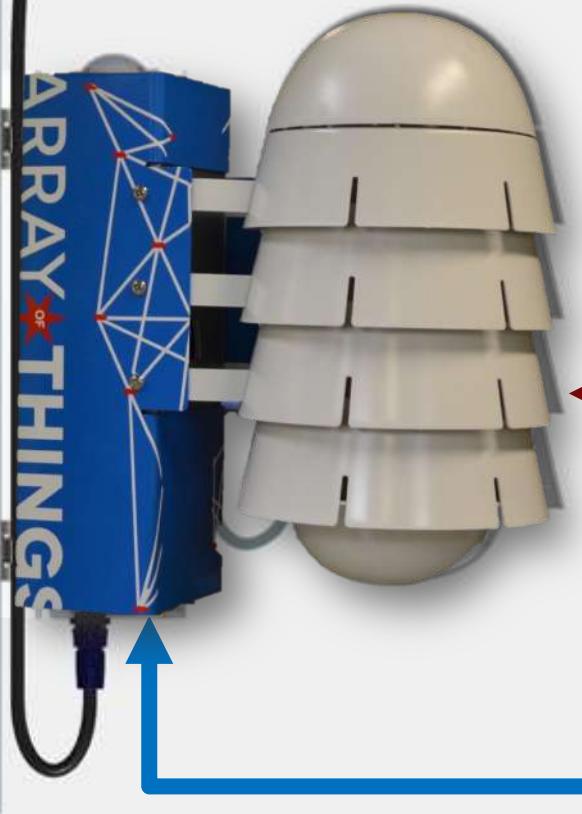
# Array of Things (AoT): Beyond Passive Measurement



AoT is an NSF-funded Major Research Instrumentation project to create an urban cyberinfrastructure "instrument" comprising hundreds of devices in partnership with the City of Chicago.



# AoT Current Configuration



## Environment

Ambient, UV, IR light  
Visibility  
Magnetic Field  
Vibration  
Sound pressure  
Temperature  
Relative humidity  
Barometric pressure

## Air Quality

PM 1, 2.5, 10, 40  
Carbon monoxide  
Ozone  
Sulfur dioxide  
Nitrogen dioxide  
Hydrogen sulfide  
Total reducing gases  
Total oxidizing gases

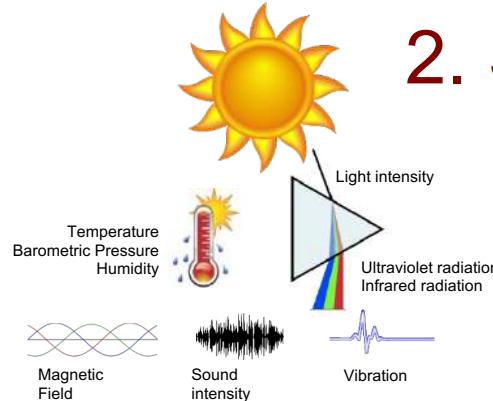


## Edge Computing

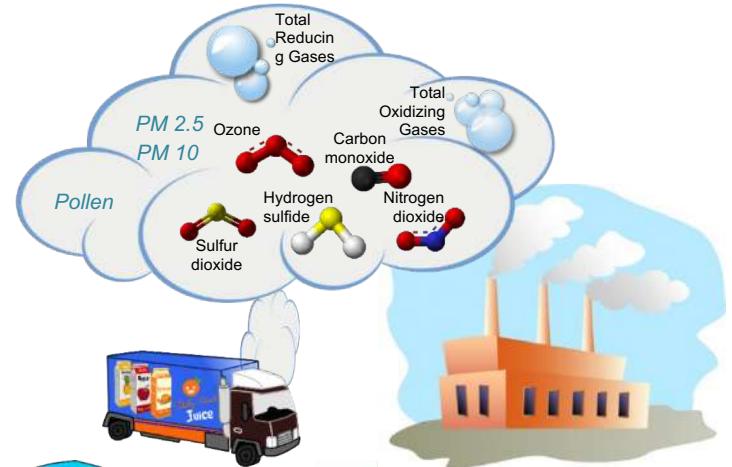
Computer Vision: Flooding, traffic flow, safety (bike helmet use, pedestrian patterns...), use patterns of public spaces, cloud cover  
Computer Audio: Noise components, sound events

# AoT: Cyberinfrastructure with Three Functions

## 1. Sensors



## 2. Sensor Testbed

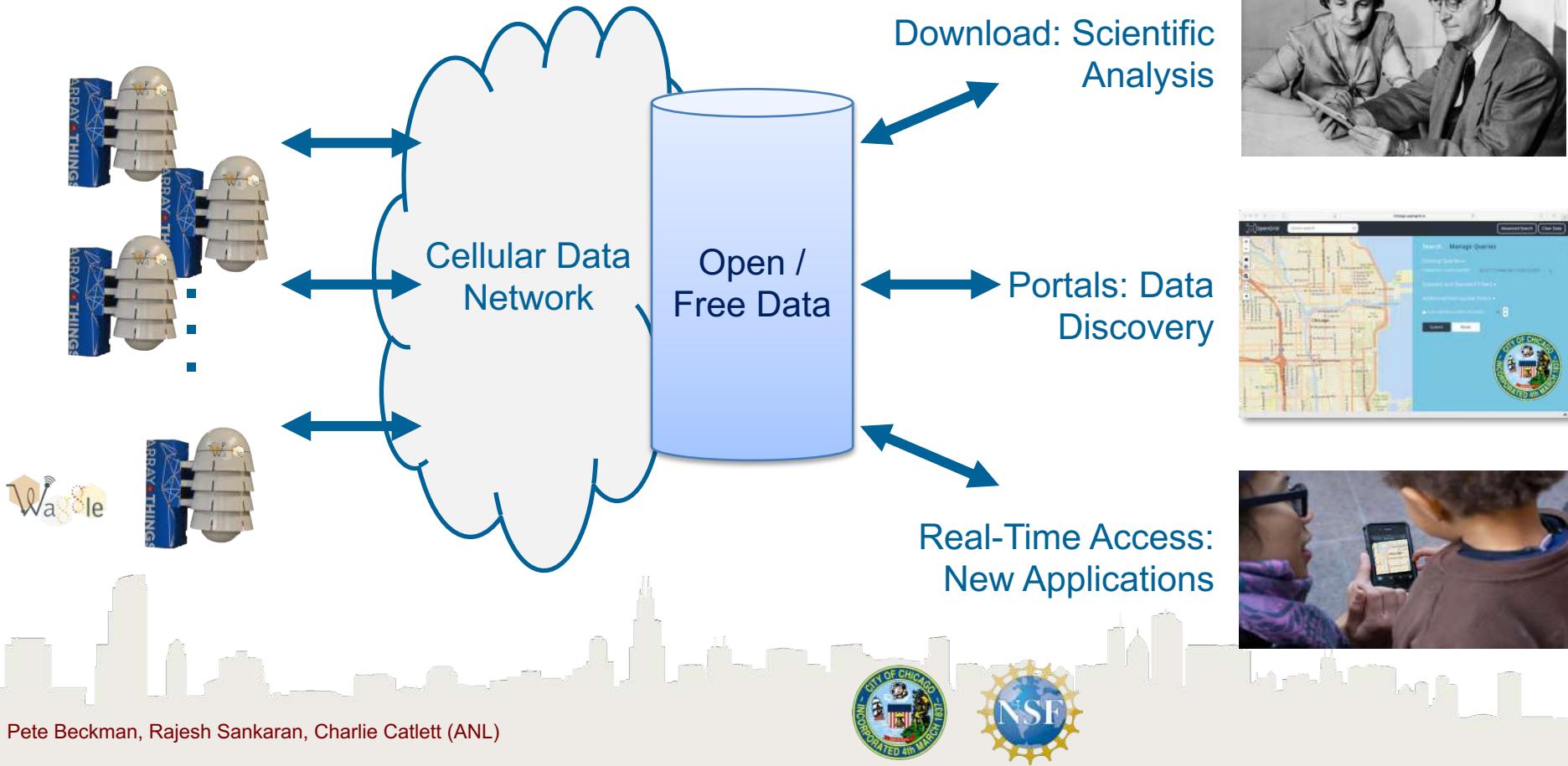


## 3. Artificial Intelligence Testbed



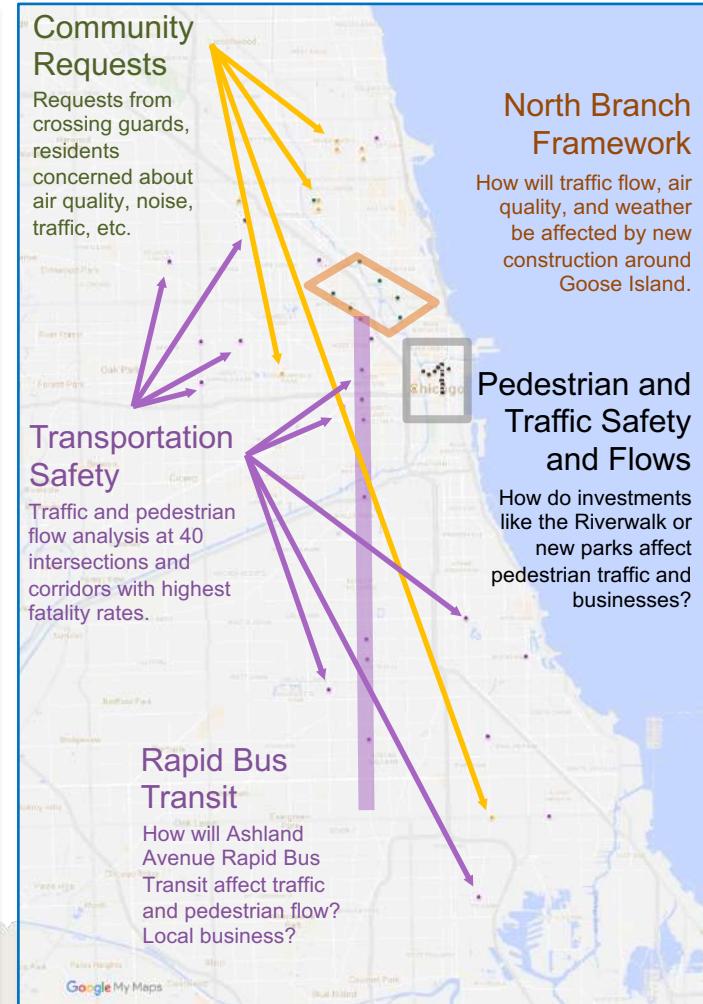
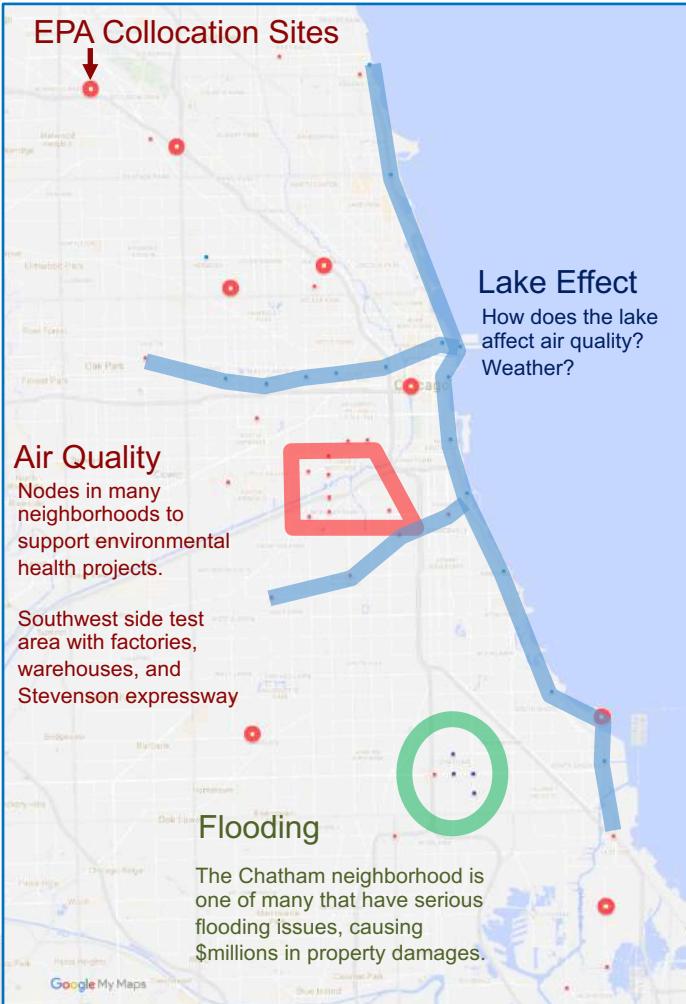
<https://aot-file-browser.plenar.io/>

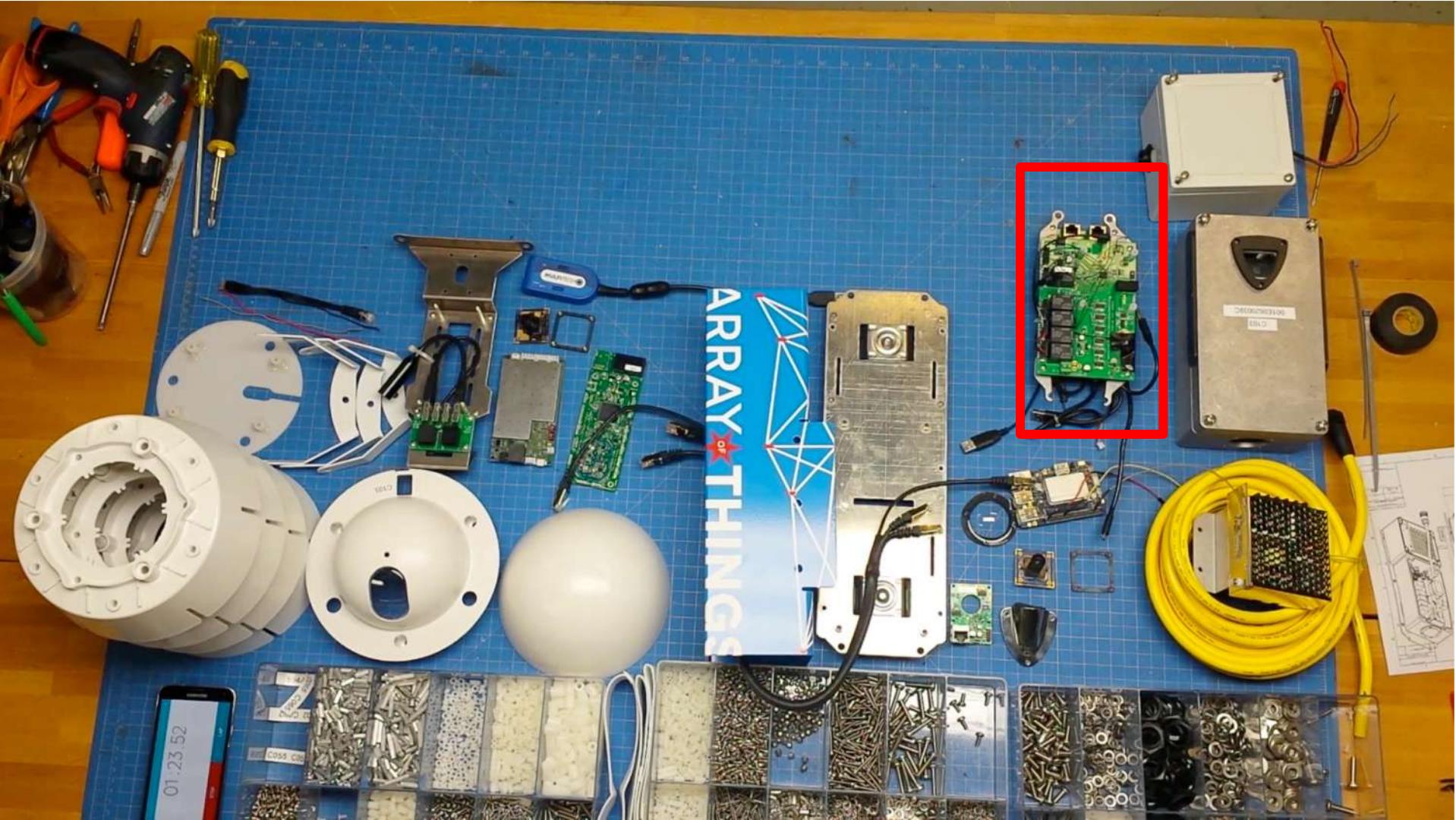
# All data is *open* and *free*.





All nodes are identical, but placement is driven by specific science or policy questions.





# The Waggle Platform

*Machine learning "at the edge" of a continuum.*



Northwestern  
University



THE UNIVERSITY OF  
**CHICAGO**

Argonne  
NATIONAL LABORATORY



# Edge+Machine Learning: A Revolution

## Sensors



## Powerful Parallel Edge Computing



Wagle

Artificial Intelligence  
Deep Learning Inference

## Actuators



**Edge computing and deep learning with feedback for continuous improvement**

## HPC/Cloud



Deep Learning Training

Reduced, Compressed data

New inference (program code)

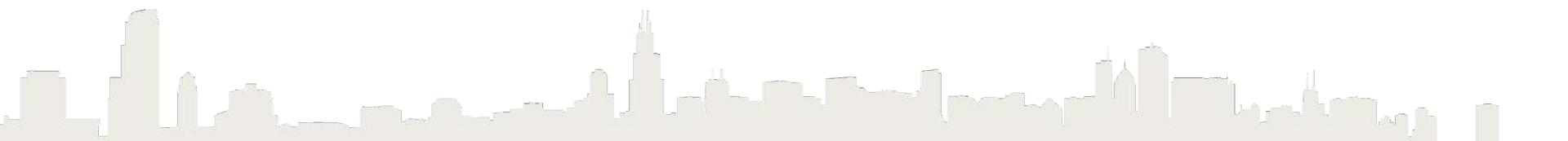
# The Computing Continuum

Fog							
	IoT/Edge				Fog		HPC/Cloud
Size	Nano	Micro	Milli	Server	Fog	Campus	Facility
Example	Adafruit Trinket	Particle.io Boron	Array of Things	Linux Box	Co-located Blades	1000-node cluster	Datacenter
Memory	0.5K	256K	8GB	32GB	256G	32TB	16PB
Network	BLE	WiFi/LTE	WiFi/LTE	1 GigE	10GigE	40GigE	N*100GigE
Cost	\$5	\$30	\$600	\$3K	\$50K	\$2M	\$1000M

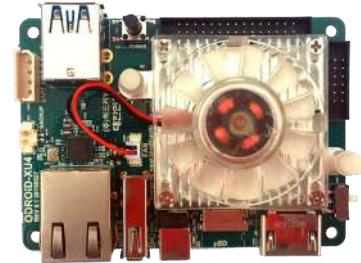
Count =  $10^9$   
Size =  $10^1$



Count =  $10^1$   
Size =  $10^9$

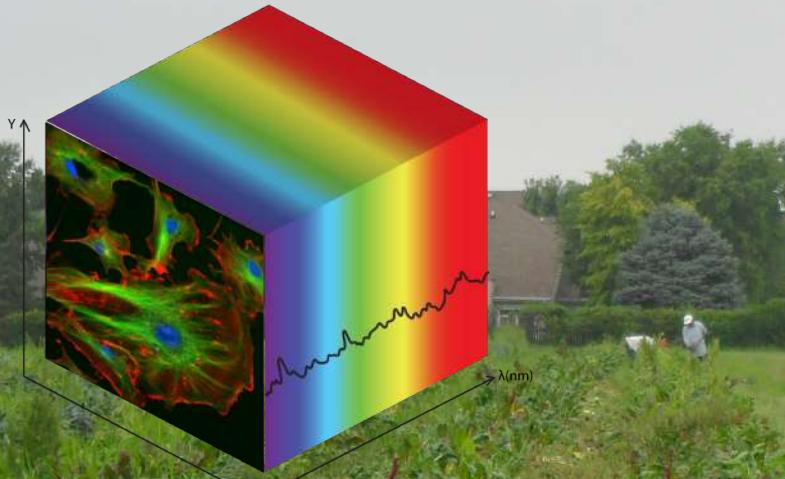


# Why Edge Computing?



- **More data than bandwidth**
  - Spallation neutron source, light source, HD Cameras, LIDAR, radar, hyperspectral imaging, grid micro-synchrophasors, etc.
- **Latency is important**
  - Quick local decision & actuation; adaptive sensing & control systems
- **Privacy/Security requires short-lived data: process and discard**
  - Compromised devices have no sensitive data to be revealed
- **Resilience requires distributed processing, analysis, and control**
  - Predictable service degradation, autonomy requires local (resilient) decision
- **Quiet observation and energy efficiency**
  - Vigilant sensors, transmit only essential observations, not big data streams





Example: SPECIM Camera:  
PFD VNIR with 768 bands  
 $(2734 \times 1312) \times 768 \times 2\text{bytes} = \mathbf{5.1\text{GB image}}$

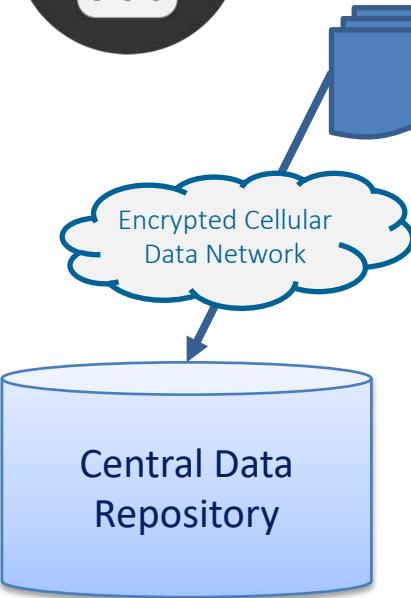
1 sample every 5 min  
Twilight to twilight on June 21 = **1TB**

**We need a parallel computer with each sensor!**

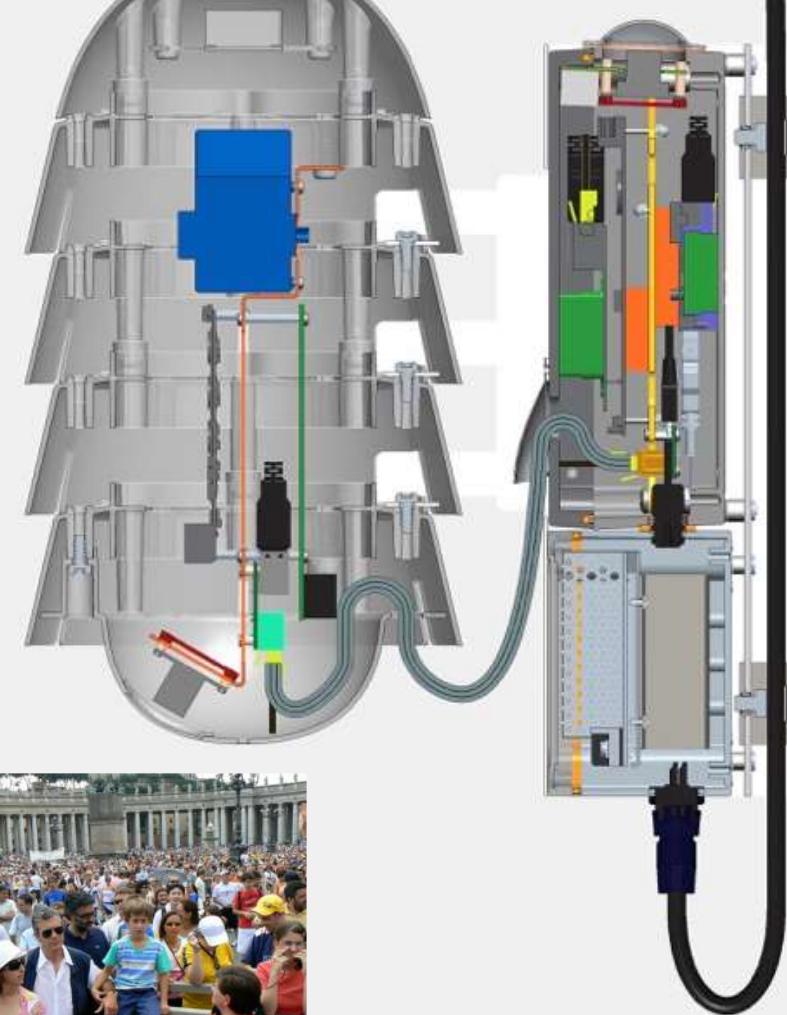


## Approved Algorithms:

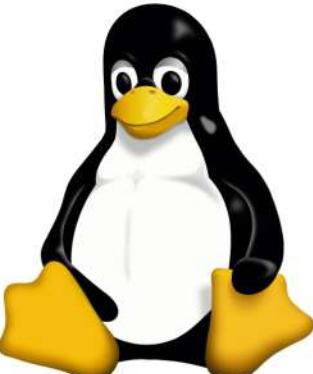
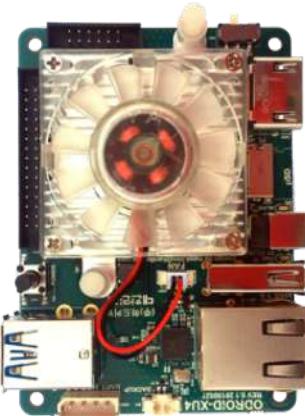
- 1. Surface water      ●
- 2. Pedestrian count    ■
- 3. Vehicle mix        ▲



After images are processed they are deleted. They are not transmitted or stored.

A photograph of a large crowd of people walking in a public square, likely St. Peter's Square in Rome, with a building featuring columns in the background.

# When a Computer + Linux is Not Enough...

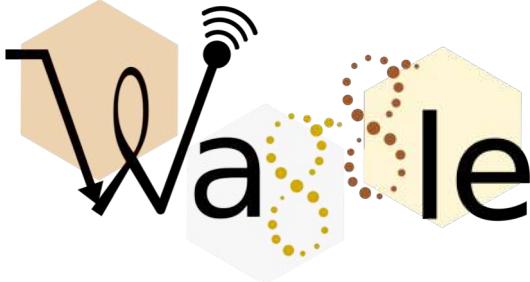


Challenging Design Contradiction

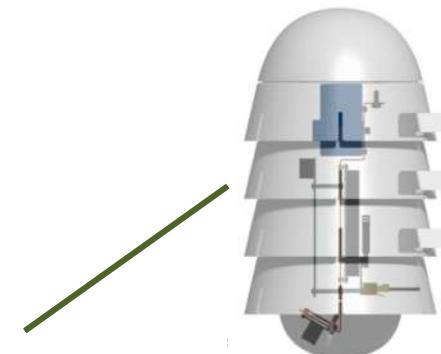
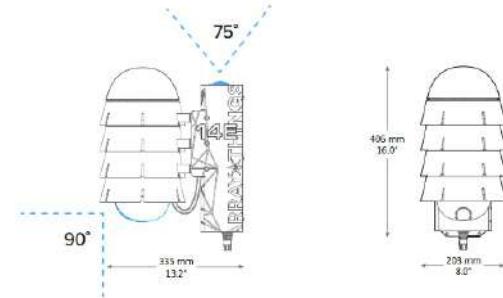


- Experimental ML/GPU software fails often
- Edge Devices are remotely deployed

```
[20728464.998512] No filesystem could mount root, tried.  
[20728464.998518] Kernel panic - not syncing: VFS: Unable to mount root fs  
[20728464.998526] CPU: 0 PID: 1 Comm: swapper/0 Not tainted 3.10.0-229.4.2  
[20728464.998532] ffffff81814288 0000000084f1a4a1 ffff880066eb7d60 ffff  
[20728464.998540] ffff880066eb7de0 ffffffff815fe71e ffffffff00000010 ffff  
[20728464.998547] ffff880066eb7d90 0000000084f1a4a1 0000000084f1a4a1 ffff  
[20728464.998554] Call Trace:  
[20728464.998565] [ffffffffff81604eaa] dump_stack+0x19/0x1b  
[20728464.998570] [ffffffffff815fe71e] panic+0xd8/0x1e7  
[20728464.998579] [ffffffffff81a4560d] mount_block_root+0x2a1/0x2b0  
[20728464.998585] [ffffffffff81a4566f] mount_root+0x53/0x56  
[20728464.998590] [ffffffffff81a457ae] prepare_namespace+0x13c/0x174  
[20728464.998596] [ffffffffff81a4527b] kernel_init_freeable+0x203/0x22a  
[20728464.998602] [ffffffffff81a449db] ? initcall_blacklist+0xb0/0xb0  
[20728464.998609] [ffffffffff815f33f0] ? rest_init+0x80/0x80  
[20728464.998614] [ffffffffff815f33fe] kernel_init+0xe/0xf0  
[20728464.998620] [ffffffffff81614d3c] ret_from_fork+0x7c/0xb0  
[20728464.998625] [ffffffffff815f33f0] ? rest_init+0x80/0x80
```



- Support “edge” computing—in situ data analytics for anomaly detection in high-bandwidth sensor streams, machine learning for computer vision, etc.
- Provide modular, secure data pipeline.
- Remotely program edge computing devices (and recover from “bricking”).
- Add sensors easily with minimal re-design (solder ports; shield expansion); rapidly swap devices for upgrades.



# Argonne's Open Waggle Platform

## A Pocket-Sized Controller for Edge Computing



*The Waggle Manager (WagMan)*

- Borrowed BG/Q control system ideas
- Designed mini “rack controller”
  - Devices can be disconnected
  - Devices can be power cycled
- “Deep Space Probe” design
  - Heart beat signals to each device
  - Alternative boot image / safe mode
  - Current and voltage monitoring
  - Environmental monitoring
- Strict cybersecurity design

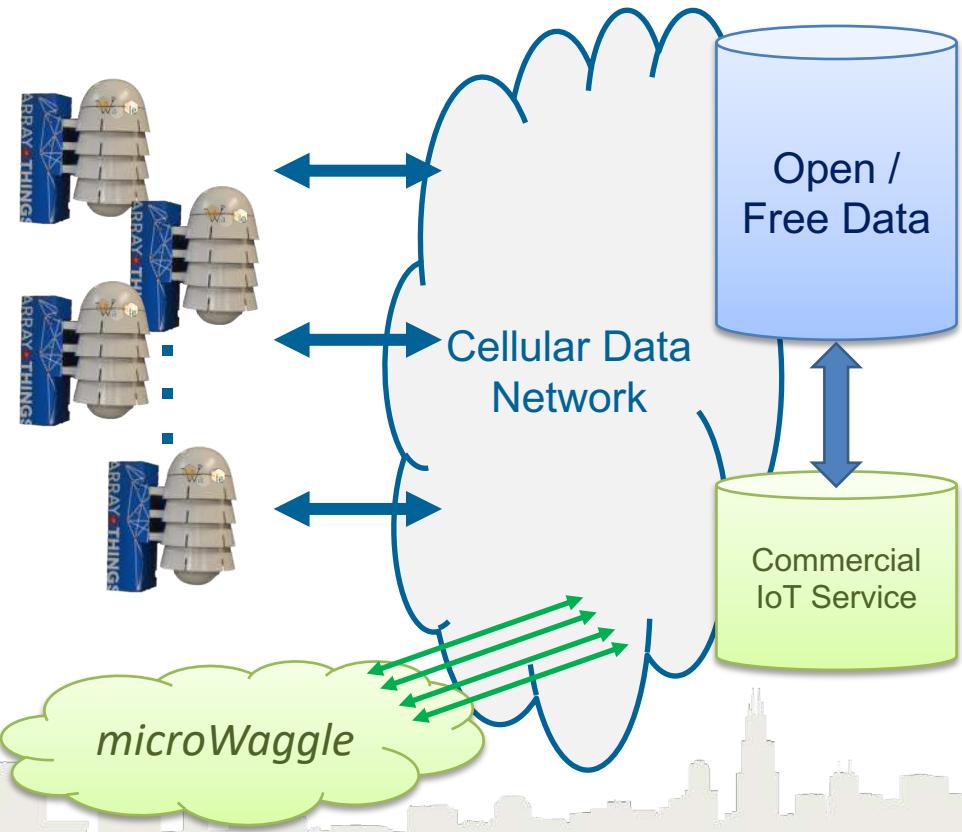
# Waggle: Argonne's *Edge Computing* Platform

*Bring Parallel Computing to the Edge*

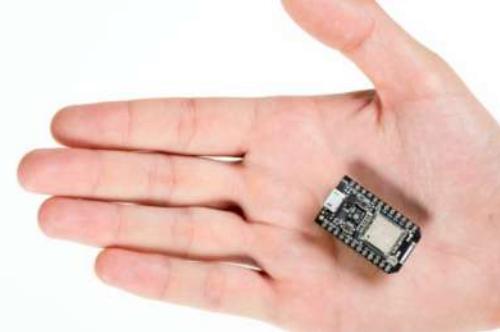
- Supports powerful, parallel computation at the edge
  - Computer vision and deep learning frameworks (Caffé, TensorFlow, OpenCV)
  - Supports edge-optimized & experimental computing
    - ML hardware, GPUs, neuromorphic, FPGAs, etc.
- Open Source, open interfaces
- Integrating advanced sensors: simple plug-in architecture
- Robust remote system management subsystem
- Manufactured at local electronics company



# Extensible Open Architecture



*microWaggle uses cellular, mesh, or WiFi to connect lower-cost sensors, reporting data in Waggle format to mix readily with AoT data.*



# A Survey of Machine Learning Opportunities

*Array of Things and Waggle present many interesting opportunities to apply machine learning to diverse science and policy questions.*



# Array of Things Coverage

July 2018

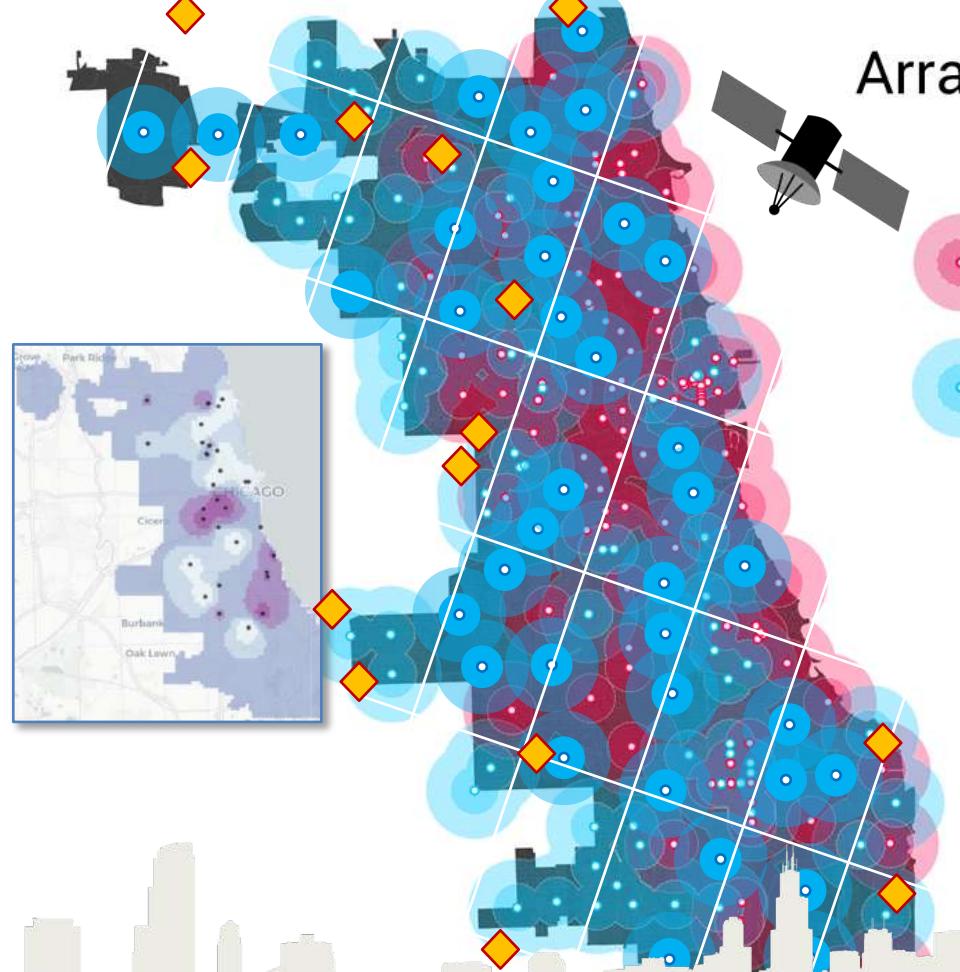
May 2018 Nodes

<https://aot-file-browser.plenar.io/>

Dec 2018 Nodes

Today there is an AoT node (red) within 2km of 80% of Chicago's population, and within 1km of 42% of the population

By early 2019 every Chicago resident will have an AoT node within 2km and at least 70% will live within 1km of a node.



Map: A. Laha (UChicago Spatial Data Science Center)





# Transportation



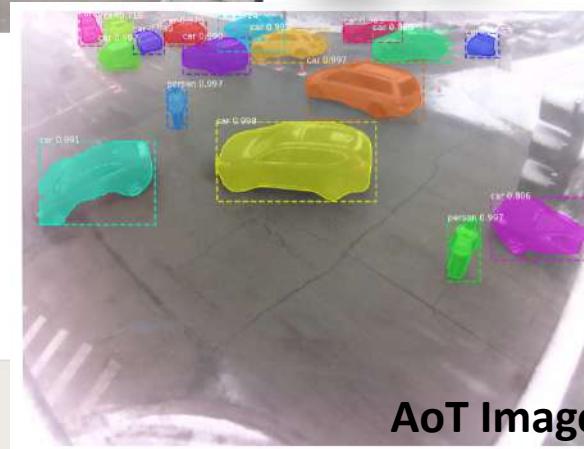
Mask r-CNN [He, 2017]  
trained on COCO dataset

Research Credits:  
Yongho Kim, Seongha Park  
(Purdue PhD Students @ ANL, 2018)  
Pete Beckman, Nicola Ferrier (ANL Scientists)

Advanced computer vision to understand pedestrian movement, eventually to predict dangerous interactions with vehicles

Research Credits:

Zeeshan Nadir (Purdue PhD Student @ ANL, 2017)  
Nicola Ferrier (ANL Scientist)



AoT Image from Lake Shore Drive

Pete Beckman beckman@anl.gov

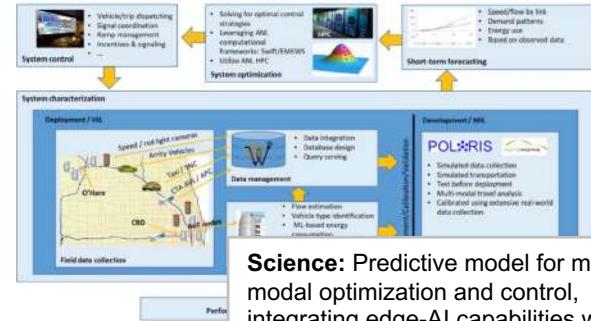


# Transportation



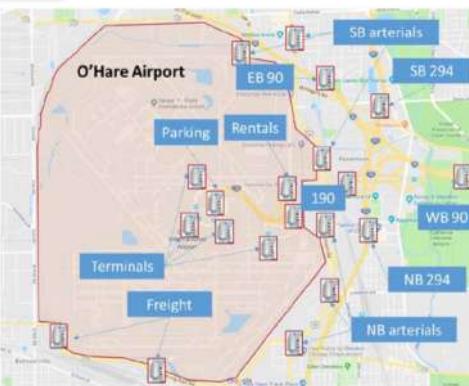
**Science:** Prototype model of at-grade crossing with impact analysis (interrupt duration and impact; emergency vehicles delayed)

**Objective:** Prioritize among hundreds of at-grade crossings in context of \$1B planned investments to improve rail throughput by eliminating key at-grade crossings.



**Science:** Predictive model for multi-modal optimization and control, integrating edge-AI capabilities with traditional transportation data, coupled with HPC models and control systems.

**Deployment:** Integrate transportation measurements from AoT/Waggle (density, flow, vehicle mix, parking) with live traffic data and traffic model around O'Hare International Airport.



**Funding:** Illinois DOT

Partners: Argonne, UChicago, Chicago Metropolitan Agency for Planning, Chicago DOT



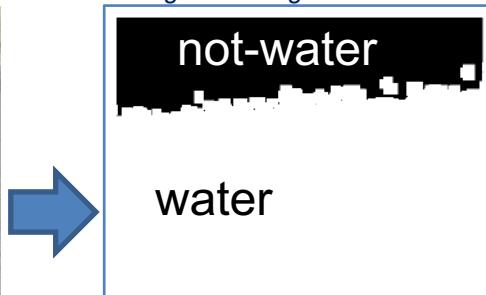
**Funding:** U.S. Department of Energy

Partners: Argonne, Chicago DOT, Chicago Dept. of Aviation, Chicago Dept. of Innovation and Technology, Arity



# Flooding

50 consecutive frames to flood water and segment image



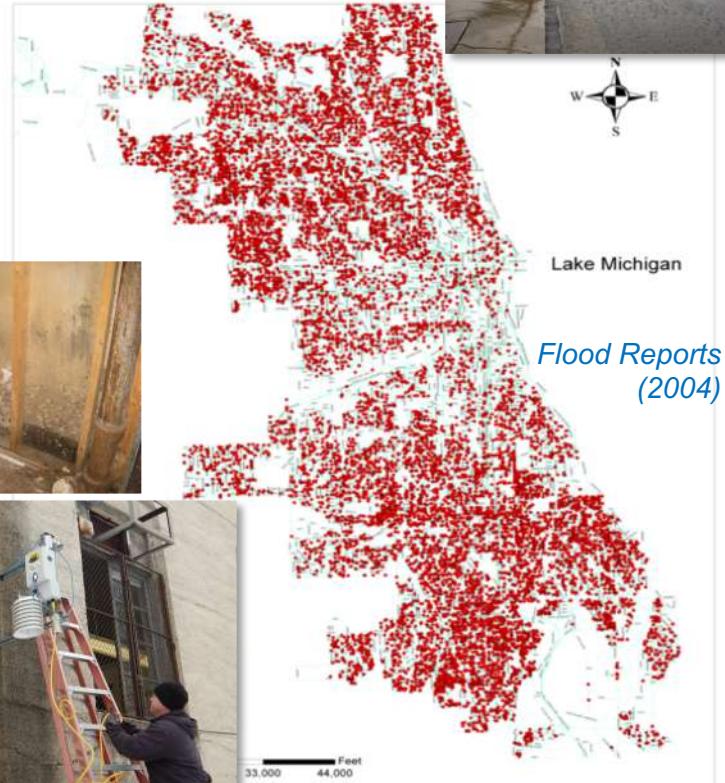
Using advanced computer vision to detect surface flooding

Ethan Trokie (Northwestern Undergrad Student @ ANL, 2017)

Vivien Rivera (Northwestern PhD Student, SCGSR 2018)

Nicola Ferrier (ANL Scientist)

Rajesh Sankaran (ANL Scientist)



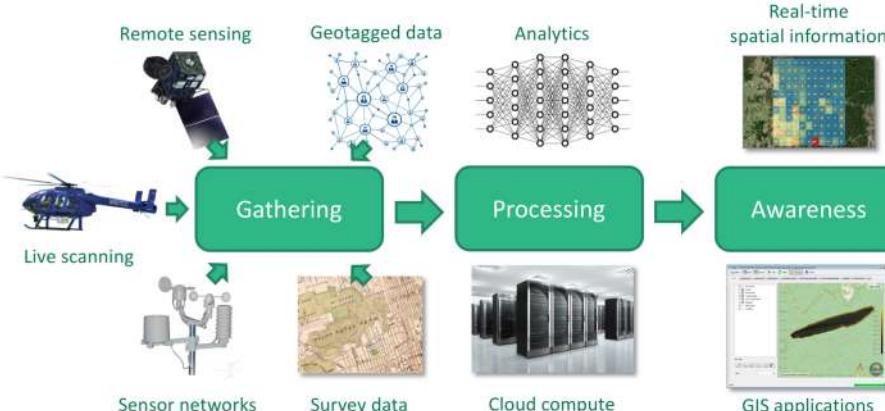
Waggle Nodes in  
Tuley Park,  
Chicago





# Disaster: Flood, Fire...

Partnership with CSIRO Australia (MOU, visiting postdoc)

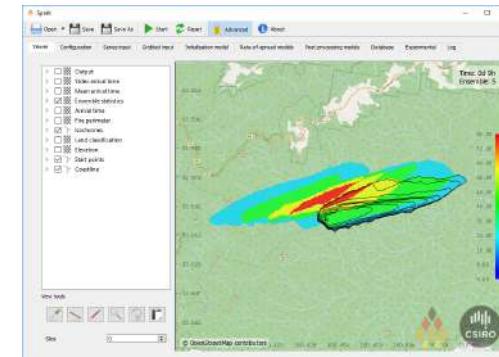


Nikhil Garg

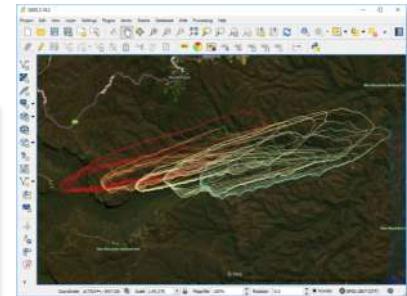


Live HPC  
Modeling  
and Prediction...

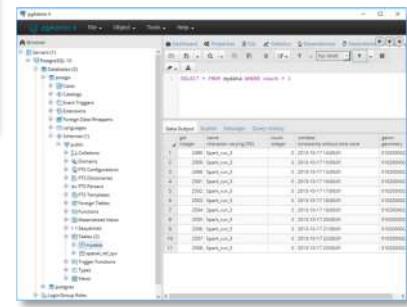
- Basis of D61 natural hazard applications:
  - Wildfire and wildfire impact (Spark)
  - Flood and coastal inundation (Swift)



Impact analysis from ensemble  
wildfire simulations



GIS visualisation



Integration with PostGIS  
database



# Environmental Science

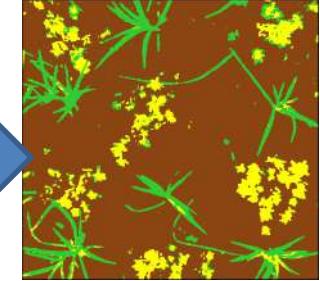


Advanced sensors and computer vision to monitor pristine prairie

Vivien Rivera (Northwestern Univ. PhD Student, 2018)  
Aaron Packman, Bill Miller (Northwestern Univ. Professors)  
Pete Beckman (ANL Scientist)



Original



Auto Segmentation

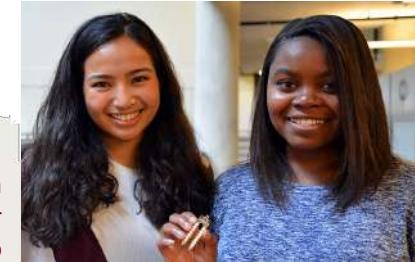
Using advanced computer vision to monitor plants

Renee Zha (Northwestern Undergrad Student @ ANL, 2017)  
Zeeshan Nadir (Purdue PhD Student @ ANL, 2017)  
Nicola Ferrier (ANL Scientist)



Chicago Botanic Garden  
Conservation Science Center

Undergrads Caeley and Jordan developed soil moisture sensor now deployed in Chicago

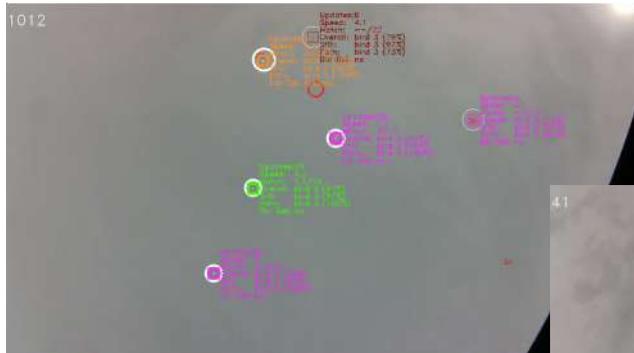




# Public Safety

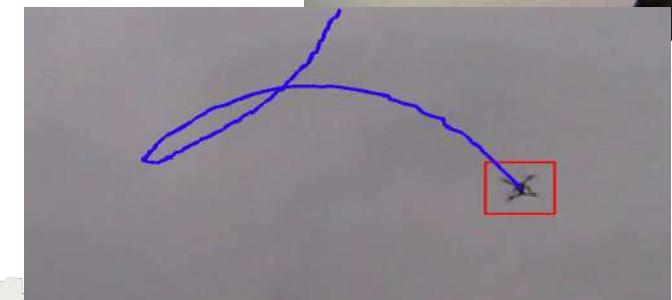
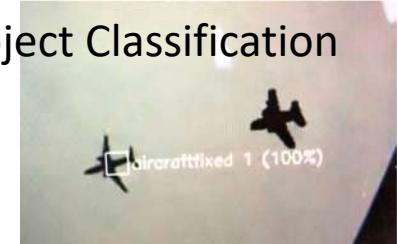
*Bird? Plane? Drone?*

Advanced computer vision and machine learning to identify drones, birds, or fixed-wing aircraft.



Sean Richardson (USAF visiting ANL)  
Adam Szymanski (ANL Scientist)

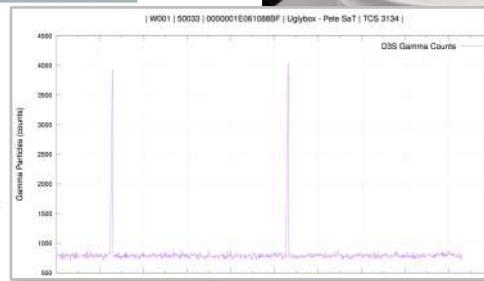
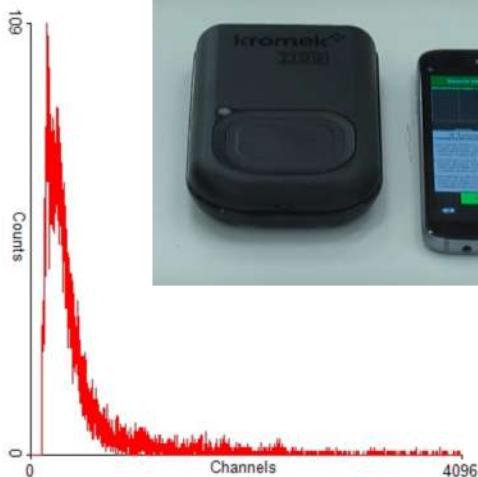
Object Classification





# Public Safety

Testing Edge / Waggle  
DARPA SIGMA+



*Benchtop integration of DARPA-funded  
radiation sensor (Kromek-D3S)*

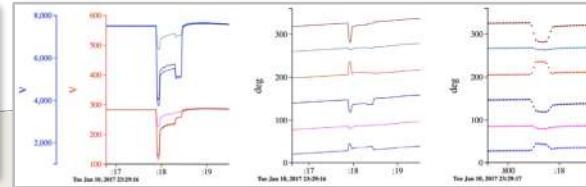
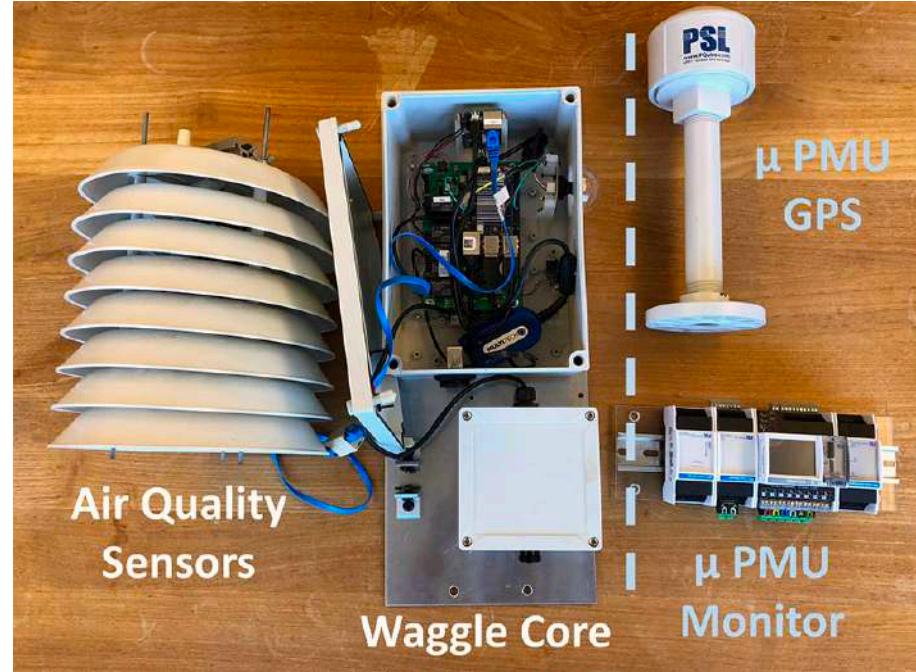




# Energy: Power Grid



- Load Forecasting
- Grid Stress
- Air Quality



DOE  
ARPA-E



# Harnessing The Computing Continuum

Science-driven  
Problems



e.g.: "Predict urban response to rainfall,  
trigger intelligent reaction..."

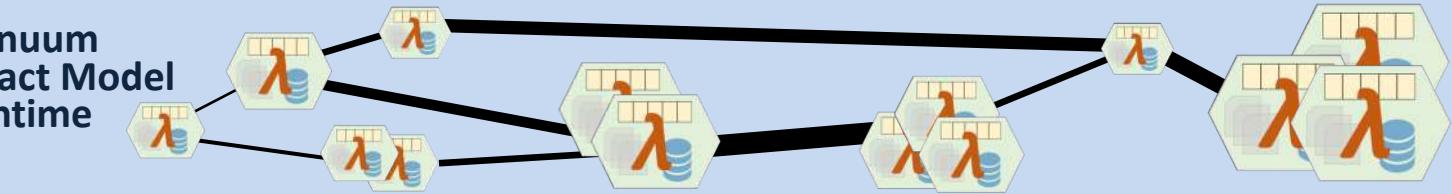


Goal-oriented  
Annotations

Notional  
Example:

trigger {flood\_actuation, resident\_warning}  
when {wx\_prediction, sewer\_model} implies  
(traffic\_capacity < 70%) or (home\_flooding > 5%)

Continuum  
Abstract Model  
& Runtime



Existing  
Resources  
& Services



# Argonne Edge Research: Edge→HPC

- Continually improving Edge-HPC Systems
  - Deep learning + lightweight training + continual improvement
  - Incremental model updates
  - Is Edge really a layer in the model?
- How will the OS/R and system software evolve for Edge-HPC?
  - Scheduling, security, resource management, streaming data
- Programming model & framework for Continuum Computing
- Optimized ML hardware for both Edge & HPC
- Theoretical foundations for failures and correctness of edge/training
- Dynamic resource management and adaptive inference priority
  - AI at the Edge is limited by power and computation – just like HPC
- **Fluid HPC** to support complex and on-demand workflows on future exascale

