

# From the Edge to the Supercomputer: Analysis along the way

Pete Beckman: Co-Director Northwestern University / Argonne Inst. for Science and Engineering

Collaborators: Ilkay Altintas, Charlie Catlett, Scott Collis, Nicola Ferrier, Eugene Kelly, Jim Olds, Mike Papka, Dan Reed, Raj Sankaran, Sean Shahkarami, Joe Swantek, Valerie Taylor, Doug Toomey, Frank Vernon, Rommel Zulueta, and many more....

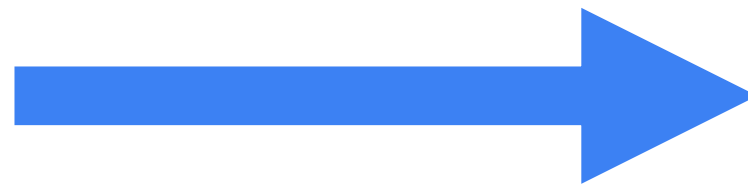




**Instrument**



**Data**



**Analysis**





# The Digital Continuum

Instrument

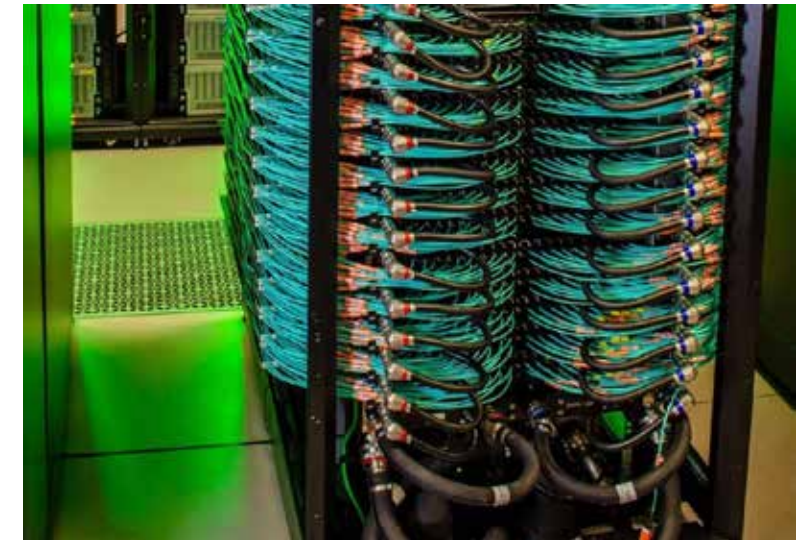
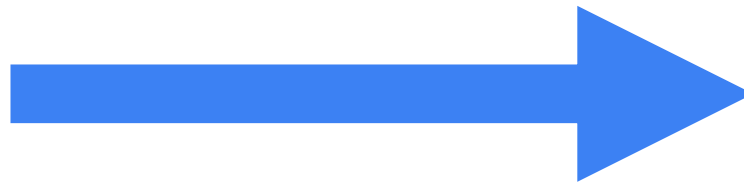
HPC/Cloud



IoT



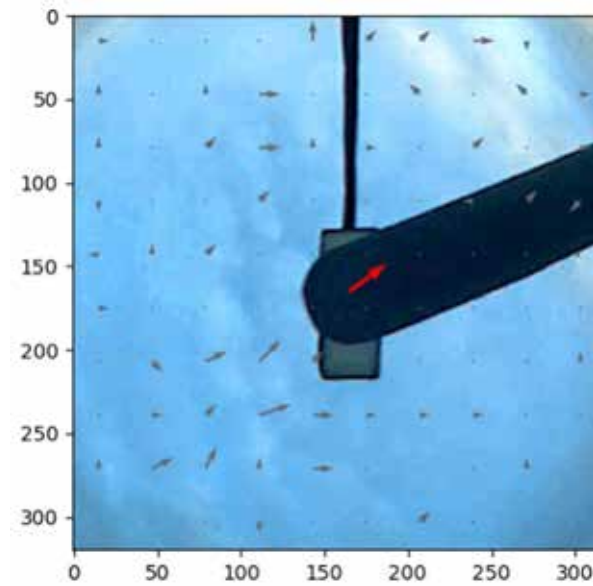
Facilities



Analysis

*Analyse full resolution data,  
find highest value data for  
the science*

# Why Live on the Edge?



*Analyse full resolution data,  
find highest value data for  
the science*

- More data than bandwidth
  - Imaging, LIDAR, SW defined radios, radar, audio, hyperspectral, large facilities, ...
- Latency is important
  - Quick local decision, experimental control & actuation: adaptive sensing
- 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-making
- Quiet observation and energy efficiency
  - Vigilant low-power sensors, transmit only essential observations

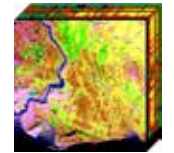
# AI@Edge: Digital Continuum

## Sensors



LIDAR

Software Defined Radios



Hyperspectral Imaging

## Facilities



## Actuators

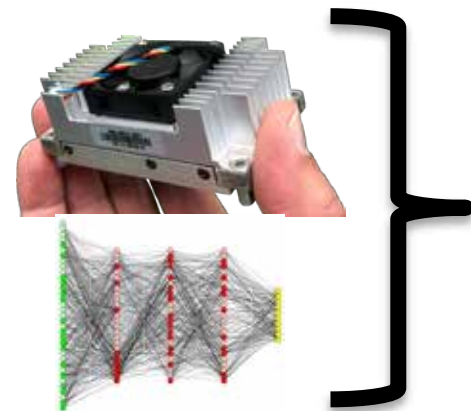


Servos

Dynamic adaptation



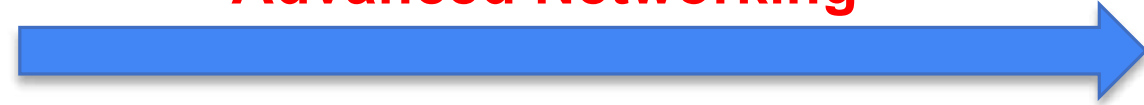
## Edge Computing



Scientific Data Analysis & Control

Artificial Intelligence  
Deep Learning Inference  
Lightweight Training  
Autonomous Action

## Advanced Networking



## New inference (model) Adaptive controls / steering



## Computation



Cloud



Data Center



HPC

Predictive Sim  
Digital Twins  
Data Analysis  
Machine Learning



# AI-Based Measurement & Anomaly Detection, & Control

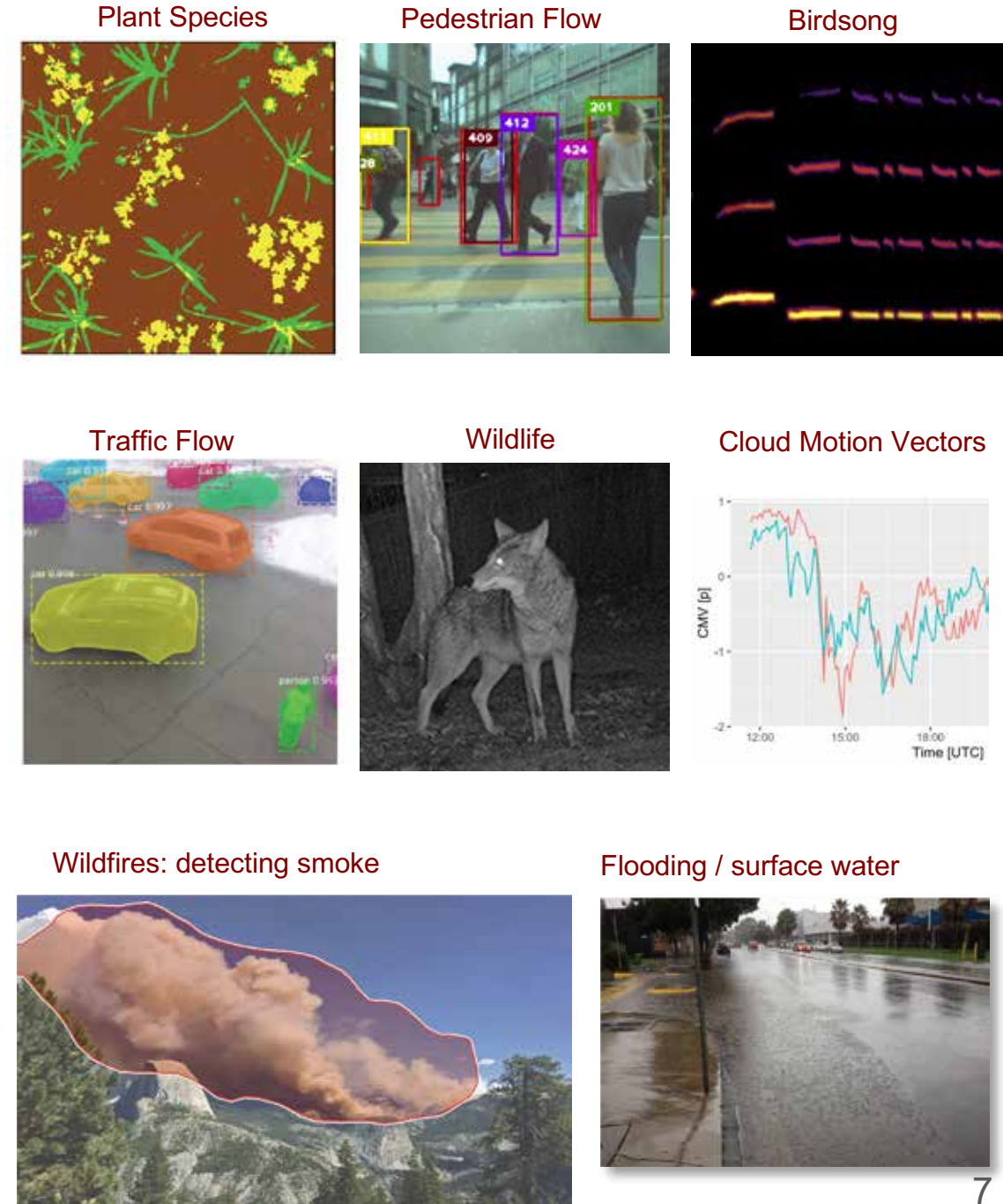
## What is a "Software Defined Sensor"?

<> "embedded computing"



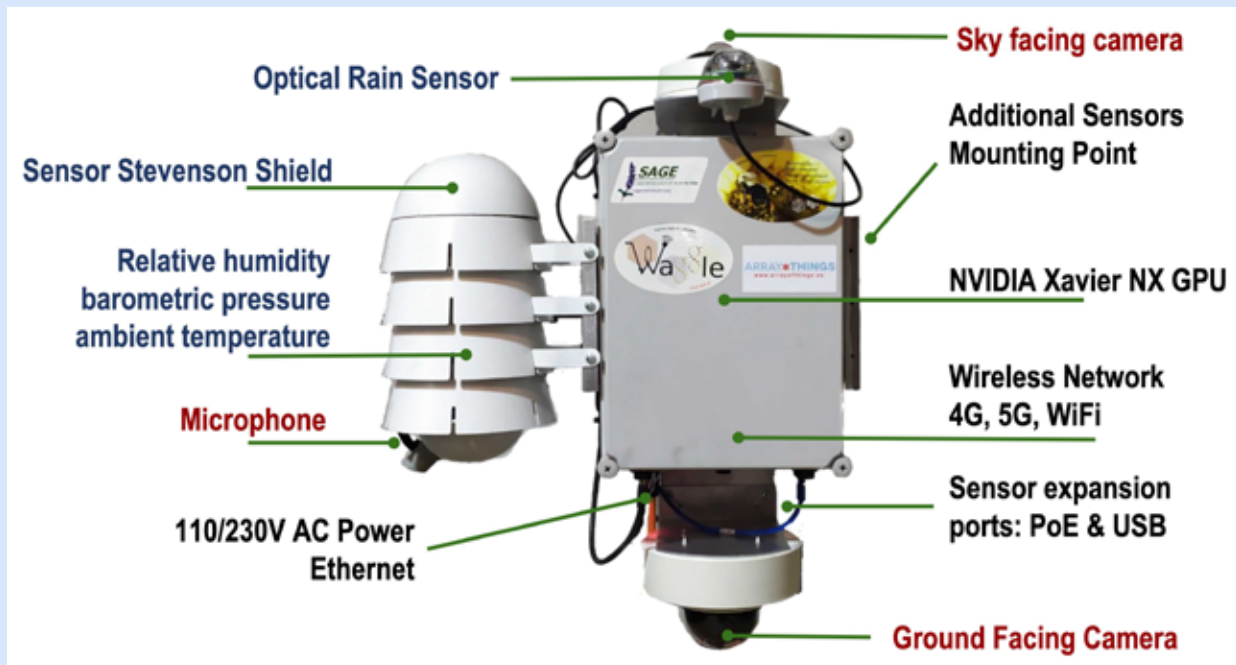
Your software container running here

Analysis produces live results



# Delivering AI@Edge: Two Forms

## Wild Sage Node



Ready for mounting **outside**, any PoE sensor can be easily added

## Sage Blade

(Sage software stack + **pure commodity server**)



Rugged server for instrument huts, new sensors easily added





# SAGE

Cyberinfrastructure for  
**AI at the Edge**  
[sagecontinuum.org](http://sagecontinuum.org)



## Leadership Team



Pete Beckman  
(NU: Director)



Nicola Ferrier  
(UC: Deputy Dir.)



Scott Collis  
(NU: Instruments,  
Atmos)



Valerie Taylor  
(UC: Edu, Broader  
Impacts)



Eugene Kelly  
(CSU; Ecosys,  
NEON)



Mike Papka  
(NIU: Edu, Broader  
Impacts)



Raj Sankaran  
(NU: Node Arch)



Ilkay Altintas  
(SDSC: Data)



Charlie Catlett  
(Uillinois: Urban)



Jim Olds  
(GMU; Life Sci,  
Risk)



Dan Reed  
(Utah;  
Architecture)



Kathy Bailey  
Proj Mgmt



Helen Taaffe  
Proj Mgmt



Joe Swantek  
NU: Software

Put AI@Edge



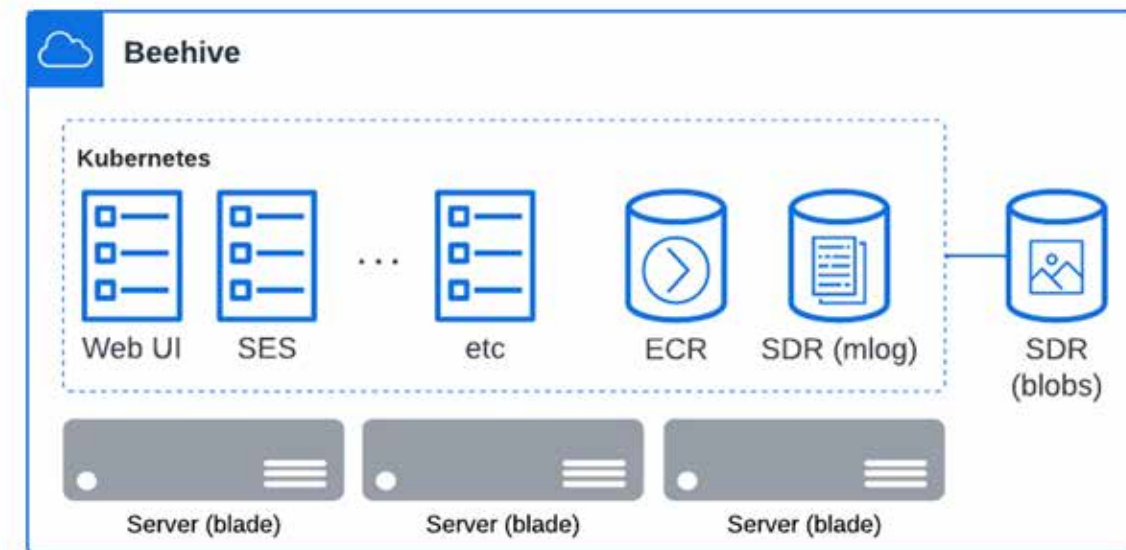
Analyse full resolution data,  
find highest value data for  
the science

# Sage Software Architecture

## Sage Nodes: "Cloud Native"



## Cloud Infrastructure



AI@Edge "Plugin" from Edge Code Repository (ECR) (the "App Store")

- Built upon standard AI Stack
- **Containers on Kubernetes**
- **Multi-tenancy**
- **"Goal-based" Scheduler**
- Local control for actuation
- Extreme cybersecurity
- Publish data to Beehive

## Beehive manages

- Sage Edge Scheduler (SES)
- Sage Data Repository (log entries)
- Sage Data Repository (binary files)
- User Interface components



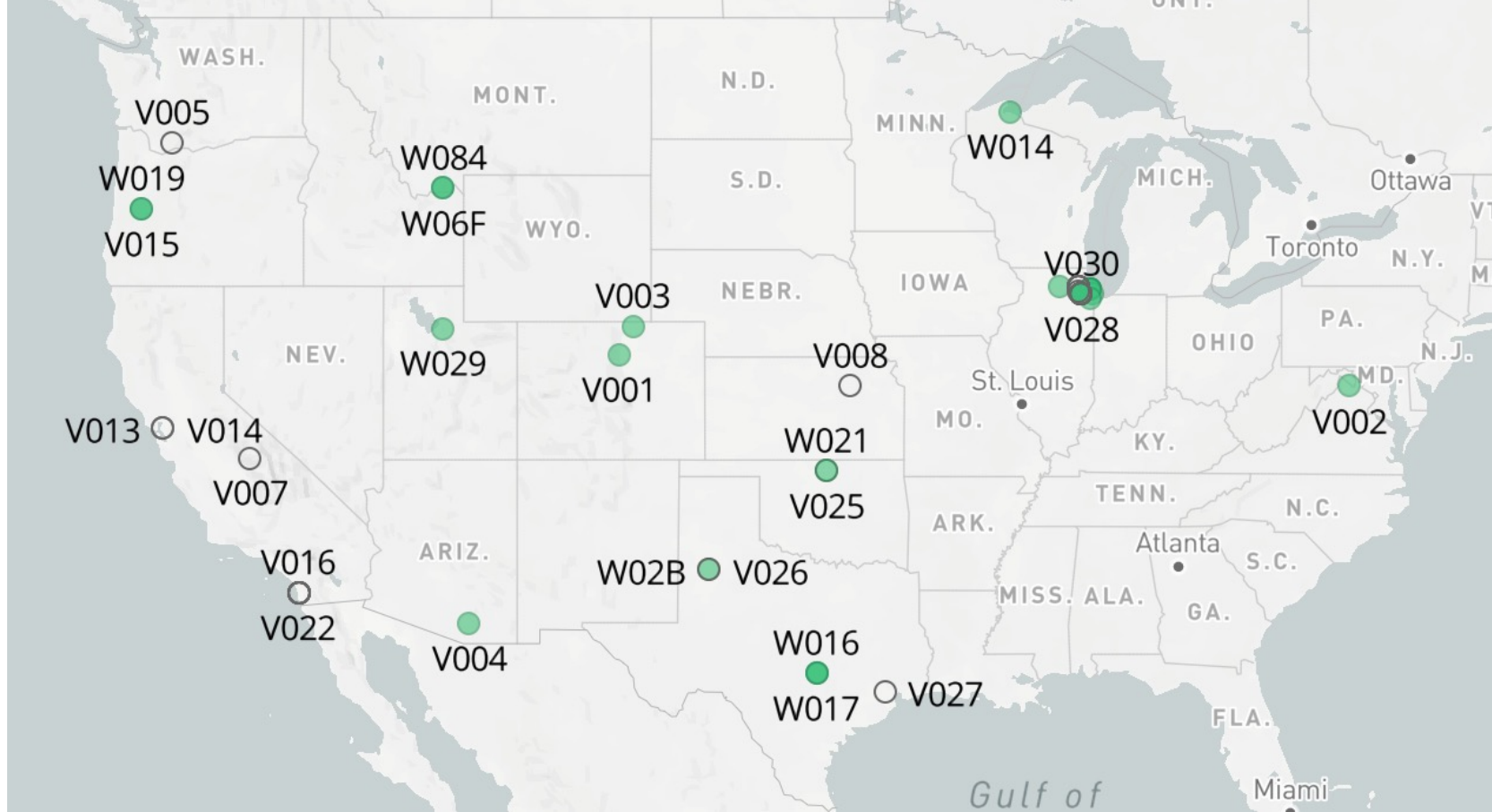
# Building Community for AI@Edge Applications

The screenshot displays the Sage App Catalog interface. At the top, the Sage logo (beta) is on the left, and navigation links for 'App Catalog', 'Job Status', and 'Data' are in the center. On the right, there are links for 'Docs' and 'Sign In'. Below the header, a search bar is present. The main content area is titled 'Featured Apps' and contains a grid of 12 application cards. Each card includes a thumbnail image, the app name, a brief description, and metadata such as the creator's name, number of tags, and the last update date.

App Name	Description	Creator	Tags	Last Updated
surface-water-detection	Surface Water Detection	seonghapark	8 tags	Updated 139 days ago
avian-diversity-monitoring	Records environmental sounds, identifies birds by such sounds and f...	dariodematties1	1 tag	Updated 158 days ago
weather-classification	An app for identifying cloud or rain coverage from the ARM Doppler ...	rjackson	13 tags	Updated 161 days ago
traffic-state	Traffic State Estimator	seonghapark	5 tags	Updated 202 days ago
motion-analysis	Motion Analysis	seonghapark	6 tags	Updated 202 days ago
motion-detection	A general-purpose motion detection system that locates and tracks m...	seonghapark	2 tags	Updated 202 days ago
solar-irradiance	Solar Irradiance Estimator Using U-Net	seonghapark	3 tags	Updated 202 days ago
cloud-cover	U-Net Cloud Coverage Estimator	seonghapark	5 tags	Updated 202 days ago
object-counter	Object Counter	seonghapark	5 tags	Updated 202 days ago
cloud-motion	Cloud Motion Estimator for the Sky Camera	bhupendraraut	2 tags	Updated 202 days ago
wildfire-smoke-detection	Wildfire Smoke Detection	seonghapark	2 tags	Updated 259 days ago
sound-event-detection	Sound event detection (SED) plugin, using YAMNet audio classificati...	dariodematties	1 tag	Updated 278 days ago

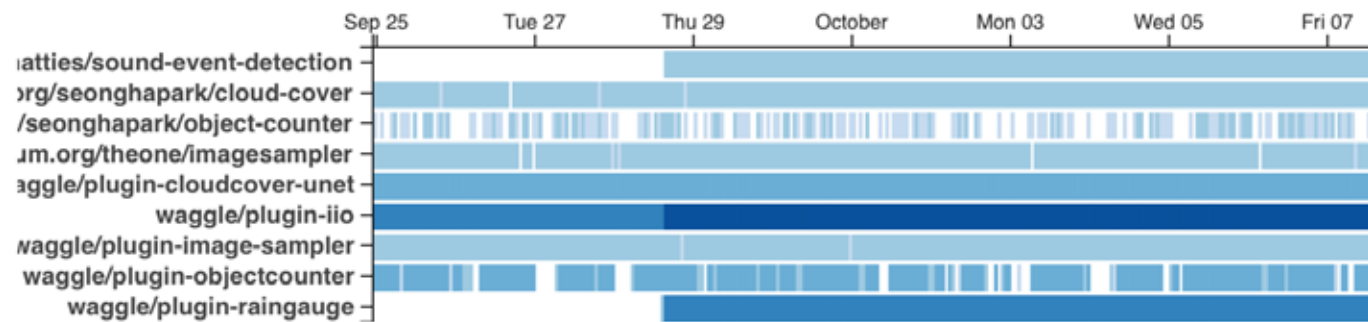
The Edge Code Repository





### W015

E. 87th & S. Cottage Grove, Chicago, IL





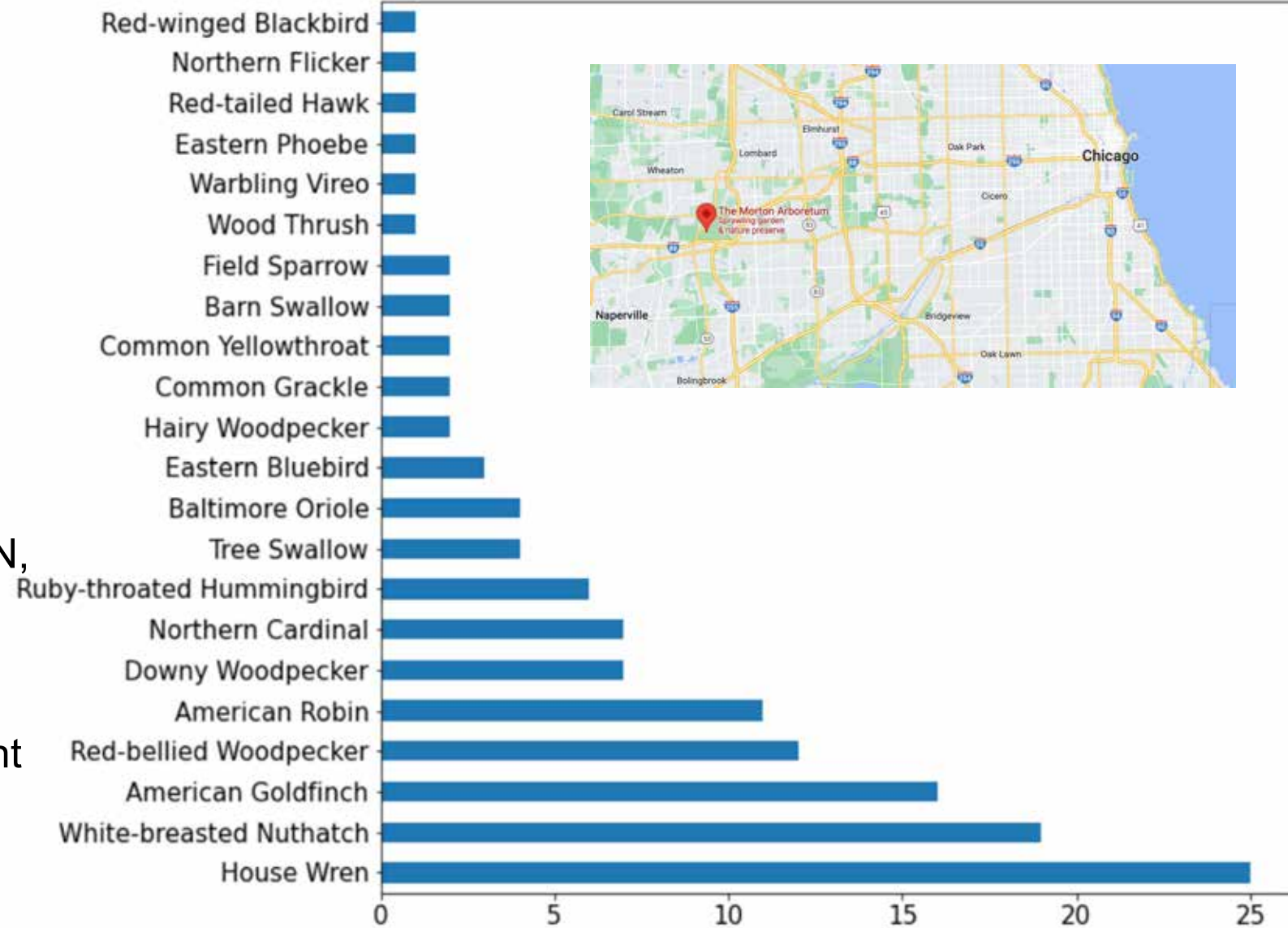
# Avian diversity monitoring



Image Creator: Becky Matsubara  
Copyright: © 2018, Becky Matsubara  
<https://creativecommons.org/licenses/by/4.0/>

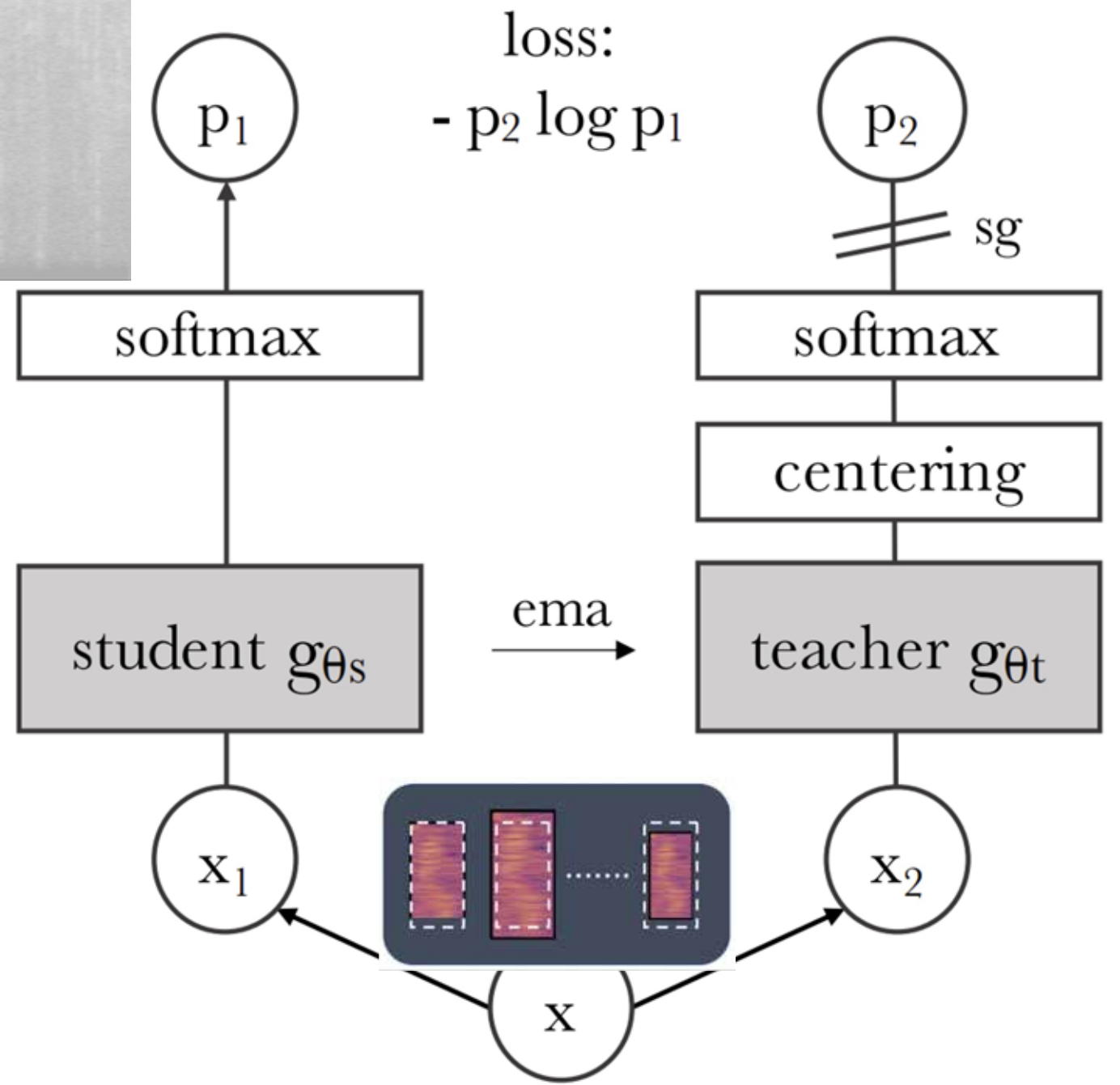
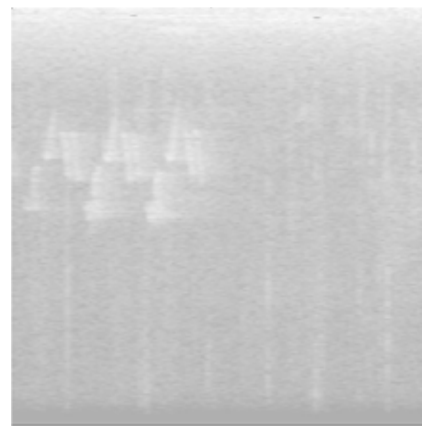
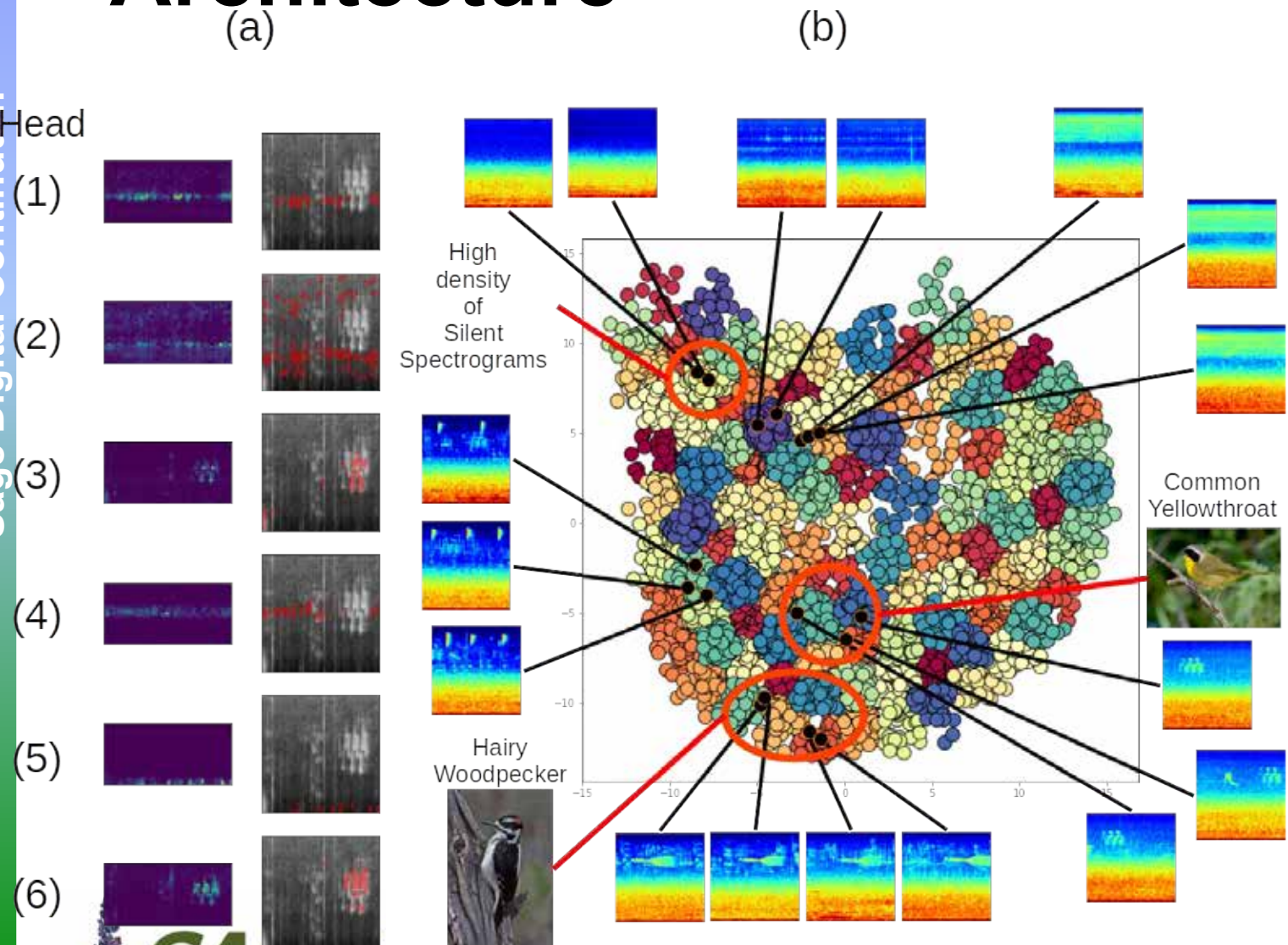
- Bird diversity changes as a metric to track the current environmental conditions
- We automate Avian Diversity Monitoring by using a DNN, called BirdNET [1], capable of identifying 984 North American and European bird species by sound. Weekly cumulative detections of non-migratory species occurrence was highly correlated with human point count observations
- It will be possible to get exposure to many organisms occupying diverse areas without needing to detect them during demanding and expensive human fieldwork

[1] Stefan Kahl, Connor M. Wood, Maximilian Eibl and Holger Klinck. BirdNET: A deep learning solution for avian diversity monitoring. Ecological Informatics Volume 61, March 2021.



Morton Arboretum Avian Detection, June 28, 2021 (24 hour)

# Self-supervised Avian Diversity Monitoring Joint-Embedding Architecture



Caron, M., Touvron, H., Misra, I., Jégou, H., Mairal, J., Bojanowski, P. and Joulin, A., 2021. Emerging properties in self-supervised vision transformers. In *Proceedings of the IEEE/CVF International Conference on Computer Vision* (pp. 9650-9660).

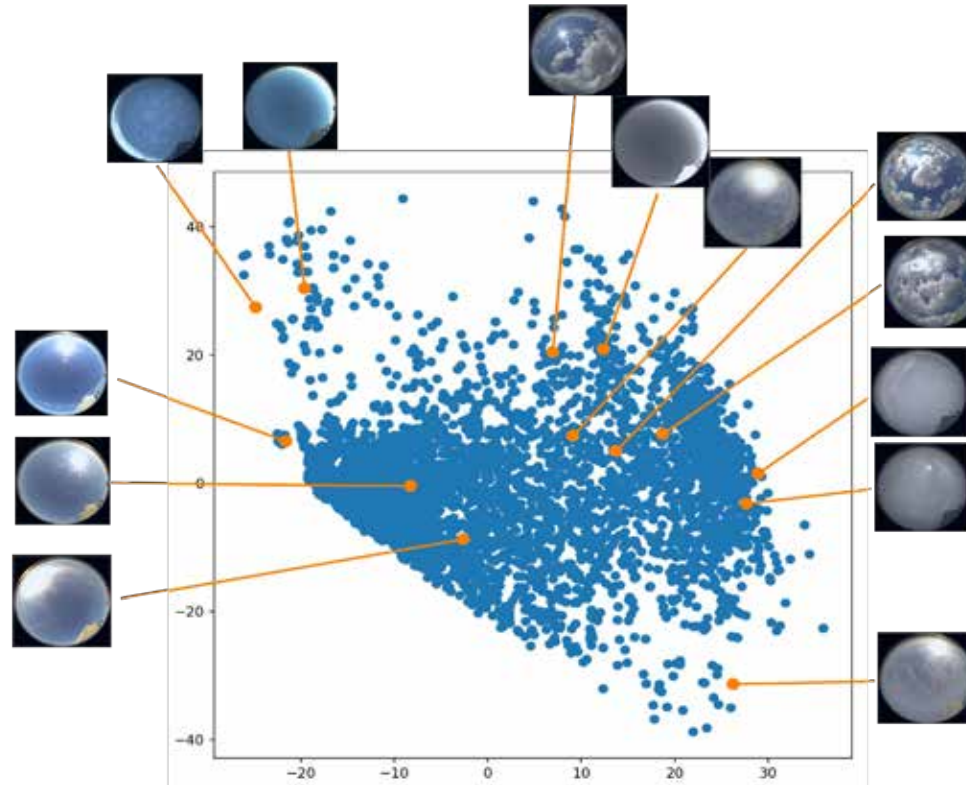


# Edge computing unleashed on understanding climate

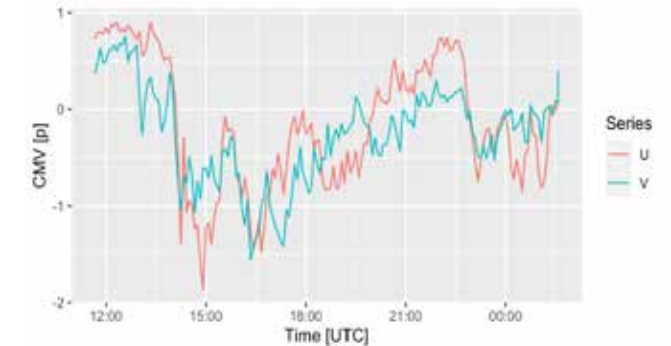


## Analyzing cloud patterns

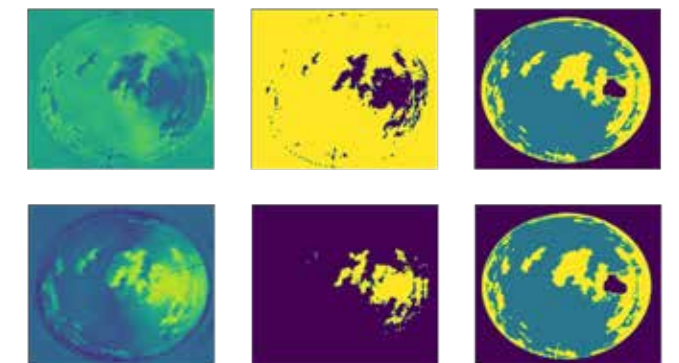
Self-learning Cloud Patterns



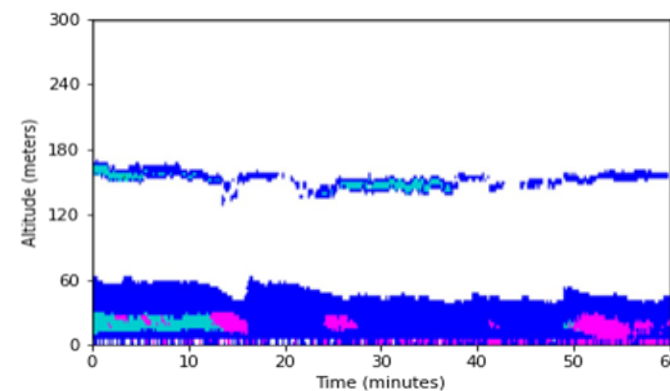
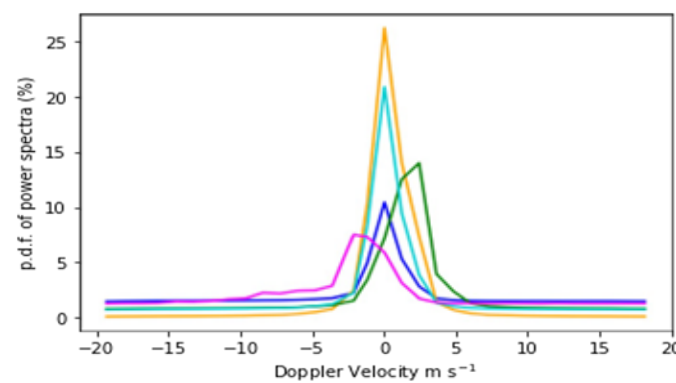
Cloud Motion Vector



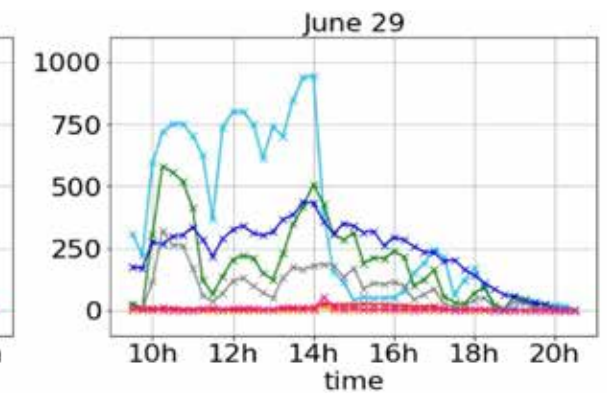
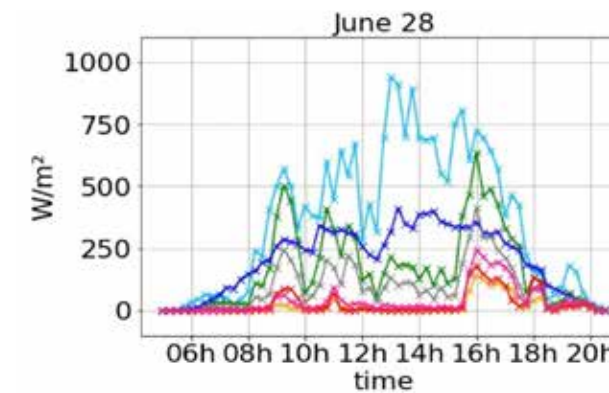
Cloud Cover Estimation



## Automated LIDAR particle identification



## Solar energy estimation





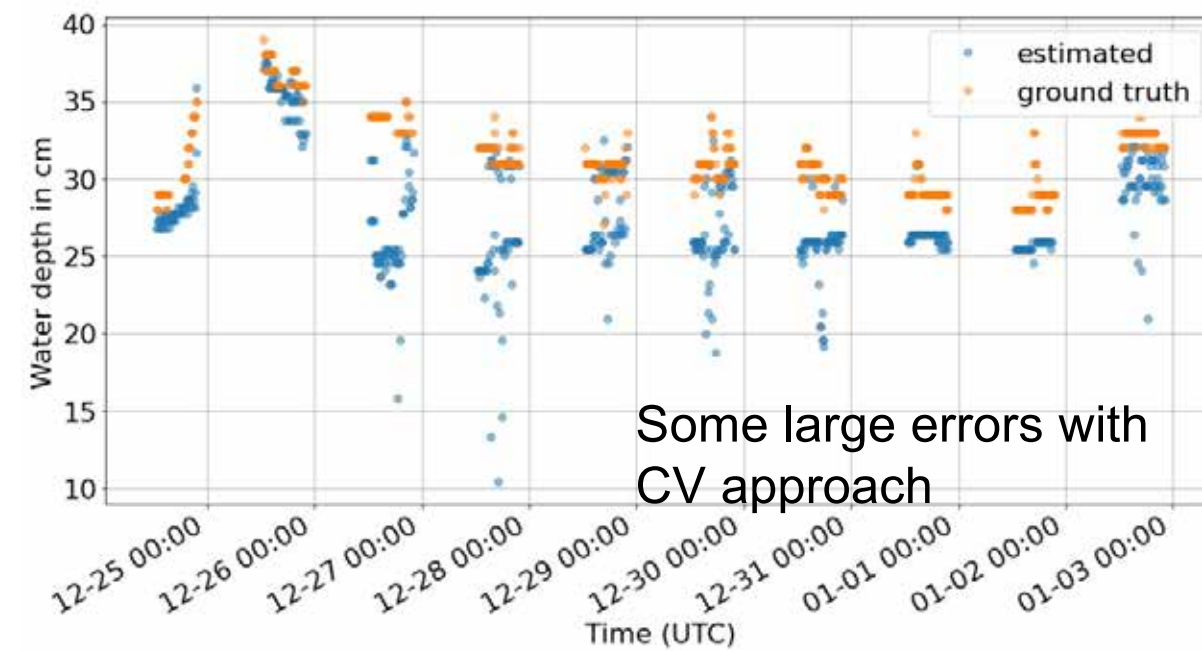
# Measuring Water and Snow Depth

We are evaluating multiple approaches to estimate the water (or snow) level from images of rulers (in of a stream at a NEON site)

- Computer vision (CV) based
- Machine Learning algorithms
  - U-Net, ResNet
  - Self-supervised Learning

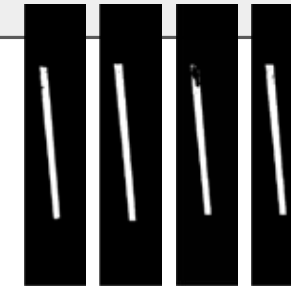


Human annotation using Labelbox

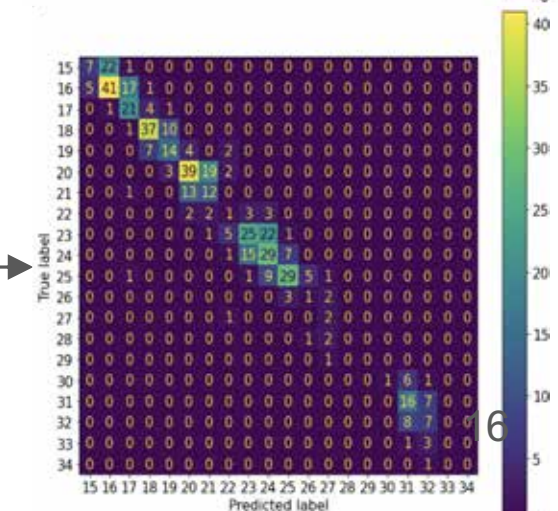
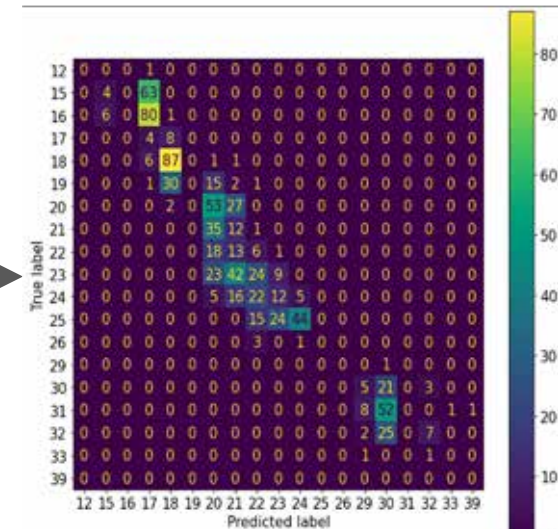


Some large errors with CV approach

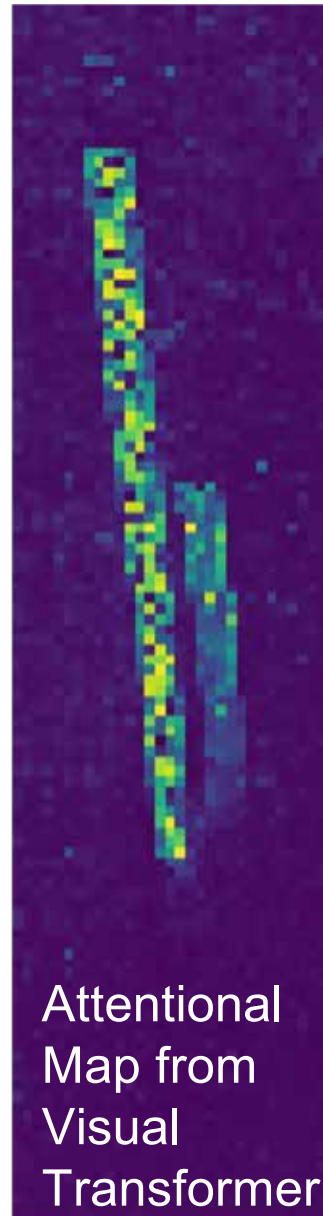
U-Net  
(segmentation)



ResNet32 based  
Regression




# Measuring Water and Snow Depth

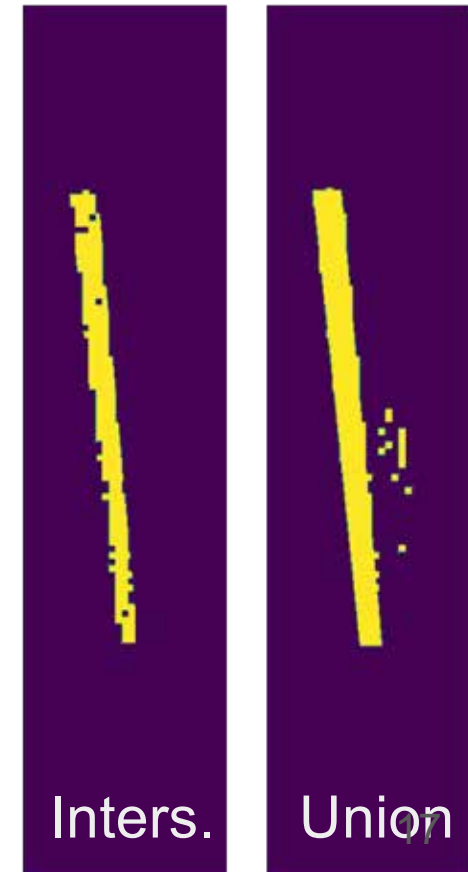


## Self-supervised Segmentation

- Exploring visual transformer ML
- ML model was trained using only images from IMAGENET (no labels and no NEON data)
- An Intersection over Union score  $> 0.5$  is normally considered a “good” prediction.

$$\text{IoU} = \frac{\text{Area of Overlap}}{\text{Area of Union}}$$


Intersection/Union  
(IoU) = 0.729





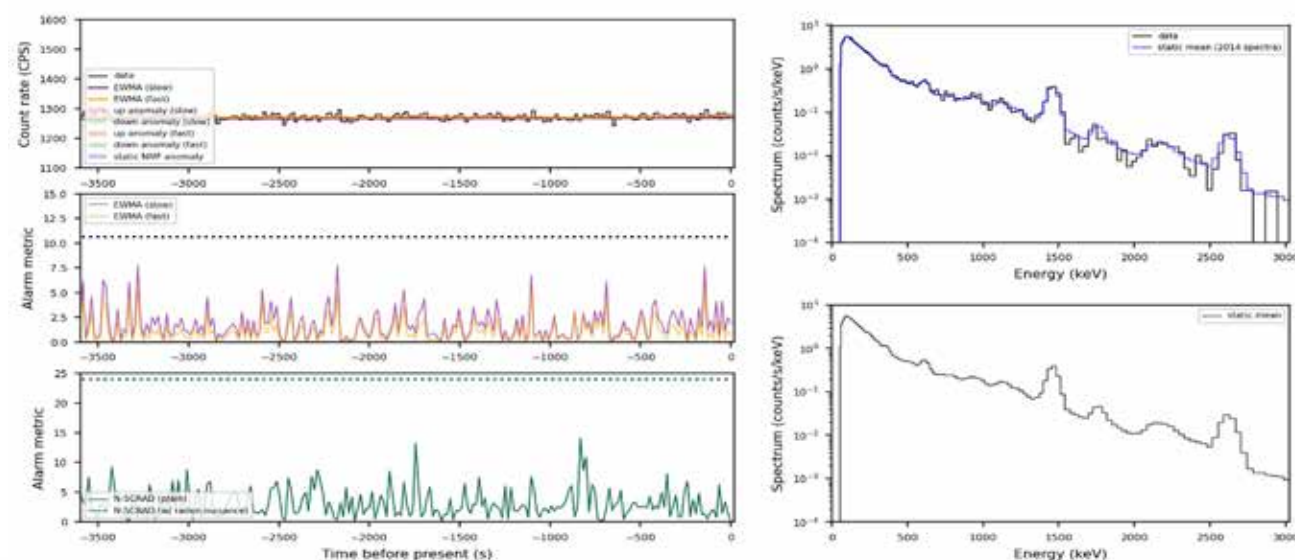
# PANDA (LBNL) and DAWN (ANL) (testing network of AI@edge – collaborative decisions)



Wildebbeest 5G  
project exploring  
integration

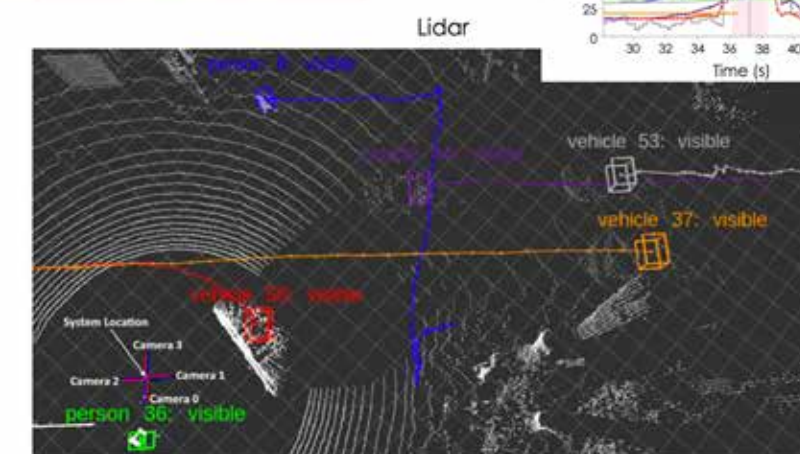
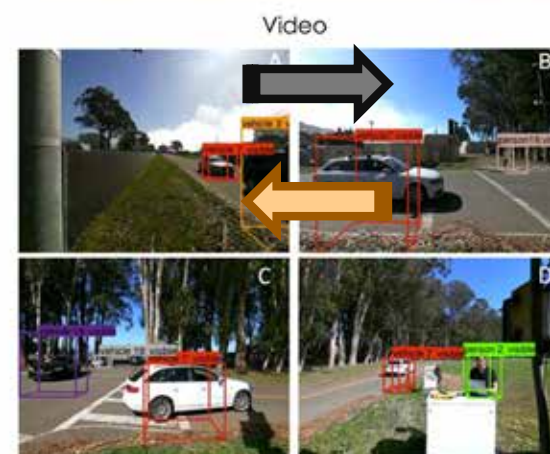
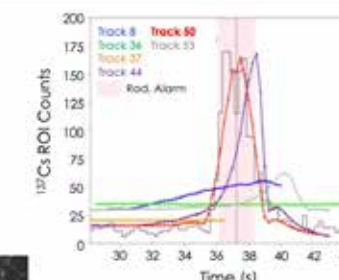
## Science begins at the instrument

- “Spectral Triage” separates radon progeny spectra from static background
- NMF background model learned ab-initio
- Model re-trained upon request



## Node Level Contextual Sensor Processing

- Video and Lidar-based object detection and tracking is used to attribute radiological anomalies to objects in the scene in real time
- Pre-trained Convolutional Neural Networks deployed at the edge for real-time object detection
- Provides localization, situational awareness, and improved detection sensitivity



M.R. Marshall et al., IEEE TNS DOI: 10.1109/TNS.2020.3047644

December 9<sup>th</sup> 2021

PANDA | EDUG 2021

8



# NNSA: 2 DAWN Nodes @ RUSH Hospital (briefings with stakeholders every 2 weeks)



**Node W01C | 000048B02D15C31F**

**Overview**

Project	DAWN
Focus	AM
Location	Harrison & Ashland, Chicago, IL
Built	2021-03-01
Commissioned	-
Registration	1/17/2022, 9:02:45 PM
GPS (from stream)	41.873822594, -87.566258729

**Hardware**

Shield	yes
Modem	yes
Modem Sim	n/a
Nx Agent	yes

**Cameras**

Top	-
Bottom	PTZ (XNV-0061Z)
Left	-
Right	-

**Last 45 days of data**

audio-sampler panda-plugin-rainbag plugin-image-tamper plugin-objectcounter plugin-rain-scope

**Sensors**

Sensor	Latest Time	Value	Last Hour
Temp	6.5 mins ago	33.95°C	
Humidity	6.5 mins ago	42.29%	
Pressure	6.5 mins ago	989.72 Pa	
Rain gauge	17.1 mins ago	1.68mm	

**Images**

Bottom

1657728959.9929066

Y (m)

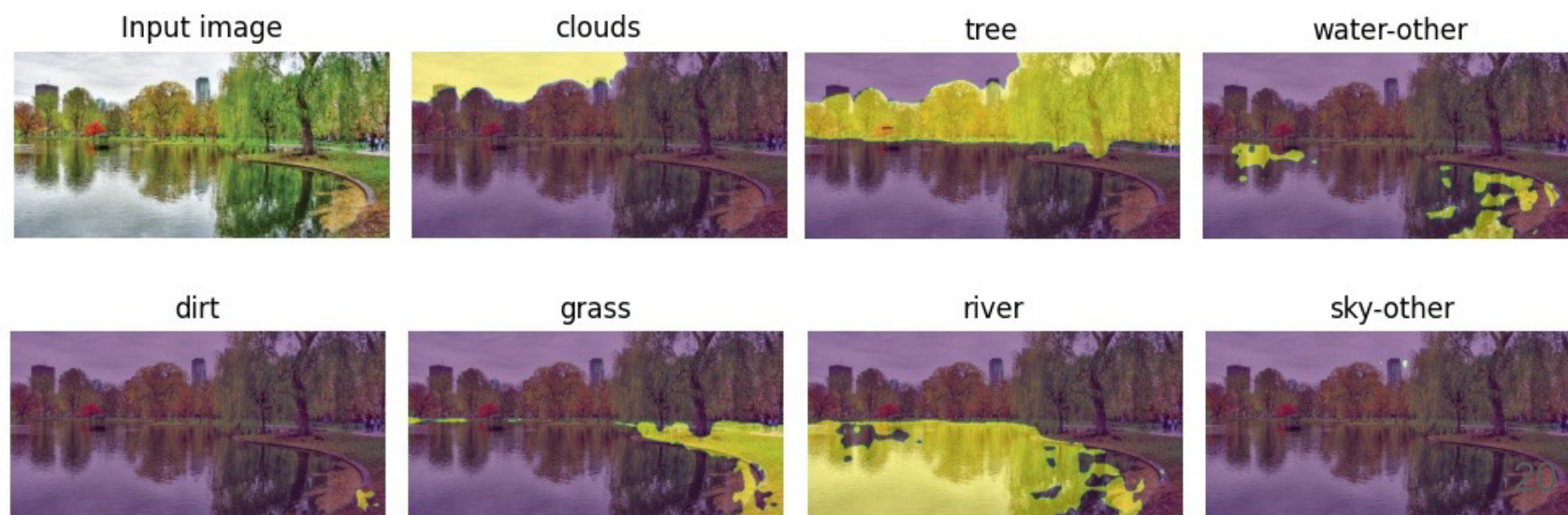
X (m)



# Surface Water Detection



**Linked with HPC, can be used to build hydrology models and predictive capabilities**

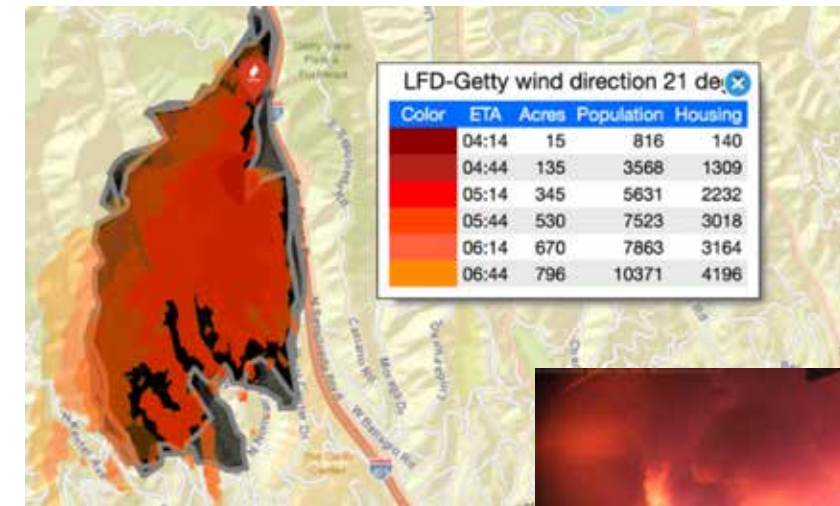




# Wildfire Detection and Prediction



## Exploring wildfire detection at the edge linked to HPC simulations



ALERTWildfire: A unique wildfire detection and monitoring system



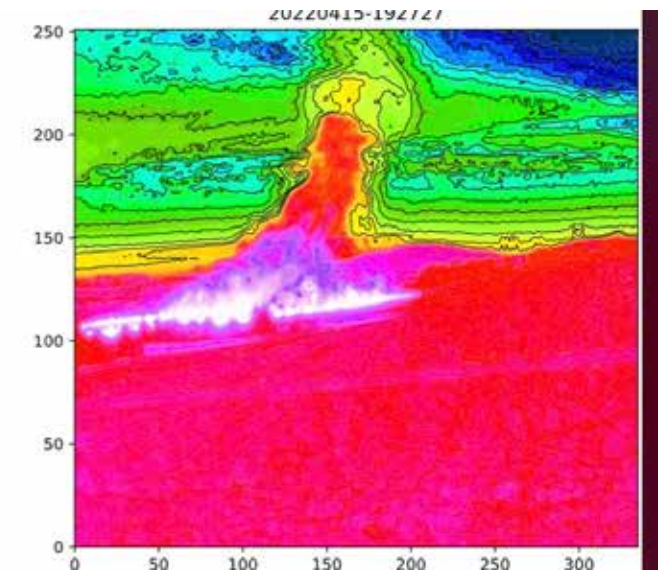
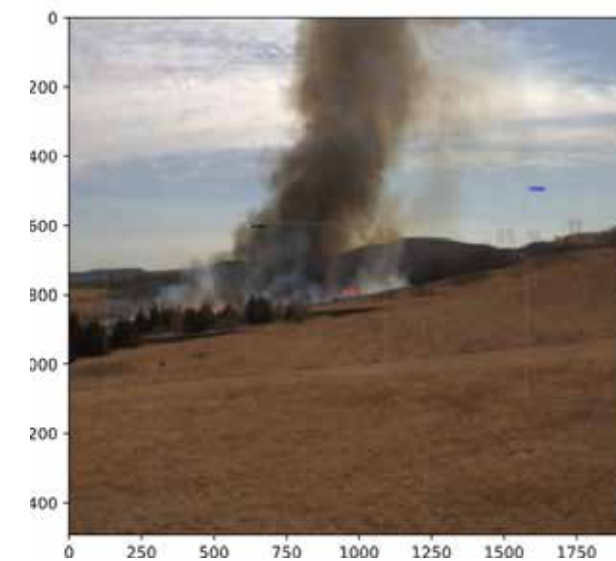
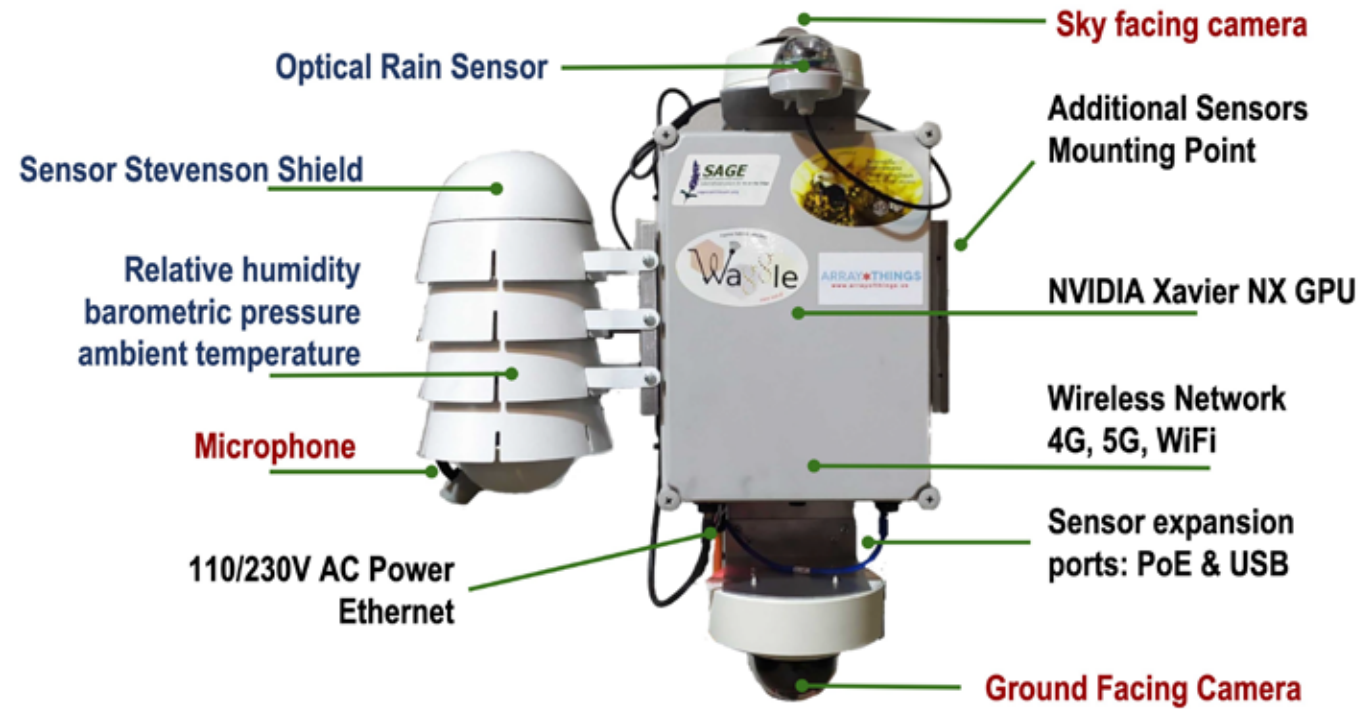
Collaboration: Doug Toomey, UOregon





# NEON Mobile Deployment Platform (MPD) with Sage Konza Prairie for controlled burn: April 2022.

Special Thanks: NEON Team!  
Rommel Zulueta @ Battelle



# **Sage NEON NSF Controlled Burn**

**Konza Prairie, Kansas**

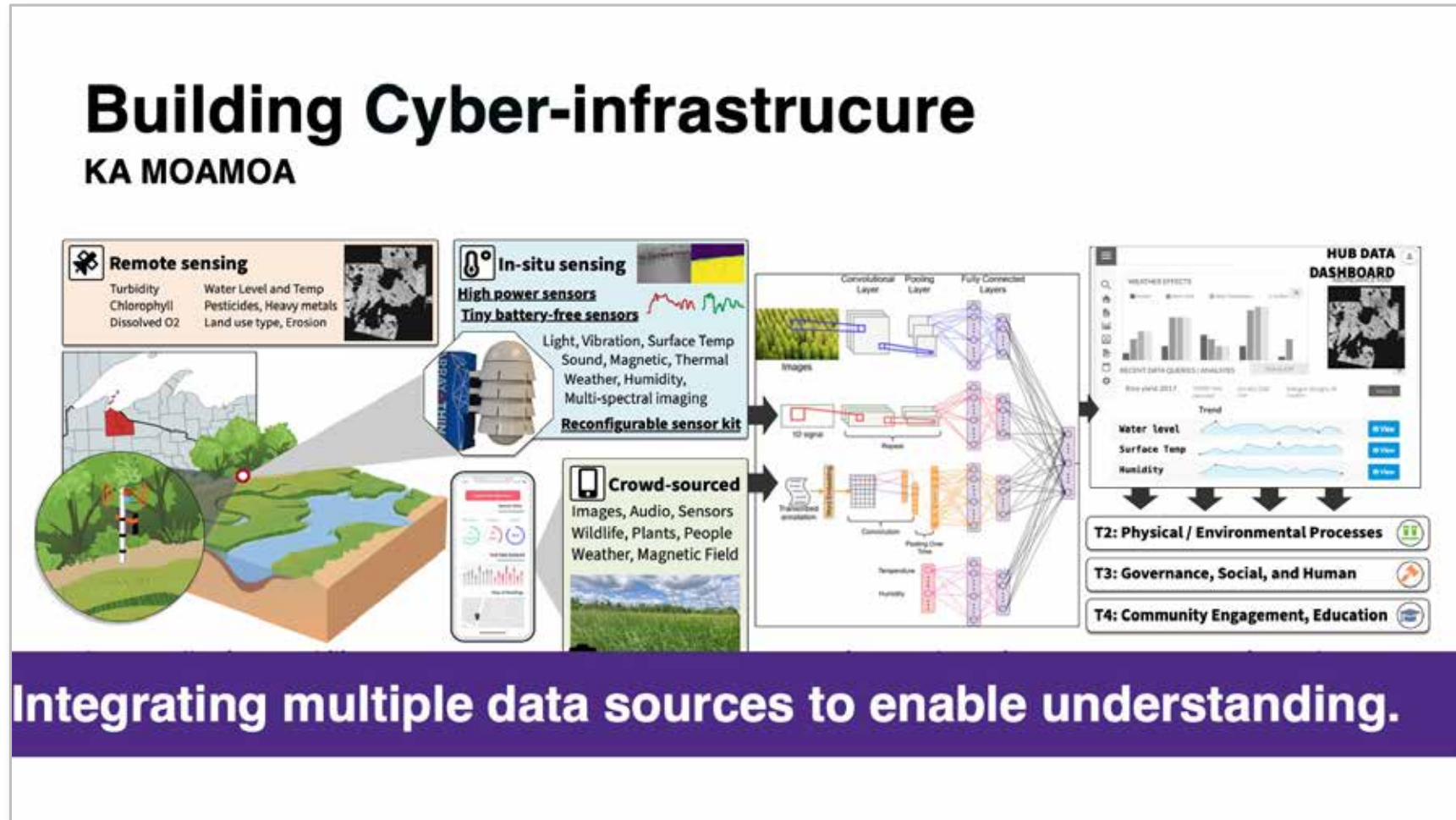






# NSF Coastlines and People

**Strengthening Resilience of Manoomin, the Sentinel Species of the Great Lakes, with Data-Science Supported Seventh Generation Stewardship**  
**PI: Josiah Hester**



**Jonathan Gilbert**, Biological Services Director, Great Lakes Indian Fish & Wildlife Commission (GLIFWC)



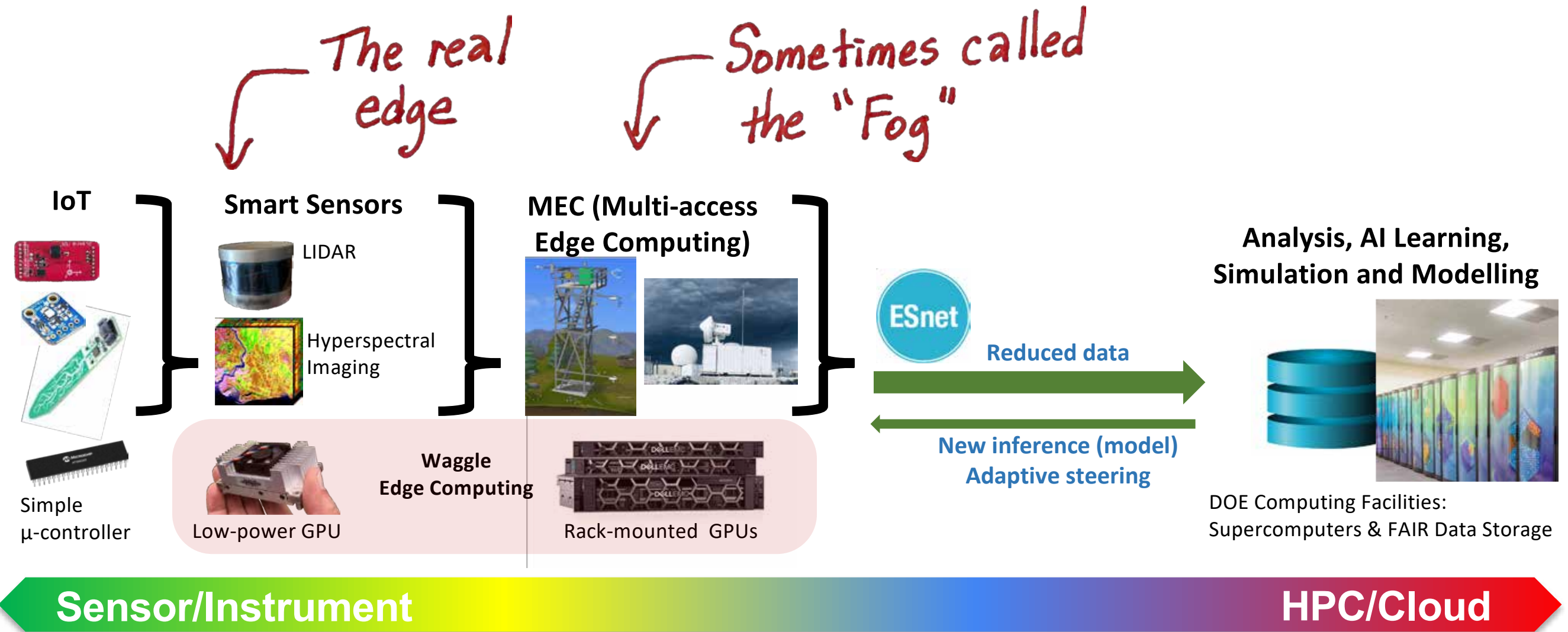


Can we leverage the tools and architectures from decades of experience building supercomputers for the continuum?



- AI@Edge open source software
  - Edge resource allocation
    - Node sharing & "Will it fit?"
  - Goal-based edge scheduler
  - Operating System management
  - Resilient data movers
- 
- HPC open source software stack
  - HPC allocation
    - CPU hours, storage, etc.
  - Job scheduler for shared access
  - Operating system management
  - Storage services & data movers

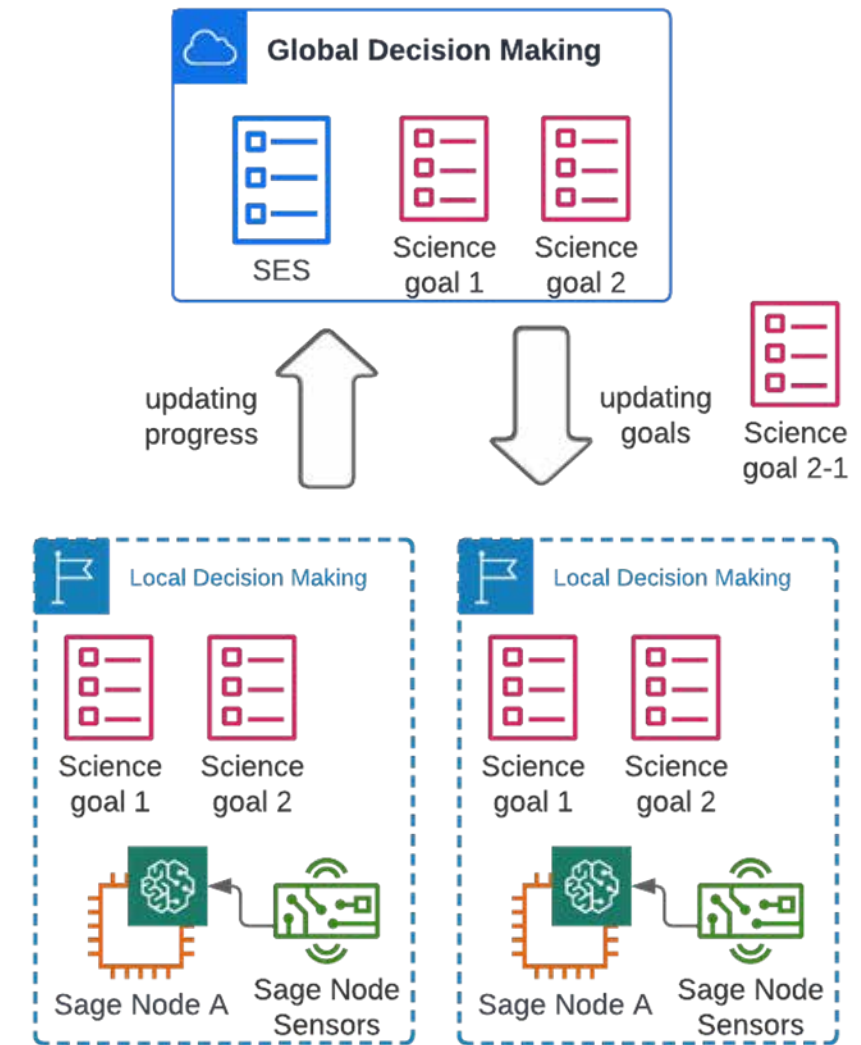
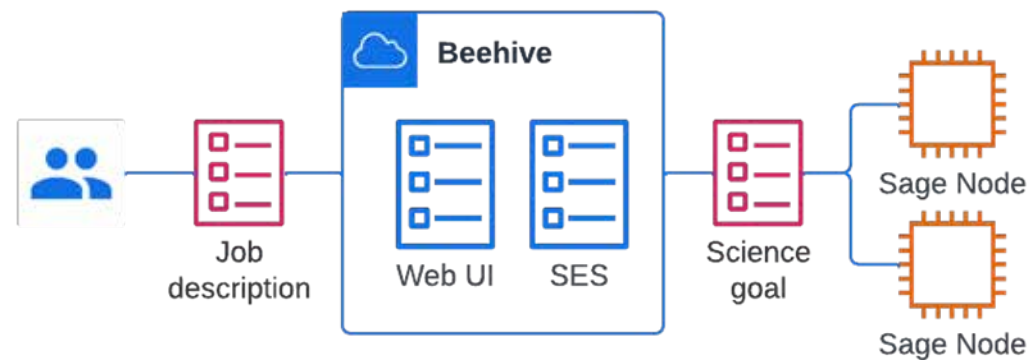
# A Challenge: We Need a Programming Model.... (think shmem :-)





# Multi-tenancy with Sage Edge Scheduler

- Jobs include a “Science Goal”
- Examples of science rules for different science studies
  - sampling: `cronjob("15m")`, `cronjob("14:00")`
  - trigger-based: `value("temperature") > 32`, `value("raingauge.uint") > 3`
  - event-based: `event("big-storm")`, `event("fire-season")`



For more detailed design, see paper:

<https://www.sciencedirect.com/science/article/pii/S0743731522001009>

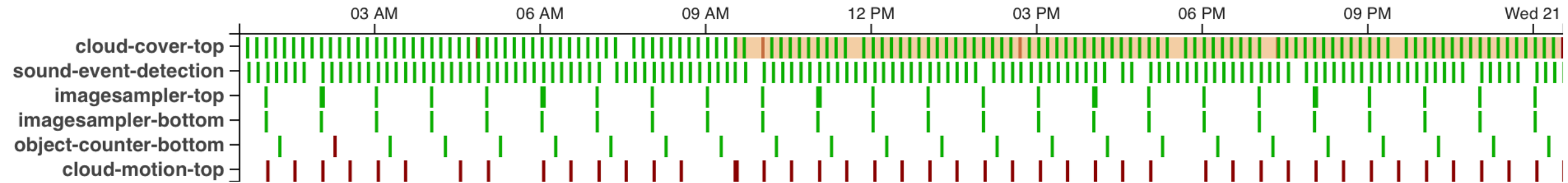
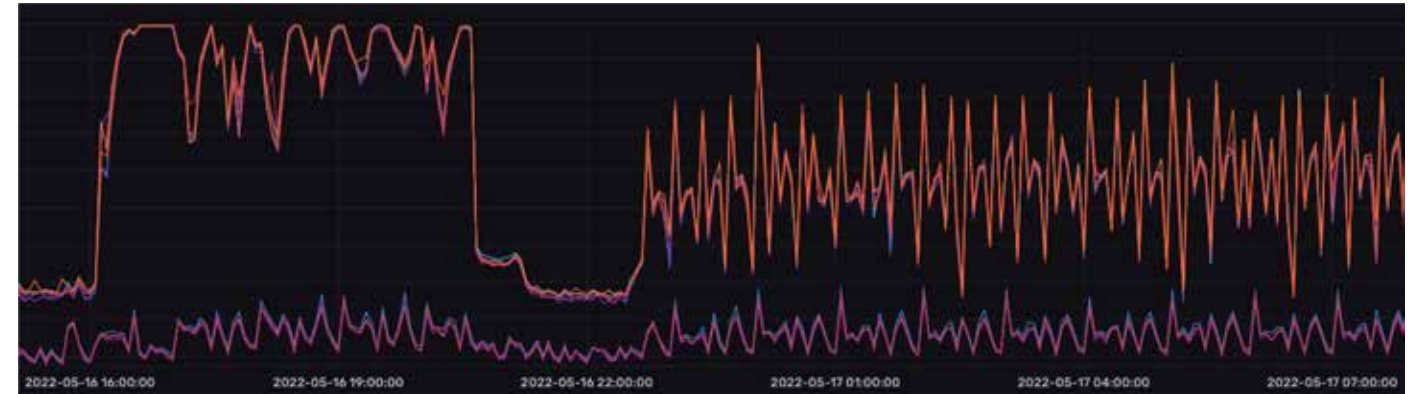
```
yongho.kim@node1:~$ sesctl stat
JOB_ID  NAME                               USER      STATUS      AGE
-----
9       sage-ttu-job                       waggle    Submitted   138h47m28s
10      sage-arm-ok-job                    waggle    Submitted   138h47m25s
11      sage_mt_image_sampler              yongho    Submitted   114h58m26s
2       vto-traffic-state                  yongho    Submitted   239h49m40s
6       sage-uoregon-job                   waggle    Submitted   138h47m39s
1       vto-video-sampler                  yongho    Submitted   305h59m36s
5       vto-job                             waggle    Submitted   138h47m46s
7       sage-urban-job                     waggle    Submitted   138h47m35s
8       sage-utah-job                      waggle    Submitted   138h47m31s
```

# Sharing the Node: Multi-tenancy

“Will it Fit” and automatic performance tuning  
Collaborative with with UOregon TAU team

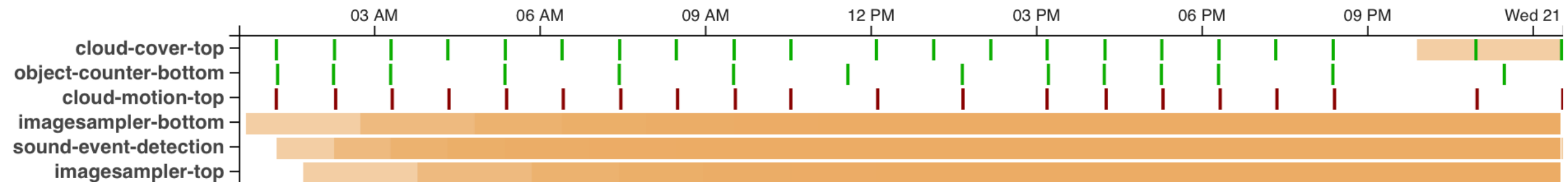
## W019

1264 Franklin Blvd, Eugene, OR



## W02B

3003 15th St, Lubbock, TX





# Tinkering with CLIP?

There's been a lot of excitement in the community about image *generation* using DALL-E 2 and Stable Diffusion...



CLIP is a building block used internally by some generative models to drive training by scoring how well an image matches a text description!

It's trained on the *many* text-image pairs available to provide meaning to text.



# How can we best use this? “Query@Edge”?

CLIP is useful for quickly experimenting with new kinds of detection without having to collect data and train a model - think minutes instead of hours or days!

```
python3 main.py \  
  "a person drinking coffee" \  
  "a person making a call" \  
  "a person jogging" \  
  "a construction crew fixing the road" \  
  "a red sports car" \  
  "a busy intersection"
```

```
12.499 0.000 a red sports car  
16.413 0.001 a busy intersection  
17.943 0.006 a construction crew fixing the road  
20.251 0.065 a person jogging  
21.546 0.237 a person making a call  
22.612 0.690 a person drinking coffee
```



# AI@Edge science problems for students.

- Measuring river depth against graduated marker
- Auto-steering of PTZ cameras based on local AI
- Measuring snow depth against graduated marker
- Measuring vegetative states, growth rates
- Self-supervised learning: IR, LiDAR, audio, and RGB
- Vehicle types and flow speeds
- Quantify flower blooming (color, count)
- Outlying conditions from previous sensor data
- Calculating biodiversity based on audio
- Measuring surface water coverage
- Measuring lightning via RF (software defined radios)
- Measuring visibility across a field
- Measuring rime ice thickness
- Measuring ice coverage on a large body of water
- Measuring water flow speed
- Classifying wildlife behaviors
- Improved wildfire detection algorithms
- Wildlife tracking in open fields (speed, direction, count)
- Ultrasonic bat detection
- Measuring pedestrian movement dynamics
- Measuring land changes (riverbeds, plant coverage)
- Measuring water turbidity, debris movement, floating waste
- Measuring vehicle dynamics: identification of sliding, crashes, mishaps
- Measuring bike usage, bike lane dynamics
- Identifying urban "near misses"
- Measuring bird flocks and dynamics

# Exciting, Hard, Challenging, CS Problems: From Instrument to the HPC/Cloud

- **Programming model for the Digital Continuum**
- Lightweight AI training, federated learning at edge
- Self-supervised learning with multiple instruments
- Microelectronics for low-power AI@edge
- Container technology for Cloud/HPC and the edge
- Cooperative sharing: multi-tenancy
- Control loops for actuation, steering
- Movement (drones, robots)
- Digital twin / MODEX for setting local edge goals
- Advanced networking: wireless and satellite

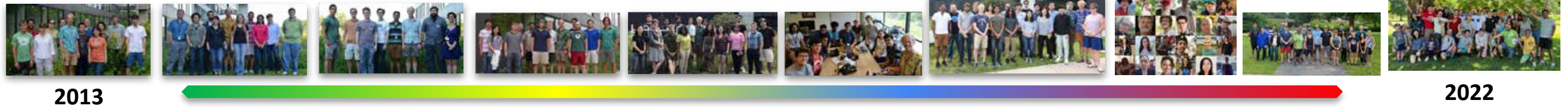


# Special Thanks



# arm Research

## Students!



Ilkay Altintas  
Kathy Bailey  
Daniel Balouek-Thomert  
Pete Beckman  
John Blair  
Eric Bruning  
Adam Brust  
Charlie Catlett  
Scott Collis  
Neal Conrad

Geoff Davis  
Dario Dematties  
Nicola Ferrier  
Jannick Fischer  
Larry Hartman  
Robert Jackson  
Eugene Kelly  
Yongho Kim  
Nick Maggio  
Seth Magle

Bill Miller  
Patrick O'Neal  
Jim Olds  
Aaron Packman  
Mike Papka  
Seongha Park  
Ismael Perez  
Bhupendra Raut  
Dan Reed  
Mike SanClements

Raj Sankaran  
Sean Shahkarami  
Sergey Shemyakin  
Joe Swantek  
Helen Taaffe  
Valerie Taylor  
Doug Toomey  
Frank Vernon  
Rommel Zulueta



# Questions

- Getting started with Sage!** - <https://docs.sagecontinuum.org>
- Sage AI@Edge Apps** - <https://portal.sagecontinuum.org/apps/explore>
- Sage Data** - <https://portal.sagecontinuum.org/data>
- Sage Konza MDP Campaign** - <https://mdp.sagecontinuum.org>
- Overall Sage system status** - <https://admin.sagecontinuum.org/status>
- Waggle Github** - <https://github.com/waggle-sensor>
- Sage Continuum Github** - <https://github.com/waggle-sensor>

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Gensburg-Markham Prairie, The Nature Conservancy  
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Dec 2015

