

Dynamic Capability Composition at the Digital Continuum from Edge to HPC

Keynote presentation for the Chesapeake Large-Scale Analytics Conference 2022
October 26, 2022 – Annapolis, MD

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Founding Director, **WIFIRE Lab**

University of California, San Diego

Joint Faculty Appointee, Los Alamos National Laboratory

UC San Diego



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at the UNIVERSITY OF CALIFORNIA SAN DIEGO
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ABOUT SDSC SERVICES SUPPORT RESEARCH & DEVELOPMENT EDUCATION & TRAINING NEWS & EVENTS

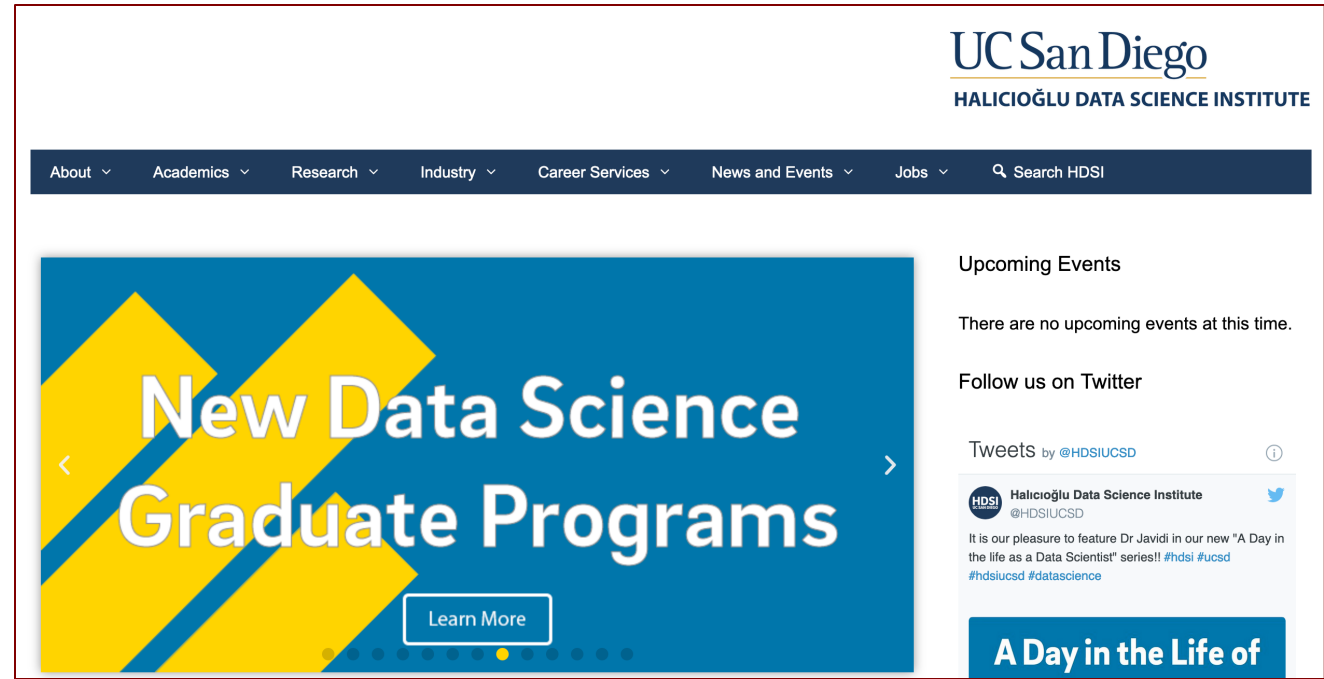
Search...

Continuing the Mission to Democratize 'ACCESS' to Cyberinfrastructure

SDSC, a national leader in high-performance supercomputing and data science, will apply its resources and expertise to the Advanced Cyberinfrastructure Coordination Ecosystem: Services & Support (ACCESS) program.

[READ MORE](#)

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Upcoming Events

There are no upcoming events at this time.

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HDSI Halicioğlu Data Science Institute @HDSIUUCSD

It is our pleasure to feature Dr Javidi in our new "A Day in the life as a Data Scientist" series!! #hdsi #ucsd #hdsiucsd #datascience

[A Day in the Life of](#)

Cyberinfrastructure and Convergence Research @SDSC

Translating cyberinfrastructure research for impact at scale

Convergence research is driven by a specific and compelling societal problem and works towards integrating innovative and sustainable solutions into society.

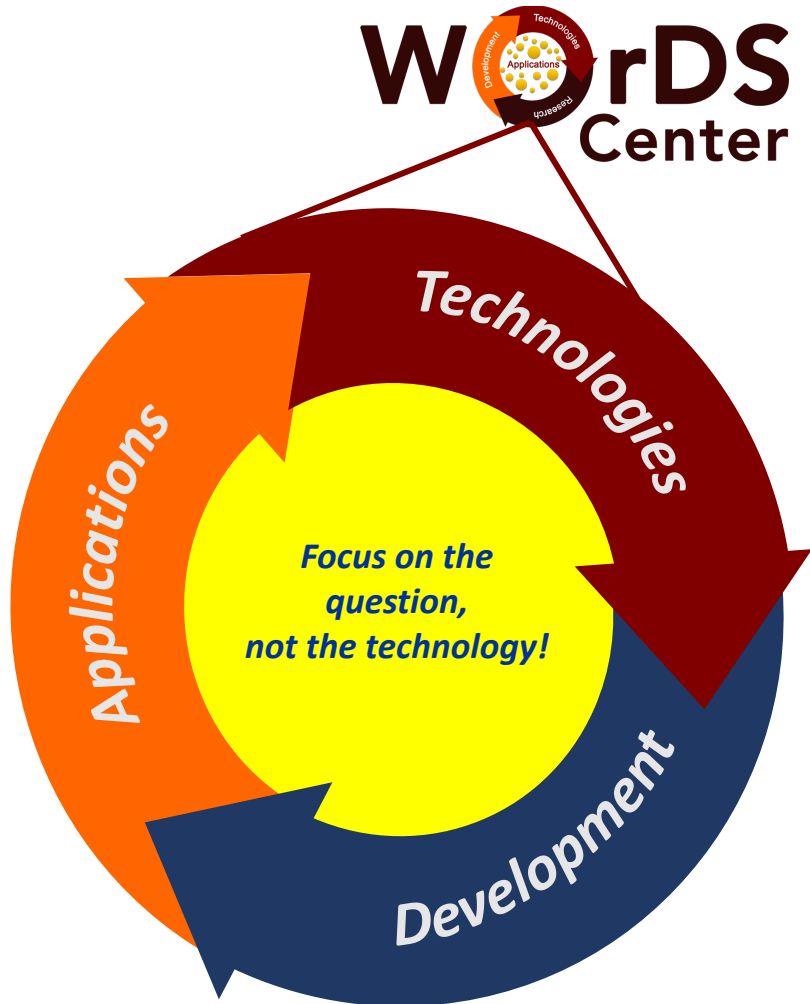
Why is data cyberinfrastructure and ML/AI critical?

- *Number of information sources*
- *Complexity of the data*
- *Number and needs of different users*



Workflows for Data Science Center of Excellence at SDSC

<http://WorDS.sdsc.edu>



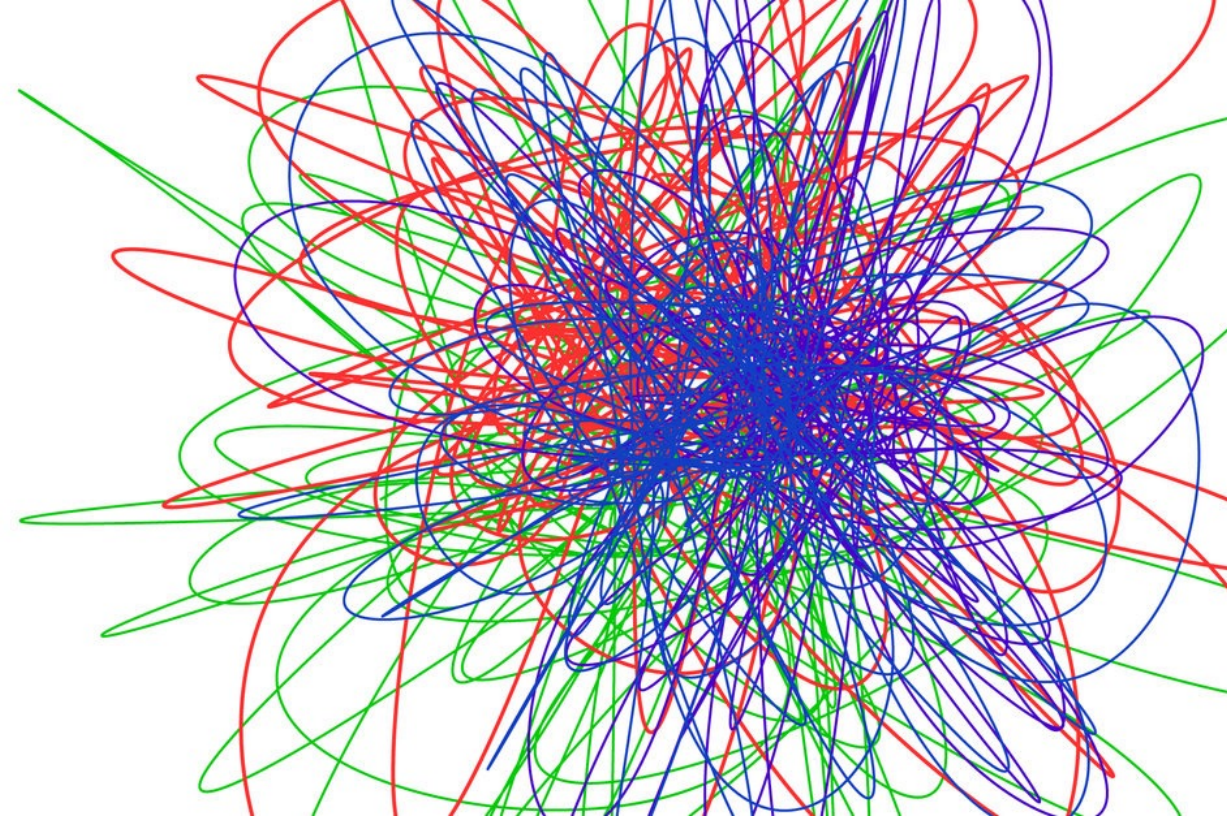
Research and Development Mission:

- *Methodology and tool innovation to enable collaborative workflow-driven science*
- *Create solutions on top of big data and advanced computing platforms.*

Evolving advanced computing and workflow ecosystem...

**“Big” Data, Computational Science,
Data Science, Cyberinfrastructure,
and Their Applications**

Application Context # 1



We are in the age of complexity!

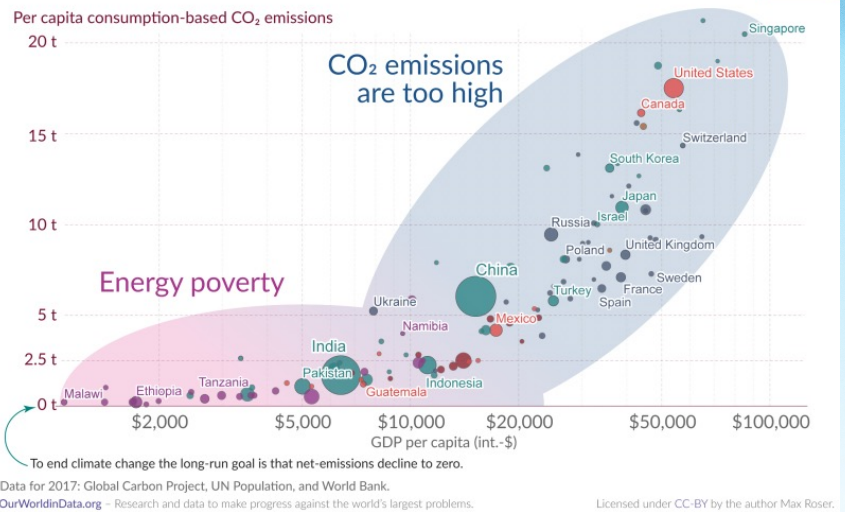
WCRP

GRAND CHALLENGES



CO₂ emissions per capita vs GDP per capita

Our World in Data



<https://ourworldindata.org/worlds-energy-problem>

\$10 trillion+ spent in global response to COVID-19

216 countries, areas or territories with cases

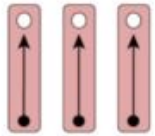
165+ COVID-19 vaccines being developed globally

<https://www.10xgenomics.com/research-areas/infectious-disease>

The biggest challenges of our time are too difficult to solve alone!

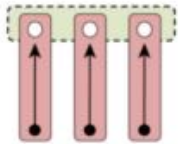
<https://www.wcrp-climate.org/learn-grand-challenges>

Advanced computing and data technologies can help, **and...** ...we need to collaborate effectively.



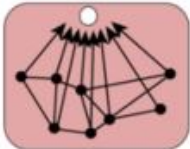
Disciplinary

- Within one academic discipline
- Disciplinary goal setting
- Development of new disciplinary knowledge



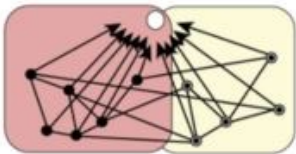
Multidisciplinary

- Multiple disciplines
- Multiple disciplinary goal setting under one thematic umbrella



Interdisciplinary

- Crosses disciplinary boundaries
- Development of integrated knowledge



Convergence

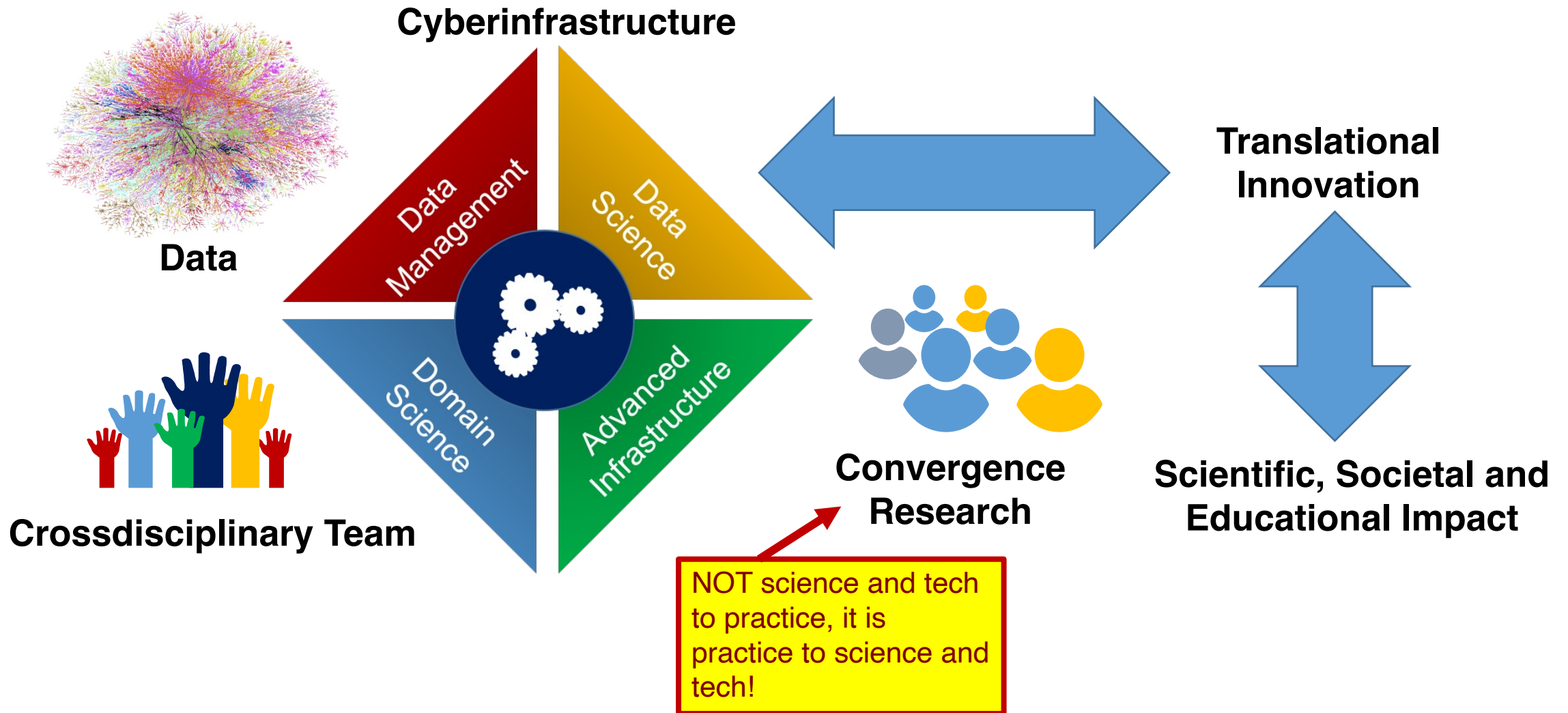
- Crosses disciplinary and sectorial boundaries
- Common goal setting
- Develops integrated knowledge for science and society
- Creates new paradigms



Adapted from Wright Morton, L., S. D. Eigenbrode, and T. A. Martin. 2015. Architectures of adaptive integration in large collaborative projects. *Ecology and Society* 20(4):5.

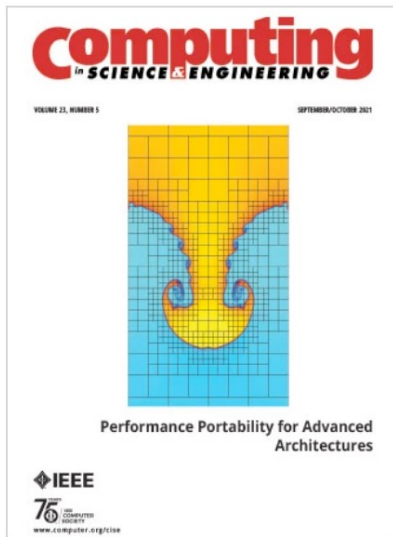


Building Cyberinfrastructure Systems for Impact



Translational Computer Science

<https://www.computer.org/csdl/magazine/cs/2021/05/09547673/1x9TDQr5c2c>



Home / Magazines / Computing in Science & Engineering / 2021.05

Computing in Science & Engineering

Translational Computer Science for Science and Engineering

Sept.-Oct. 2021, pp. 5-6, vol. 23

DOI Bookmark: [10.1109/MCSE.2021.3109962](https://doi.org/10.1109/MCSE.2021.3109962)

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**Systems should enable collaborative
teams and thinking!**

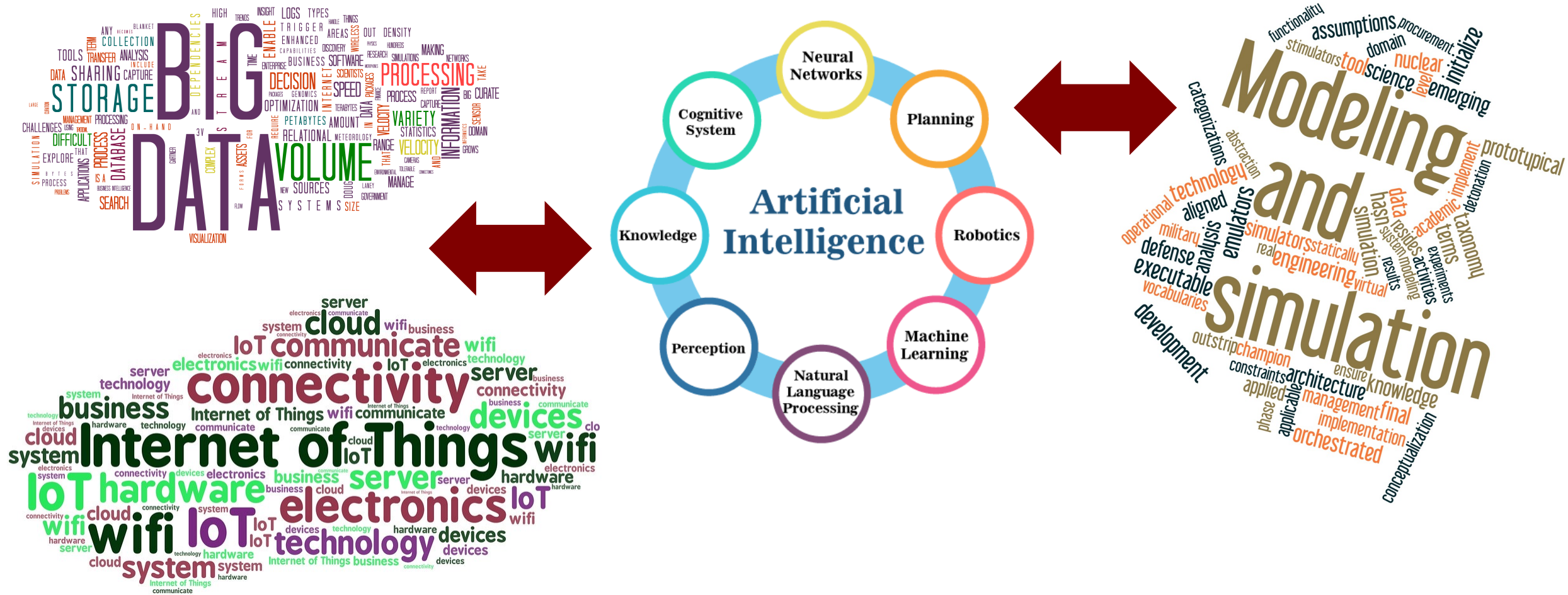
**Think "teamflows" along with
individual workflows.**



Application Context # 2

We are in the age of big data and AI!

AI-Integrated Applications at the Digital Continuum



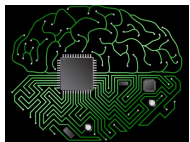
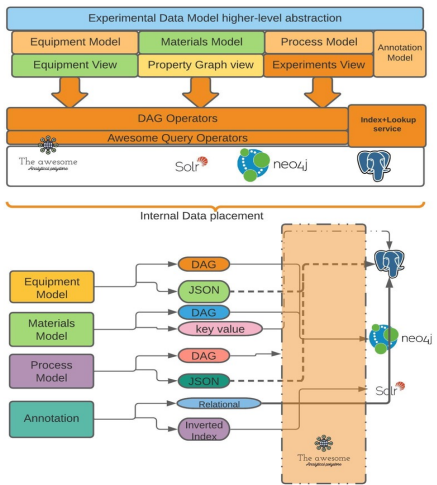
Big Data and IoT

Artificial Intelligence

Modeling and Simulation

Capability

Capacity



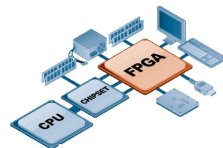
xPU → GPU, CPU, TPU, IPU, QPU, ...



Big Data



Edge

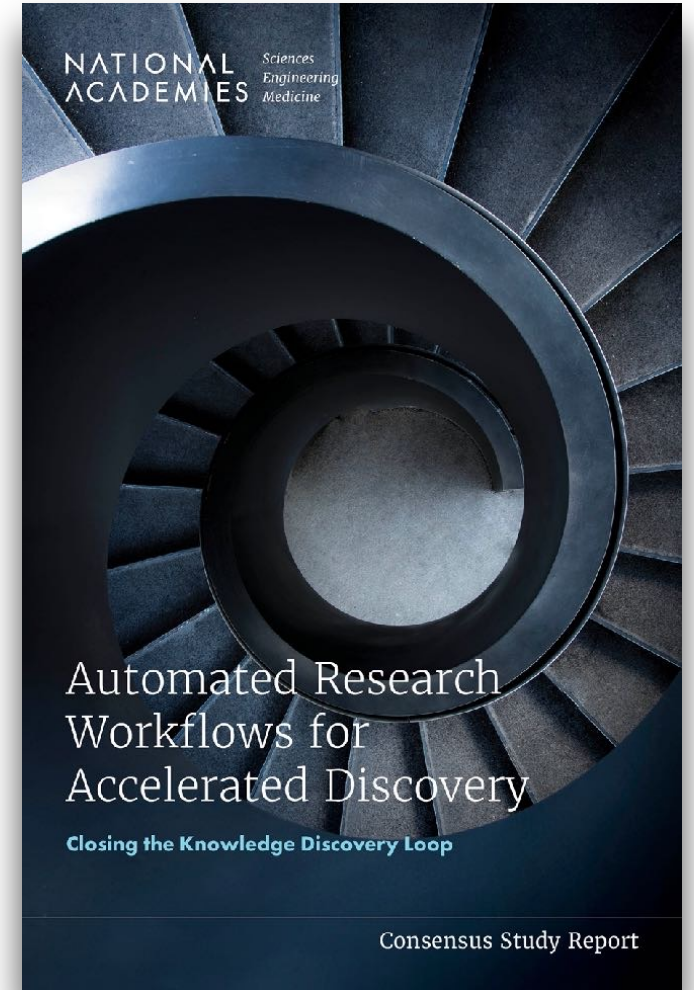


FPGA

Cloud, HPC, Storage

Evolution of Applications with AI

- Thinking in workflows moves AI closer to solutions in integrated applications.
- Teamwork is key to these workflows -- think workflows and teamflows!
- Enabling science and solutions requires open and transparent sharing and use of big data.
- Continuous integration of computing, data and AI into workflows needs dynamic systems capacity.



<https://www.nationalacademies.org/our-work/realizing-opportunities-for-advanced-and-automated-workflows-in-scientific-research>



**Systems should enable seamless integration of
AI-integrated application workflows!**

Application Context # 3

We are in the age of many systems!

August 7, 2020

Aug. 7, 2020 — This summer, five new National Science Foundation (NSF)-

ded advanced computing systems have been awarded to partner institutions across the country, all with their own unique specialties and applications. Once deployed, all of these systems will be primarily allocated through XSEDE to help connect researchers, regardless of physical location in the United States, to the system that best suits their research needs.

The recent announcements are a clear indicator of the NSF's renewed interest in HPC. NSF is a major indicator of the NSF's re-



ing the Fastest Computers
People Who Run Them

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our Authors

agement and operation of the resource

Jetstream 2 (Indiana University / Texas Advanced Computing Center / University of Arizona / Arizona State University / University Corporation for Atmospheric Research / University of Hawaii)

The NSF has awarded a \$10 million grant to support the development of a distributed cloud computing system to support research in artificial intelligence, and enhanced large-scale data analysis and visualization. The project is led by the Pervasive Computing Research Center at Indiana University. Jetstream 2 is a follow-up to the Jetstream project, which was funded in 2014 as the NSF's first project to support engineering research cloud system for the

Delta (National Center for Supercomputing Applications)

NCSA will integrate Delta into the national XSEDE infrastructure through XSEDE. Integration into XSEDE will allow researchers to access a substantial portfolio of services and support tools through XSEDE, and deliver unprecedented advances in research and education. This collaboration will promote synergy among campus, national, and commercial HPC resources.

Anvil (Purdue University)

Anvil, which is funded by a \$10 million award, will significantly increase the capacity available for research across a wide range of disciplines. Anvil will serve researchers for five years and will be supported by the NSF will support Anvil's operations and

Neocortex (Pittsburgh Supercomputing Center)

A \$5 million NSF award will allow PSC to develop performance artificial intelligence (AI) systems and fundamentally new hardware to greatly speed up research organization of Carnegie Mellon University. The new supercomputer, called Neocortex, will build the new supercomputer with Cerebras Systems and Hewlett Packard Enterprise.

Voyager (San Diego Supercomputing Center)

The NSF has awarded DSC at UC San Diego a high-performance resource for research across a wide swath of science. Called Voyager, the system will be the first NSF resource portfolio. In addition to the equivalent amount of funding is expected for the management and operation of the resource

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NCSA Gives Update on FABRIC Research Infrastructure Project

May 16, 2022

May 16, 2022 — When the internet launched in the last century it was an amazing feat of science and technology. But it was not designed for the massive data sets, machine-learning tools, advanced sensors and Internet of Things devices that have become central to many research and business endeavors and our homes.

Funded with a \$20 million grant from the National Science Foundation, FABRIC (Adaptive Programmable Research Infrastructure for Computer Science and Science Applications) is exploring ways to replace an internet infrastructure that's been showing its age for the last 20 years.

NCSA is one of 13 collaborating institutions helping create a platform for testing novel internet architectures that could enable a faster, more secure internet better suited for today's users and future needs. One that's also able to do things not possible now. The FABRIC project is led by the Network Research and Infrastructure Group at RENCi at the University of North Carolina at Chapel Hill.

NCSA: First with the Gigs

Last fall, NCSA installed a 100-gigabit network connection that's dedicated solely to the FABRIC project, the first FABRIC collaborator to do so.

NCSA already had six 100-gigabit internet connections, says David Wheeler, leader of NCSA's ICI Data Management and Delivery Division and the

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Some recent news...

Business Wire

GigaIO Awarded Lonestar6 Contract in TACC's First Bid for Composable Disaggregated Infrastructure

Business Wire



GigaIO Selected to Bring Composability to Bold New National Research Platform

June 6, 2021



The Case for an Edge-Driven Future for Supercomputing
By Oliver Peckham

September 24, 2021

"Exascale only becomes valuable when it's creating and using data that we care about," said Pete Beckman, co-director of the Northwestern-Argonne Institute of Science and Engineering (NAISE), at the most recent HPC User Forum. Beckman, head of an Argonne National Laboratory edge computing project called Waggle, was insistent on one thing: edge computing is a crucial part of delivering that value for exascale.

Beckman had opened with a quote from computer architect Ken Batcher: "A supercomputer is a device for turning compute-bound problems into I/O-bound problems." "In many ways, that is still true today," Beckman said. "What we expect from supercomputers is that they're so blindingly fast that really it's bottlenecked on either reading or writing from input or output."

"If we take that concept, though, and flip it over," he added, "then we end up with this idea that edge computing, therefore, is a device for turning an I/O-bound problem into a compute-bound problem."

San Diego, Calif., July 16, 2021 — Building on the successes of past collaborations with the San Diego Supercomputer Center (SDSC) located at San Diego, GigaIO, the creators of next-generation data center rack-scale computing solutions for artificial intelligence (AI) and high-performance computing (HPC), is proud to be announcing its low latency universal dynamic disaggregation solution, FabreX, was selected as the technology of choice for the new type National Research Platform (NRP). This National Science Foundation-funded cyberinfrastructure ecosystem is an innovative, all-in-one platform—computing resources, research and education networks, edge computing devices and other instruments—designed as a testbed to expedite discovery and enable transformative discoveries.

Complex computational and workflows underpin many of the most important scientific research challenges we hope to address through the NRP," says Dr. Frank S. Lee, PI of NRP, and co-director of the San Diego Supercomputer Center. "In areas such as public health, high energy physics and wildfire response, this research requires that we aggregate disparate computational elements, such as CPUs, GPUs, x86 processors and storage systems into highly usable and



The problems we are solving are ...

- data and AI-integrated
- heterogeneous
- collaborative

EXISTING SYSTEMS:

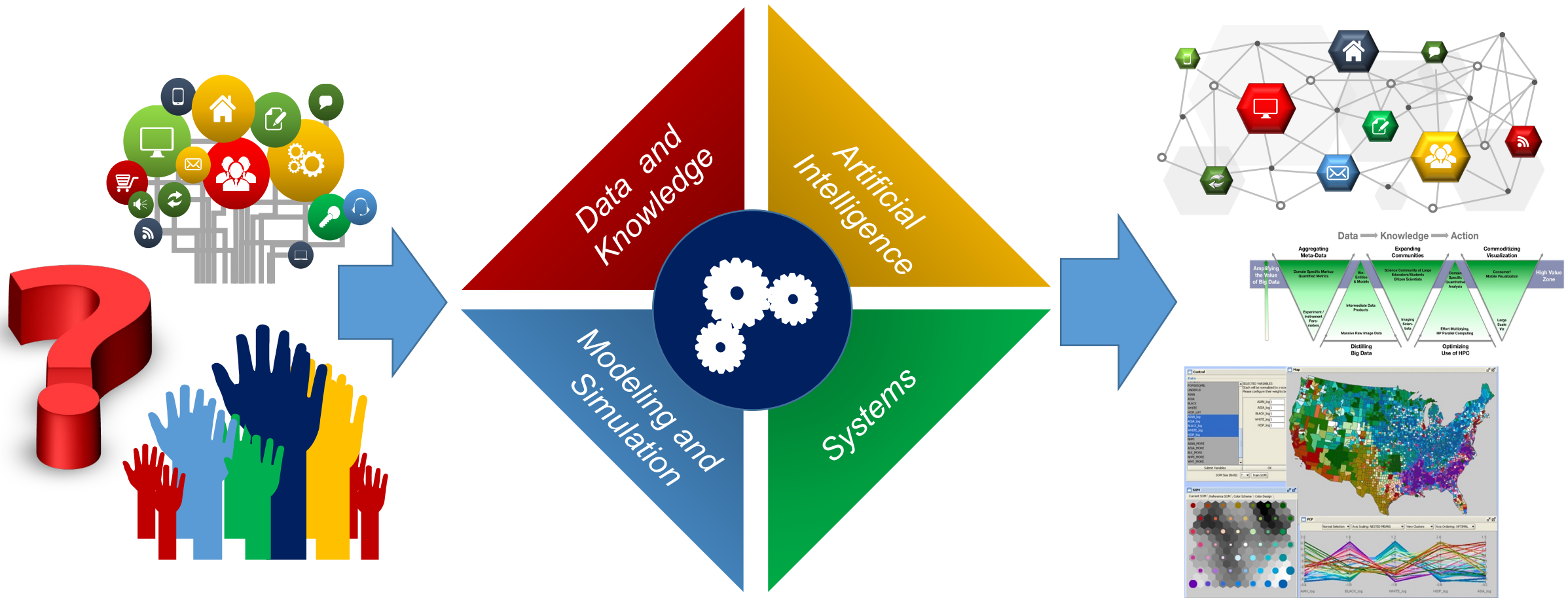
- **data** (storage + data/knowledge management)
- **continuum** of distributed computing (edge, cloud, HPC)
- high speed **connectivity** (programmable networks)

GROWING NEEDS:

- Interoperability
- Dynamic scalability
- On-demand interactive access
- Performance measurement
- Composability



**We need to think and build
strategic ecosystems
made up of reusable services
and solutions toward
capability integration!**



A Typical Application Integration Ecosystem

Integration requirements...



Dynamic composability matters.

Systems and services are only useful if groups can integrate them into applications.



TEAMWORK

Tools that enhance teamwork and use need to be coupled with responsible AI systems.



Dynamic composability matters.

COMPOSABLE SERVICES

e.g., model and data archives, learning and analytics, simulation, training

RESOURCE MANAGEMENT

e.g., container orchestration, optimization

COMPOSABLE SYSTEMS

e.g., GPU, CPU, Big Data, quantum, neuromorphic, SDN, storage

DATA LIFECYCLE MANAGEMENT

*e.g., active data repositories, long-term archives,
knowledge networks, data reuse services*

Systems and services are only useful if groups can integrate them into applications.



WORKFLOW MANAGEMENT

*e.g., application integration, coordination, optimization,
communication, reporting*

COMPOSABLE SERVICES

RESOURCE MANAGEMENT

COMPOSABLE SYSTEMS

RESPONSIBILITY

e.g., accuracy, privacy, explainability, ethics

REPRODUCIBILITY

TEAM SCIENCE

USE-INSPIRED INTERFACES

e.g., for science, education and scalable practice

Tools that enhance teamwork and use need to be coupled with responsible AI systems.

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e.g., accuracy, privacy, explainability, ethics

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e.g., GPU, CPU, Big Data, quantum, neuromorphic, SDN, storage

Smart Composability from Systems to Services

Use-Inspired Solutions

Services and Applications



Composition Software and APIs



Compute, Storage, Network

Data Management

- Dynamically measure, manage and provision resources
- Create plug and play microservices
- Run across many systems
- Capability-based integration
- Improved FAIR data capacity

Some Composable Systems

EXPANSE

COMPUTING WITHOUT BOUNDARIES
5 PETAFLOP/S HPC and DATA RESOURCE

HPC RESOURCE

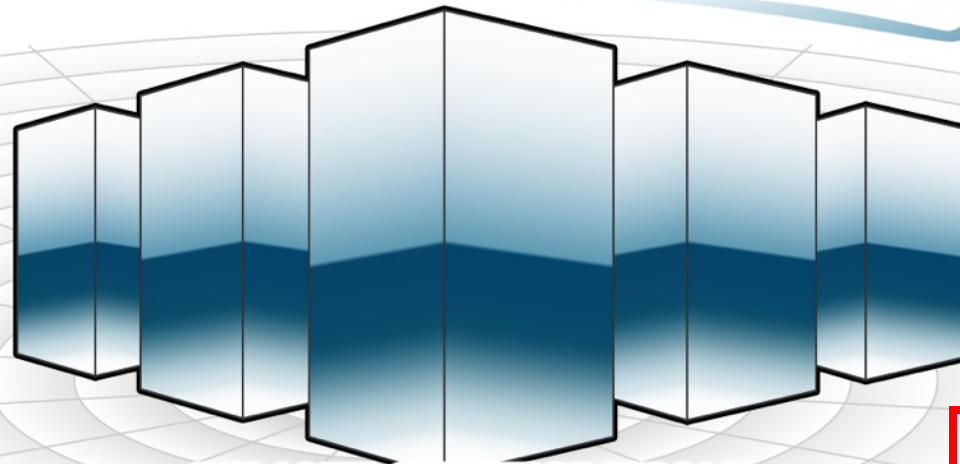
- 13 Scalable Compute Units
- 728 Standard Compute Nodes
- 52 GPU Nodes: 208 GPUs
- 4 Large Memory Nodes

LONG-TAIL SCIENCE

- Multi-Messenger Astronomy
- Genomics
- Earth Science
- Social Science

INNOVATIVE OPERATIONS

- Composable Systems
- High-Throughput Computing
- Science Gateways
- Interactive Computing
- Containerized Computing
- Cloud Bursting



REMOTE CI INTEGRATION

- ### DATA CENTRIC ARCHITECTURE
- 12PB Perf. Storage: 140GB/s, 200k IOPS
 - Fast I/O Node-Local NVMe Storage
 - 7PB Ceph Object Storage
 - High-Performance R&E Networking



- Two partitions exist in Expanse
 - Slurm
 - Kubernetes
- Dynamic provision via cm-scale (Bright Computing)
- User Web Portal (ACCESS)

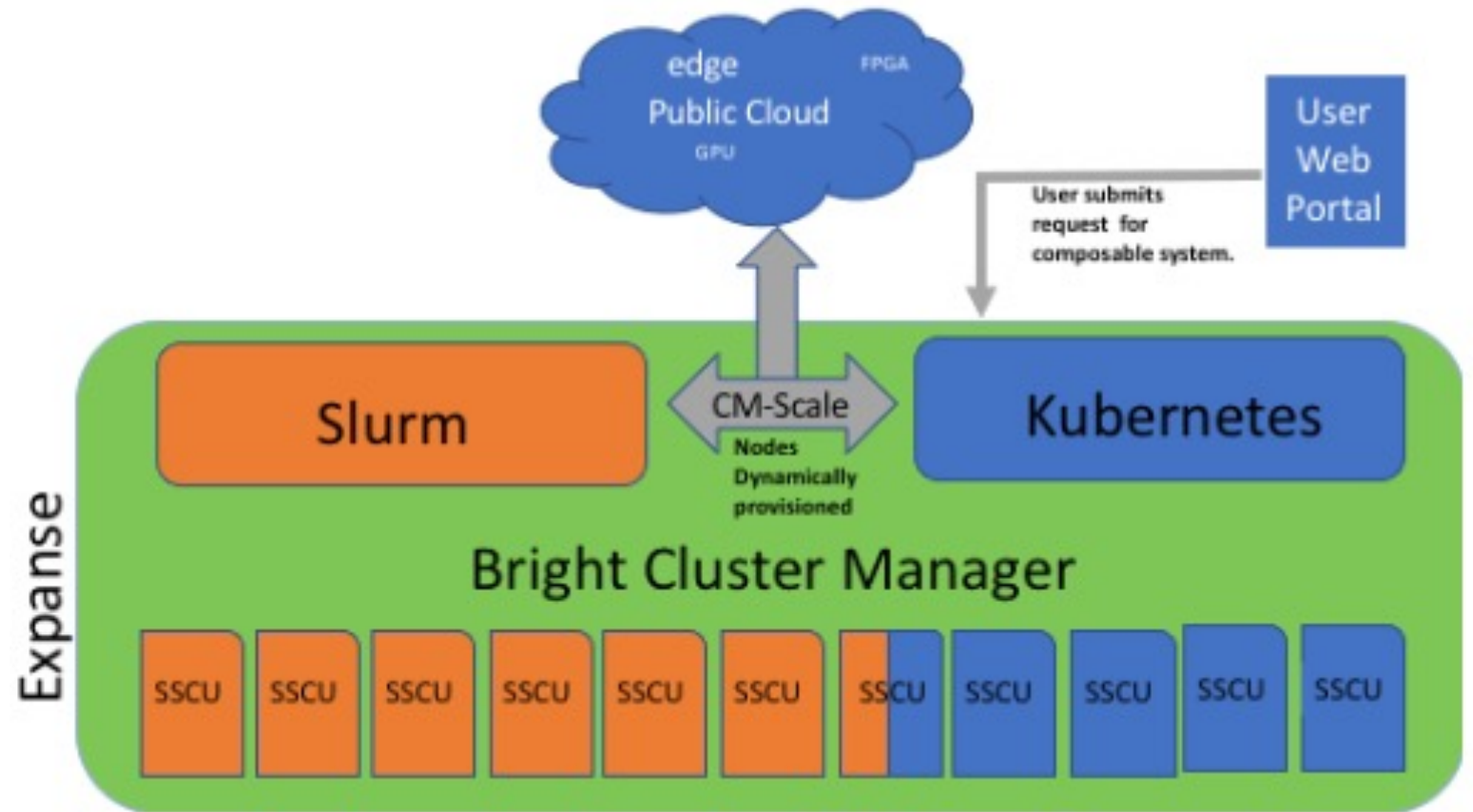


Figure 5.1 Expanse Composable Systems Framework

Nautilus – PRP and NRP



Image Source: <https://pacificresearchplatform.org/nautilus>

Nautilus is a Multi-Institution Hypercluster Connected by Optical Networks



*~200 FIONAs on 25 Partner Campuses
Networked Together at 10-100Gbps*

Installed CPUs

7870

Installed GPUs

549

Rotating Storage

4000 TB

Namespaces

600

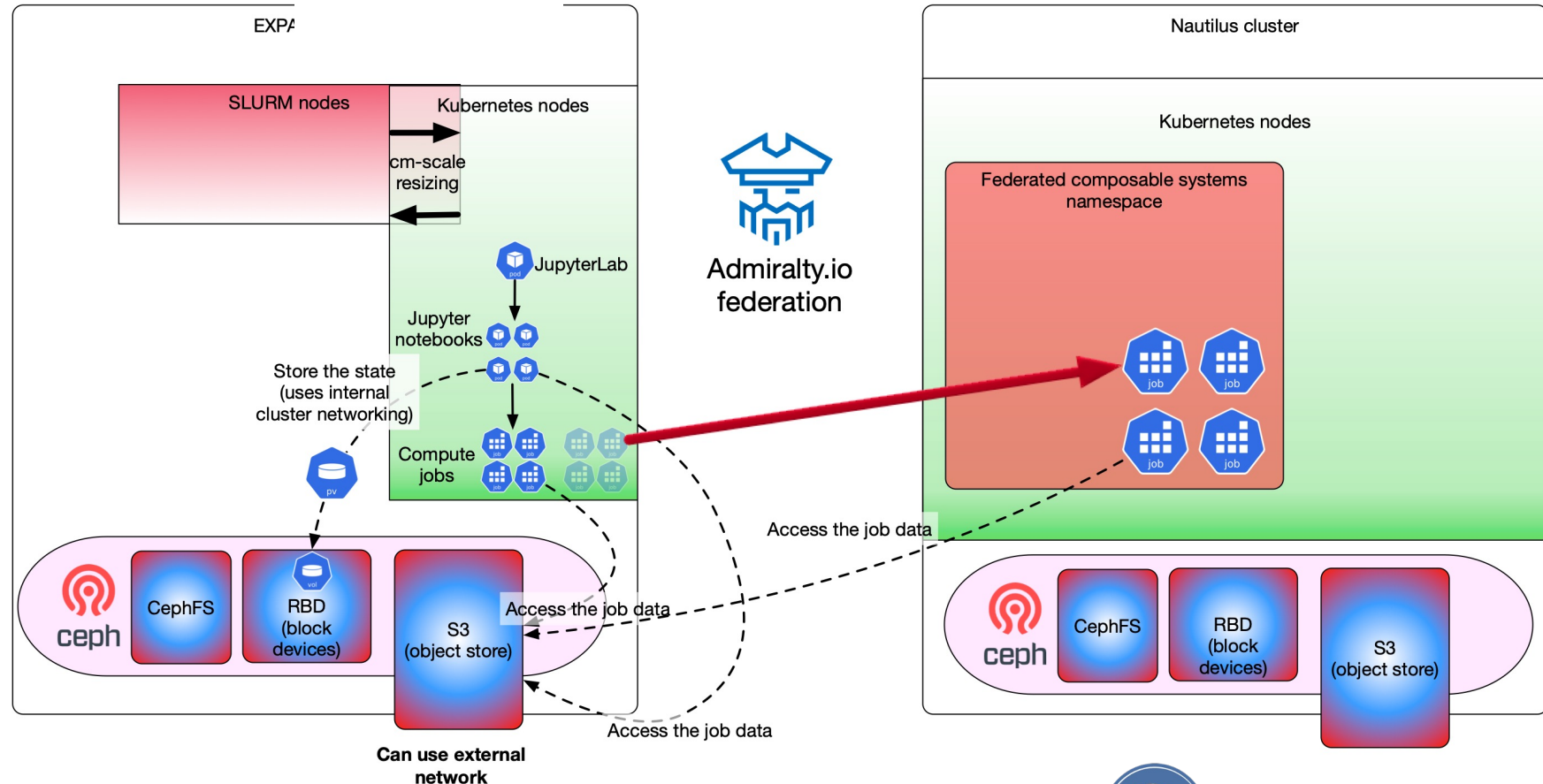


Slide Source: Larry Smarr, UCSD

Composable System using Admiralty for Federation

EXPANSE (Enthalpy) + CHASE-CI (Nautilus)

**EXPANSE
(Enthalpy)**



Admiralty Federation Topologies

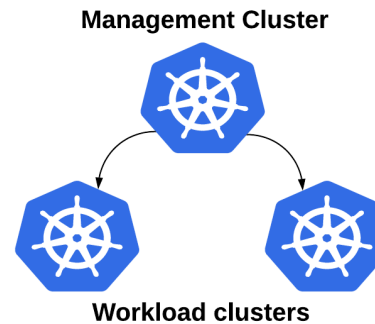
- Topologies

- a) Central control plane

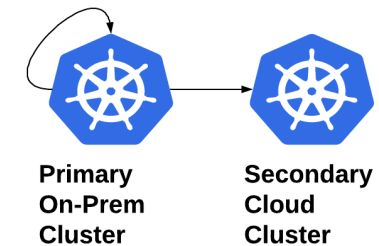
- b) Cloud Bursting

- c) Decentralized

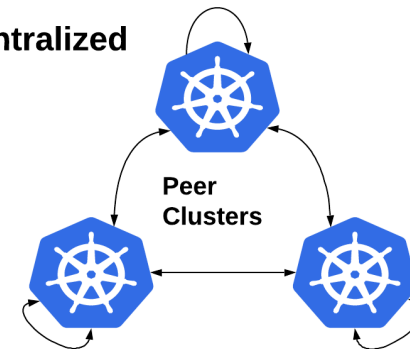
a) central control plane



b) cloud bursting



c) decentralized

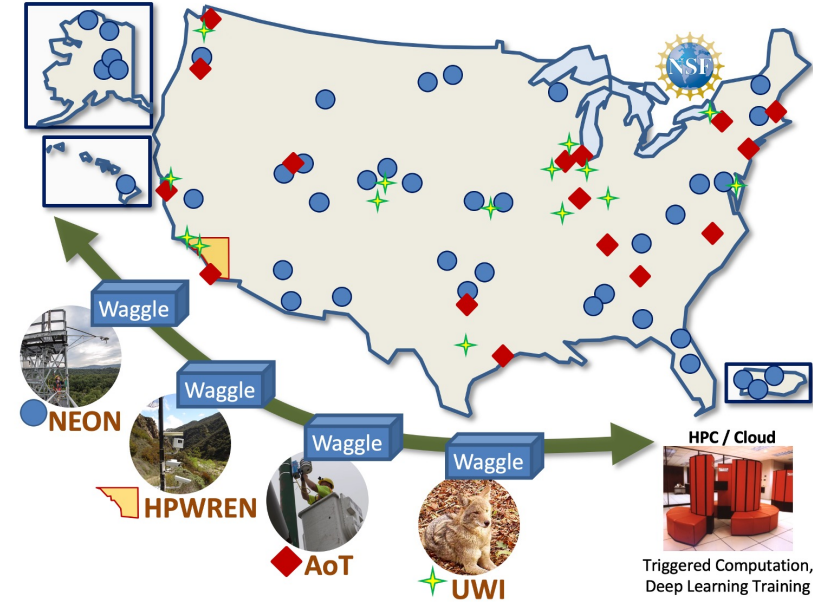


AI@Edge and the Digital Continuum

Slide Source: Pete Beckman, ANL



SAGE
Cyberinfrastructure for
AI at the Edge
sagecontinuum.org



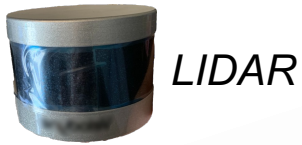
Leadership Team

 Pete Beckman (NU; Director)	 Nicola Ferrier (NU; Deputy Dir.)
 Ilkay Altintas (SDSC; Data)	 Charlie Catlett (Illinois; AoT)
 Scott Collis (NU; ARM)	 Valerie Taylor UChicago; Broader Impacts)
 Jim Olds (GMU; Life Sci, Risk)	 Dan Reed (Utah; Architecture)
 Eugene Kelly (CSU; NEON)	 Irene Qualters (LANL; Advisory Committee Chair)

Education & Training



Sensors



LIDAR



Software Defined Radios

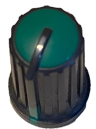


Hyperspectral Imaging

Facilities



Actuators



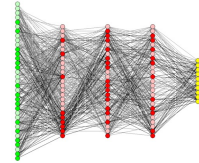
Servos

Dynamic adaptation



Scientific Digital Continuum in Sage

Edge Computing

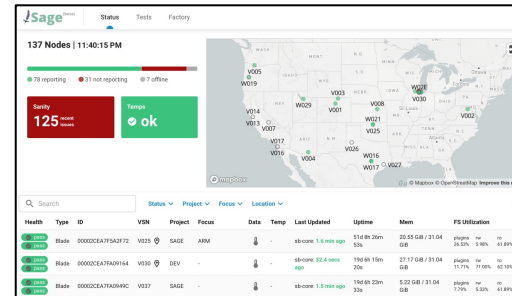


Scientific Data Analysis & Control

Artificial Intelligence
Deep Learning Inference
Lightweight Training

Slide Source: Pete Beckman, ANL

admin.sagecontinuum.org



Computation



Cloud



Data Center



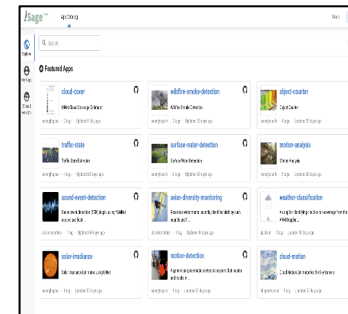
HPC

Digital Twin
Data Analysis
Machine Learning

Advanced Wireless & Networking



New inference (model)
Adaptive steering

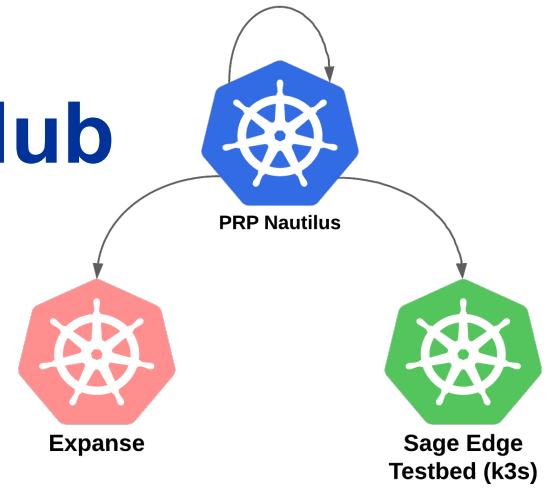
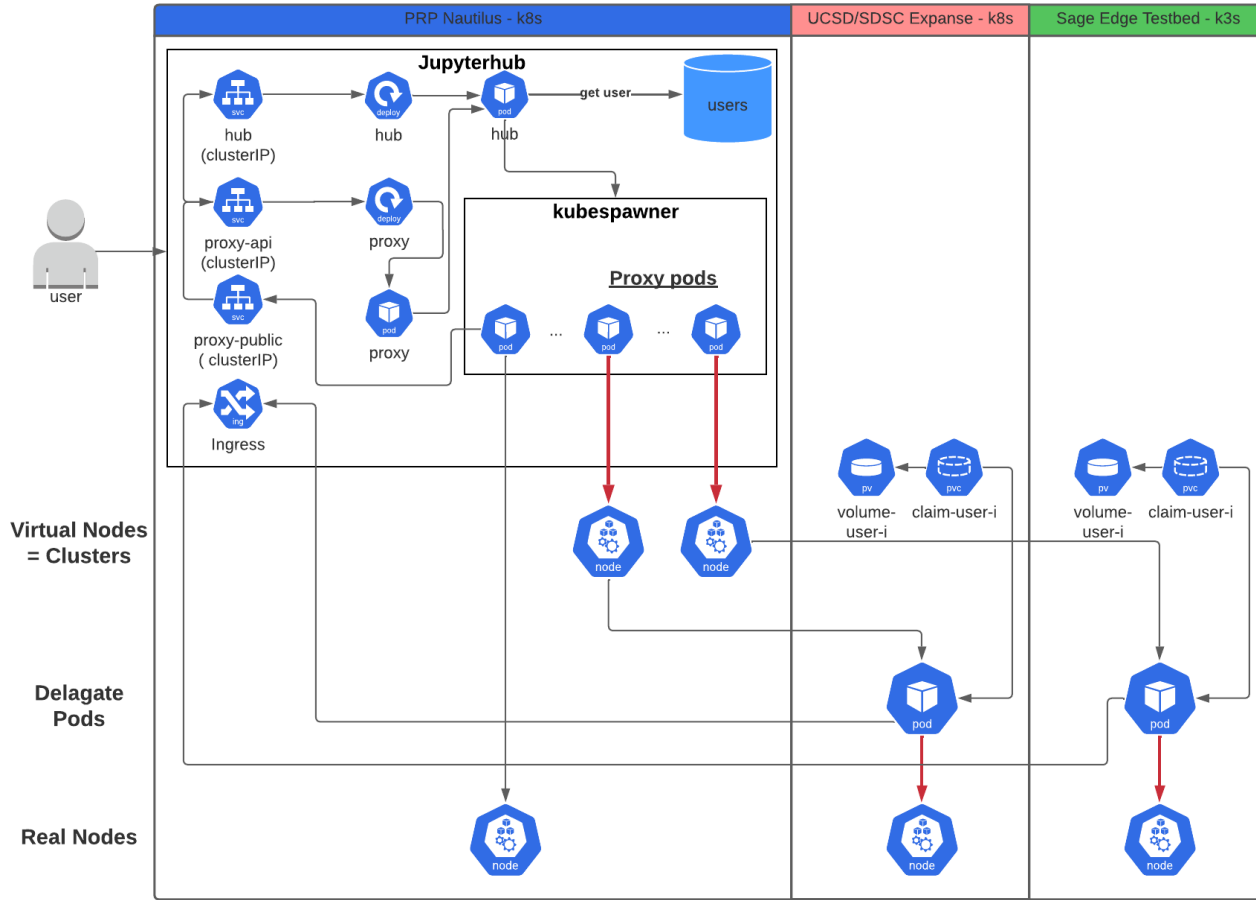


portal.sagecontinuum.org

Using Expanse, NRP and Sage for Edge AI Development

Integration of NSF EXPANSE, PRP and Sage

A Composable System Deployment of JupyterHub



- Edge-Cloud Unified Environment for prototyping AI models to deploy on the Edge
- A user can easily be provided the right environment for developing their AI Edge Application

Spawner Options

/home/jovyan is persistent volume, 5GB by default. Make sure you don't fill it up - jupyter won't start next time. You can request increasing the size in [Matrix](#)

GPUs

Cores

RAM, GB

GPU type

/dev/shm for pytorch

Mount CephFS (if assigned)

You can request assignment in [Matrix](#)

Stack options are described in [docker-stacks](#)

Image

- Stack Minimal
- Stack Minimal + Desktop GUI
- Stack Scipy
- Stack R
- Stack Tensorflow
- Stack Tensorflow + PRP added libs
- Stack Datascience
- Stack Pyspark
- Stack All Spark
- Tensorflow 1.14 (deprecated, choose one above)

Spawn



Kubernetes
Pod
Spawned
for
Exploration

Name	Last Modified
bin	4 months ago
dask-worker-space	5 months ago
data	5 months ago
include	4 months ago
kubernetes	5 months ago
lib	4 months ago
pgsql	4 months ago
postgresql-11.0	4 months ago
rclone-v1.53.1-linux-a...	5 months ago
share	4 months ago
Tempredict-Shared-P...	4 months ago
usr	a year ago
1	4 months ago
dbConnString.ipynb	5 months ago
dbConnstring.py	5 months ago
DeveloperNB-Timesca...	5 months ago
GitDemo-Tempredict-...	a month ago
KCLT.csv	5 months ago
mydask.png	5 months ago
Ops-TimescaleDB-Ta...	17 hours ago
ordered-clustering-da...	5 months ago
postgresql-11.0.tar.gz	2 years ago
PPTDemo-TimescaleD...	2 days ago
rclone-current-linux-a...	6 months ago
tempredict-oura-500-...	a month ago
Tempredict-timescale...	seconds ago
TimescaleDB-Dask-C...	5 months ago
TimescaleDB-Dask-C...	5 months ago
TimescaleDB-Dask-ps...	5 months ago
Untitled.ipynb	4 months ago
Untitled1.ipynb	16 days ago

Import Libraries

```
[6]: import pandas as pd
import numpy as np
import dask
import distributed
from sqlalchemy import create_engine
from dbConnstring import *
import dask.array as da
import os
import time
```

Define Database Connection Details

```
[6]: # TimescaleDB username, password, and database name
TimescaleDB_USERNAME = '' ## YOUR TimescaleDB_USERNAME = 'abc'
TimescaleDB_PASSWORD = '' ## YOUR TimescaleDB_PASSWORD = 'xyz'
# Create the connection
postgres_str = conn_str(TimescaleDB_USERNAME, TimescaleDB_PASSWORD)
cnx = create_engine(postgres_str)
```

Create a DASK Cluster

```
[12]: from dask import dataframe as dask_cluster_dd
```

```
[13]: N_WORKERS = 8
```

```
[14]: from dask.distributed import Client, LocalCluster

if __name__ == '__main__':
    # Create a Dask Cluster
    cluster = LocalCluster(n_workers=N_WORKERS, threads_per_worker=1, processes=True)
    client = Client(cluster)
```

Client	Cluster
Scheduler: tcp://127.0.0.1:40939	Workers: 8
Dashboard: http://127.0.0.1:8787/status	Cores: 8
	Memory: 135.06 GB

Create a 'delayed' function with DASK Cluster

```
[16]: df = dask_cluster_dd.read_sql_table('hrv_500',
postgres_str,
npartitions=8,
index_col='sensortime')

df
```

Data federation is still a big deal!

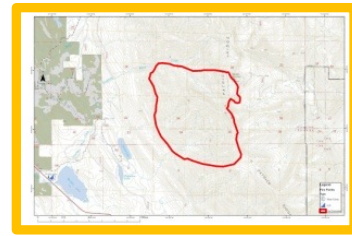
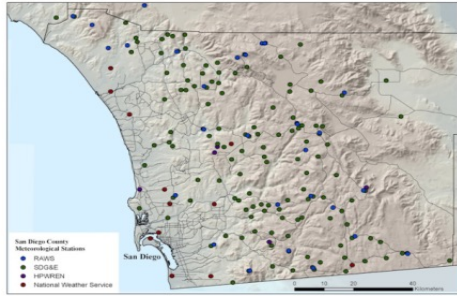
**Needs to be coupled with data systems,
knowledge networks, and commons frameworks.**

Application of Composable Systems in Fire Science and Management

**Fire is an
inevitable part of
nature, but
megafires are not.**



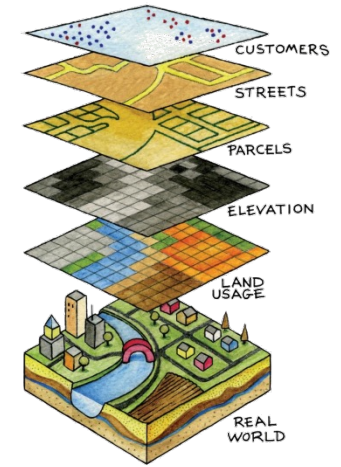
Next generation fire prediction combines emerging fire science with data from many sources.



Fire perimeter

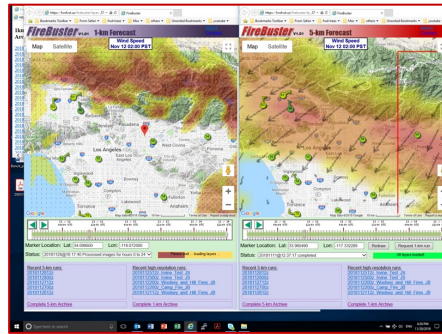


Remote sensing



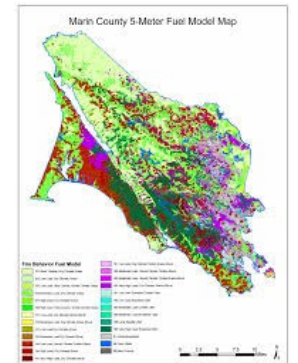
Landscape data

Ground-based real-time weather and camera imagery



Weather forecast and modeling

Data-Driven Fire Preparedness, Mitigation and Response



Land cover and fuel

WIFIRE Lab at-a-Glance

wifire.ucsd.edu

FIREMAP

Initial Attack Fire Response



Edge AI services and data plugins



Prescribed Fire Optimization



Powered By

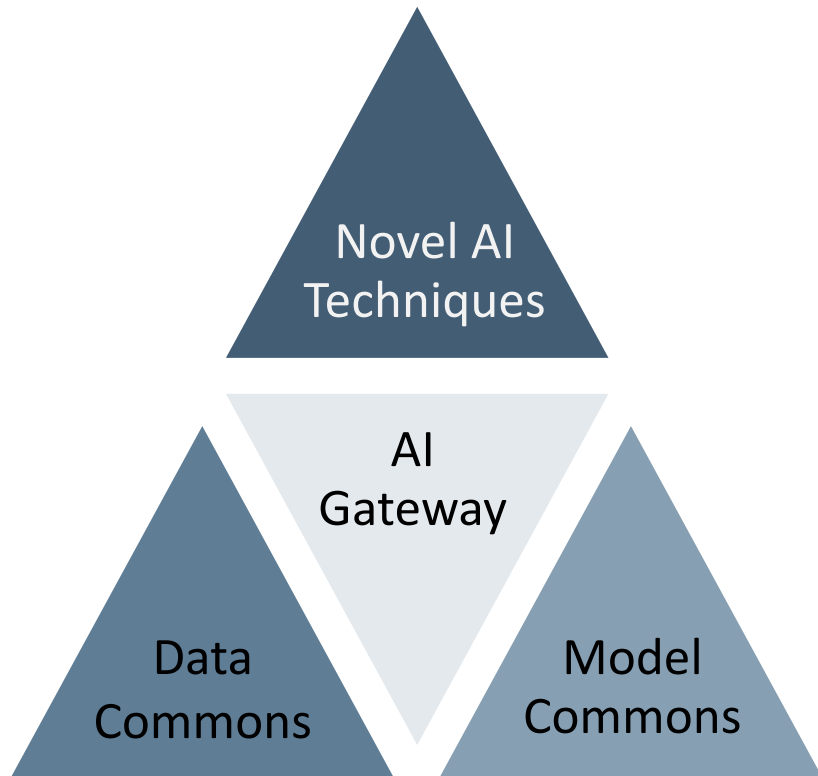


FAIR Data and Model Catalog and AI Gateway

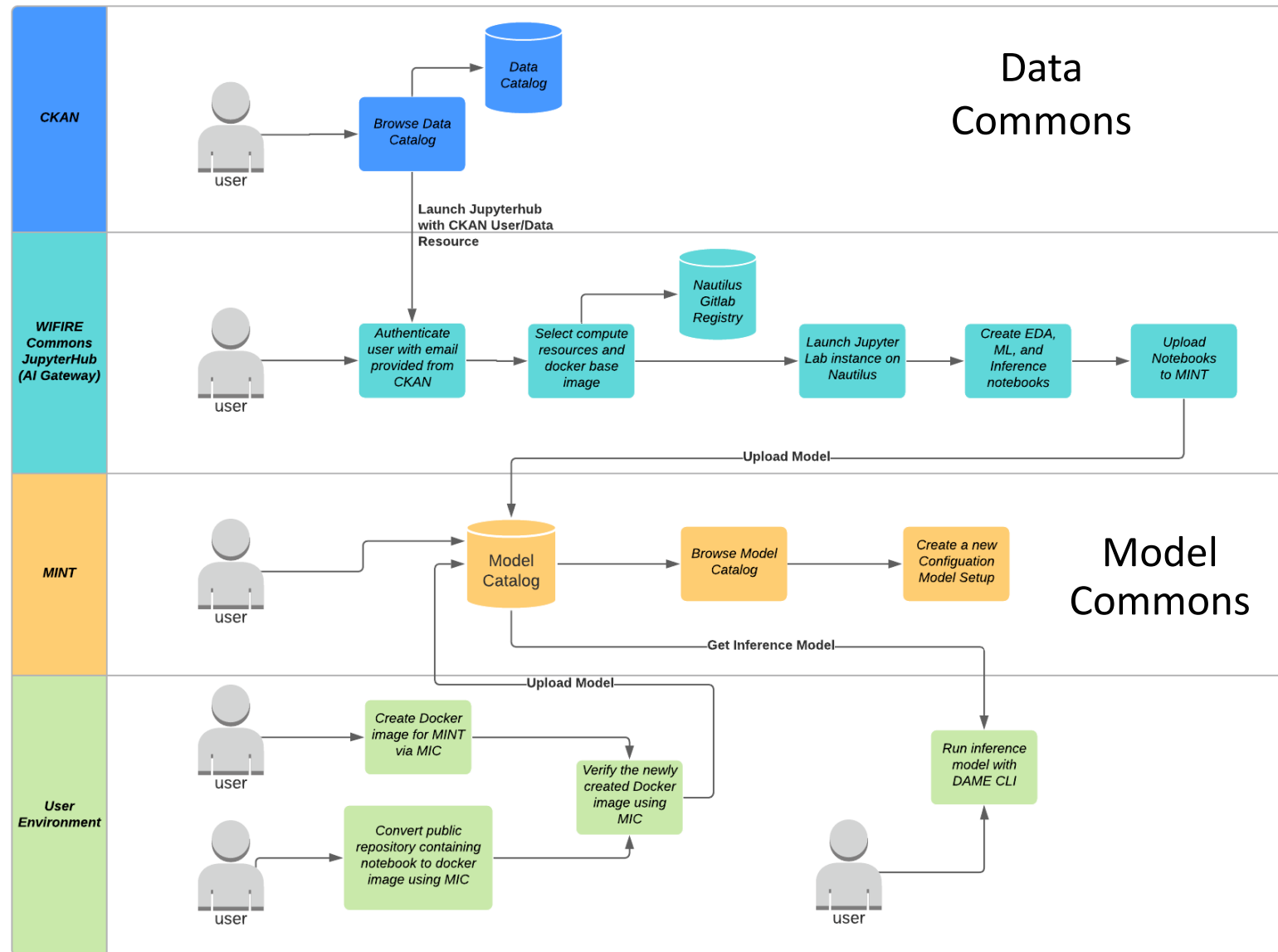
Other funders:

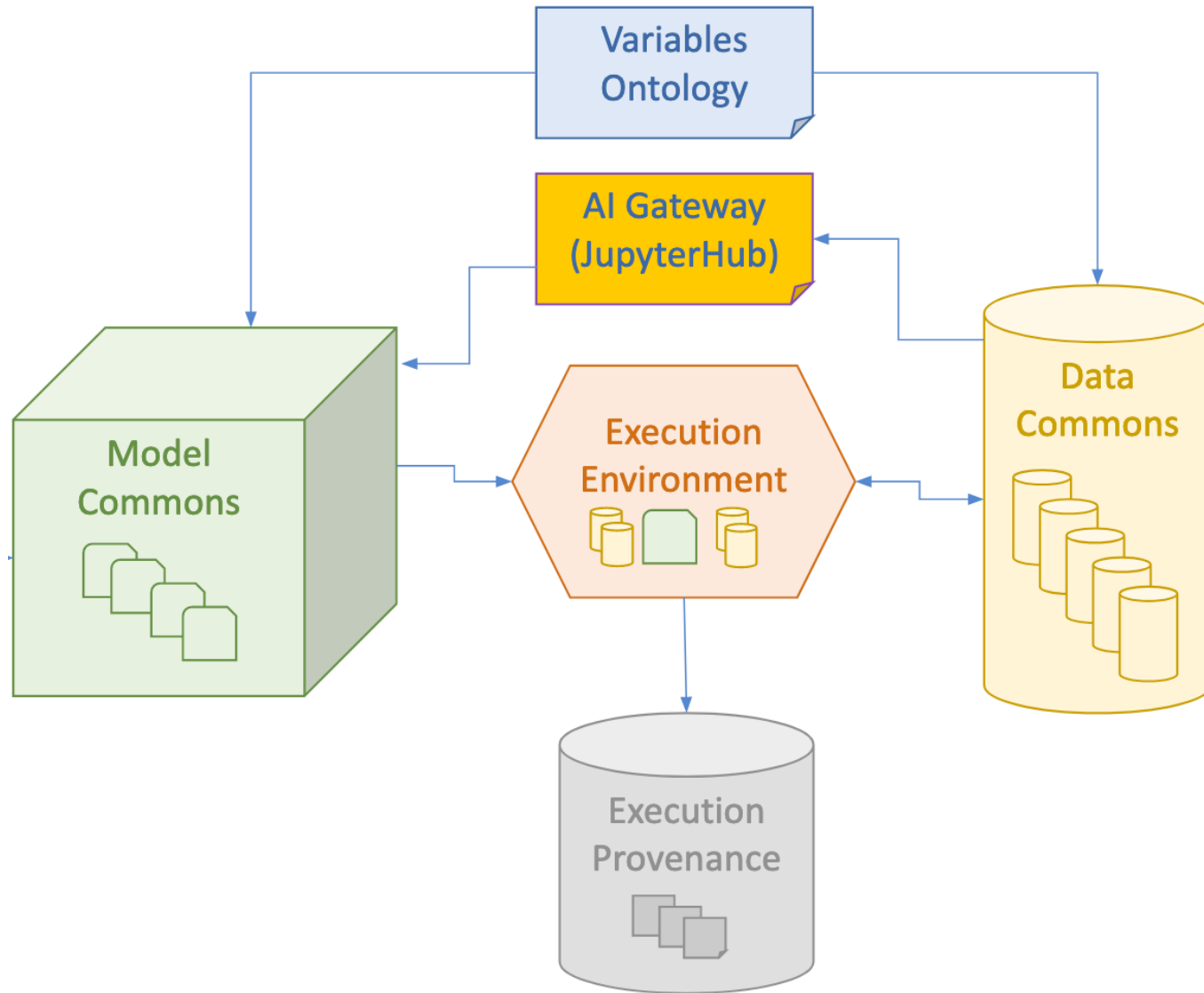


Teamwork and AI



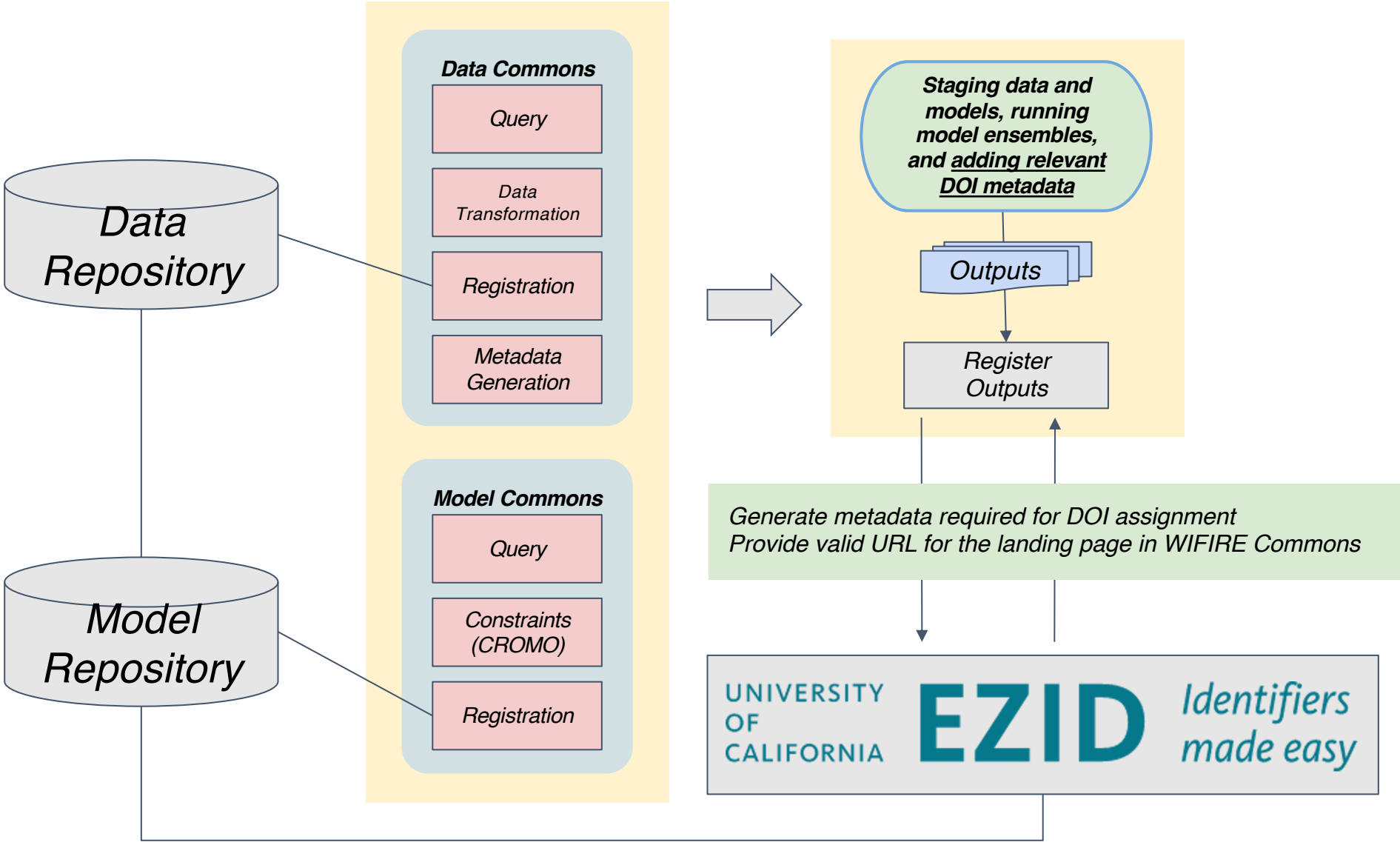
<https://wifire.ucsd.edu/commons>



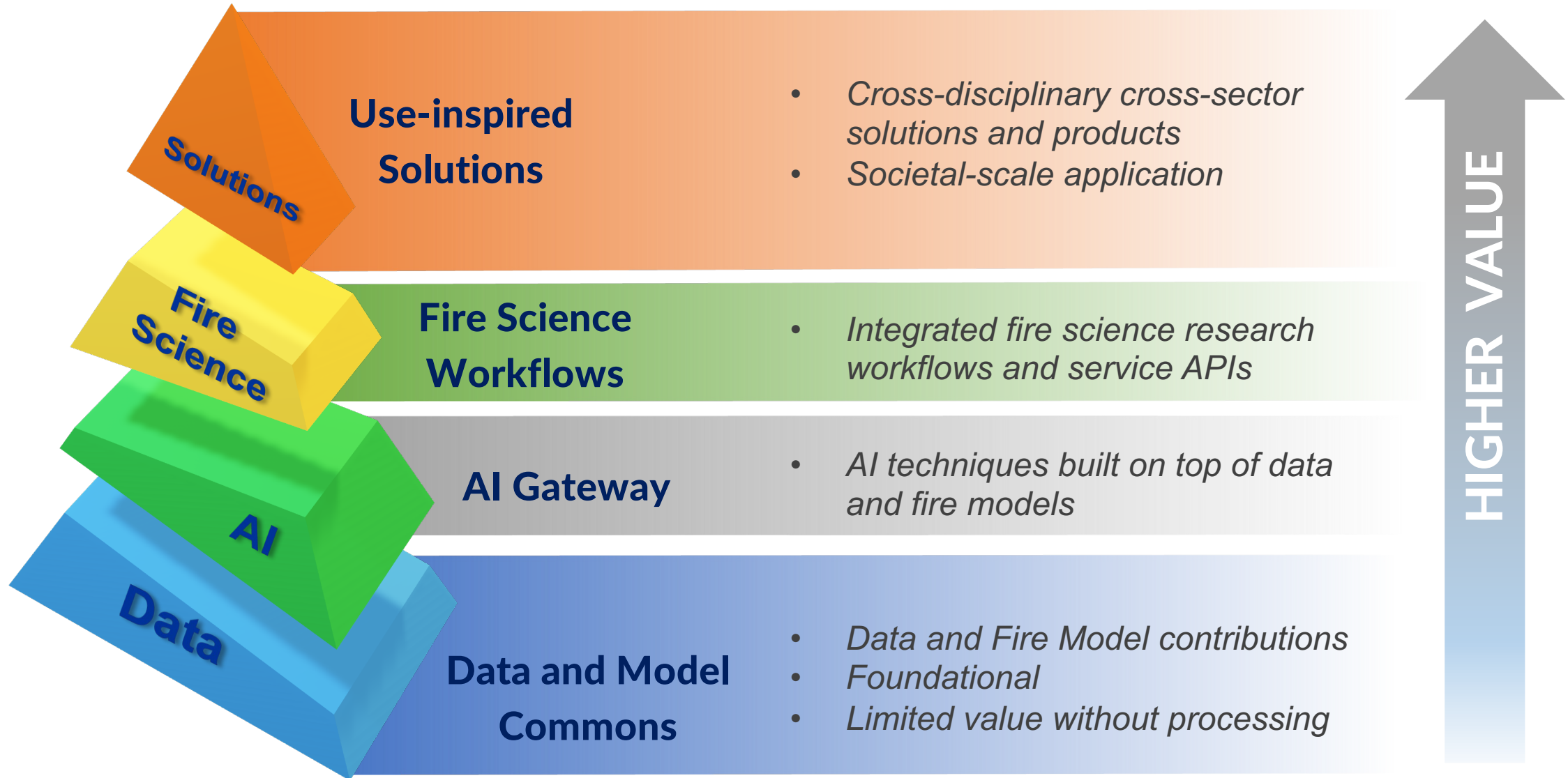


Linking Data and Models to Open Knowledge Networks

DOIs for Models and Datasets



Data to Value in WIFIRE Commons

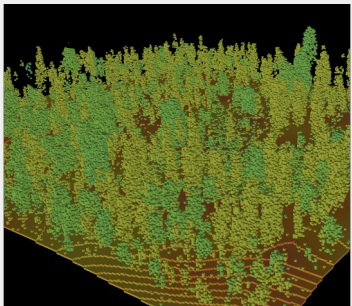


AI Gateway → Pangeo Customized for AI and Fire Science

<https://wifire-commons-pangeo-jupyter.nrp-nautilus.io/>

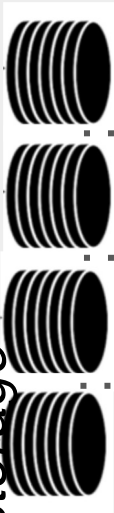
AI-Ready Ensemble Datasets stored in cloud-optimized Zarr format

FastFuel



distributed

storage



Tech stack supporting scalable analysis of large wildfire datasets

Nautilus PRP



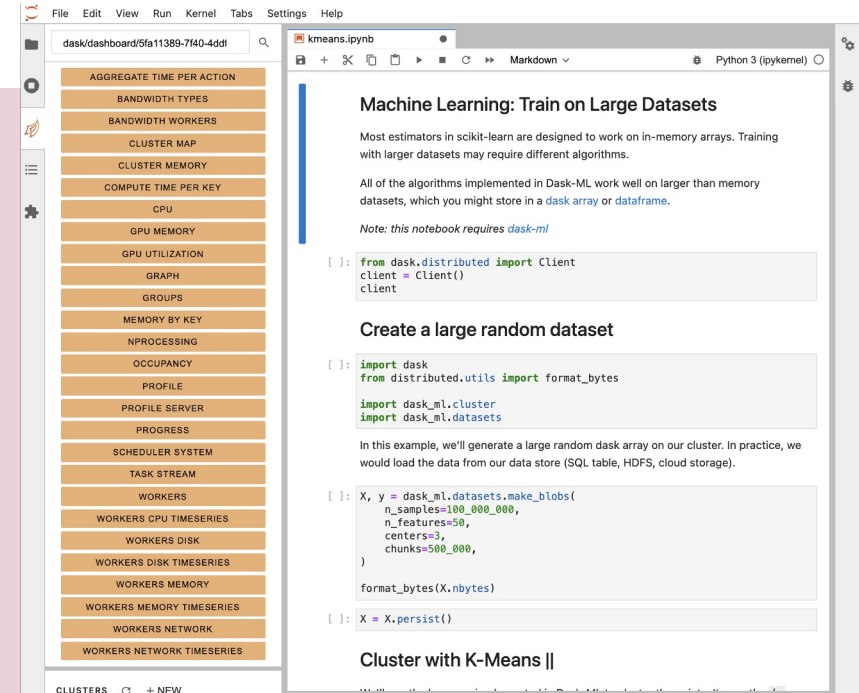
- Label-based indexing and arithmetic
- Wide range of input/output options: netCDF, HDF, geoTIFF, Zarr



- Data cataloging utility



- Parallel processing



JupyterHub facilitates interactive data analysis on remote systems



web browser



end user



Focus on Two Ways to Manage Wildland Fires



REACTIVE

Fighting Severe Wildfires

to

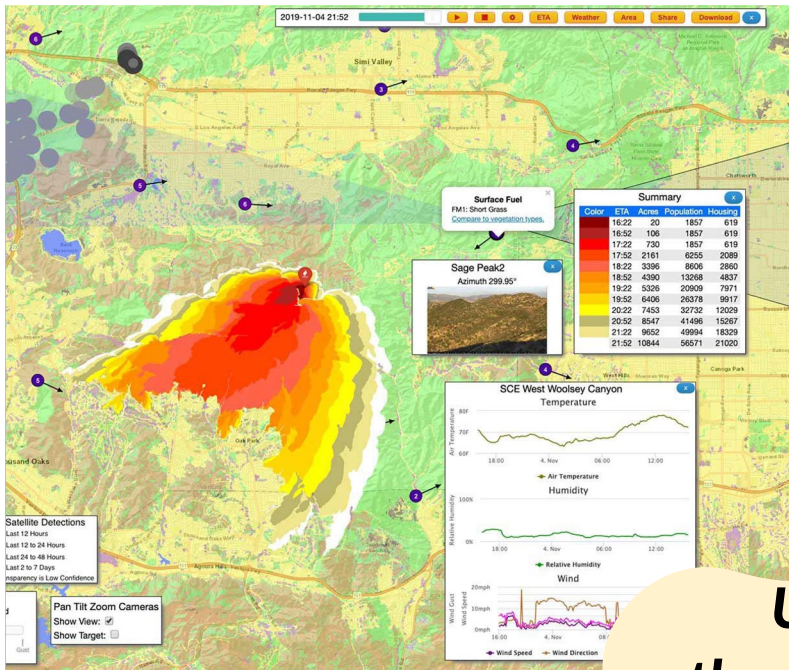


PROACTIVE

**Conducting Prescribed
Burns**

AI-Integrated Real-time Fire Response In WIFIRE

WIFIRE Firemap

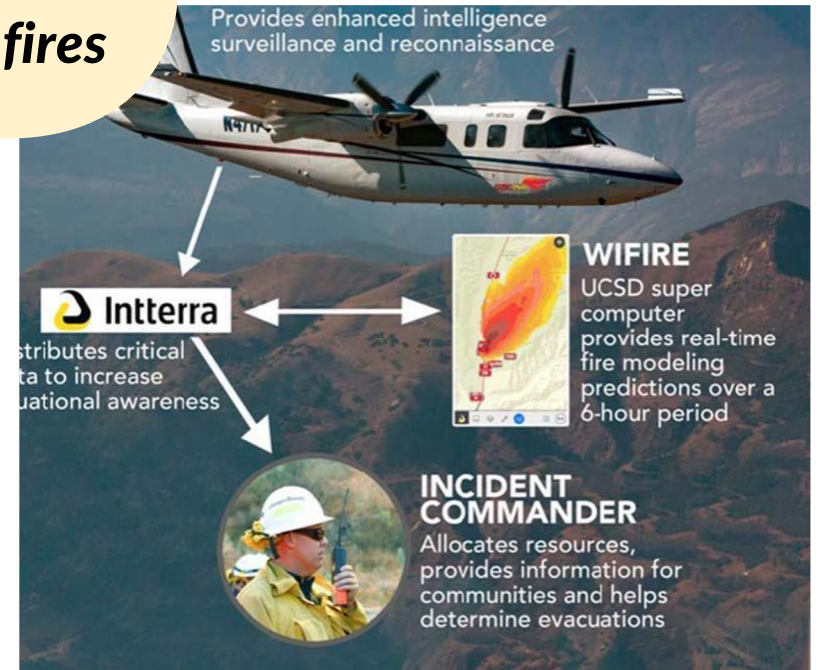


Used by thousands of firefighters

Operational in 100+ fires



FIRIS



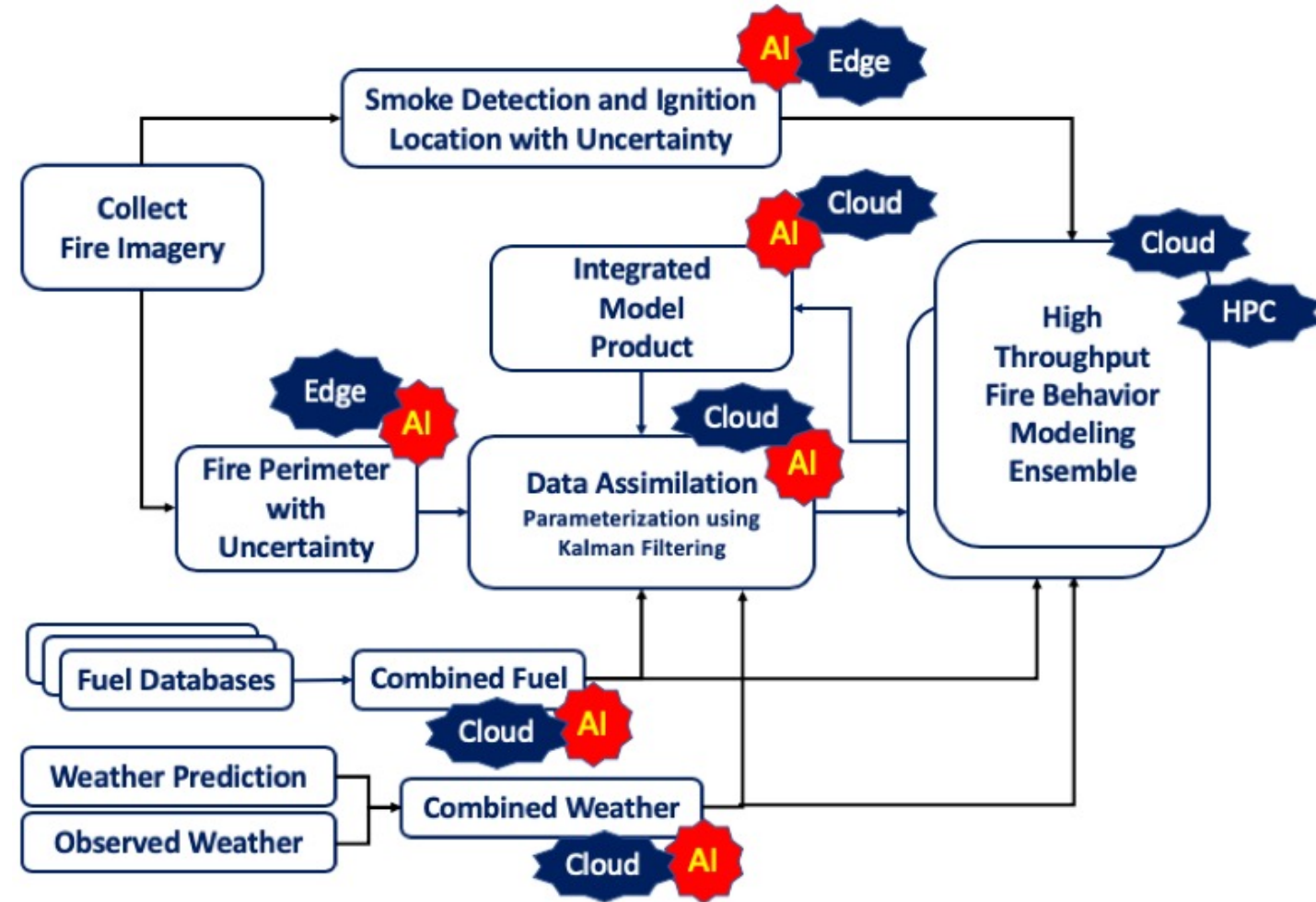
Dynamic Data-Driven Fire Modeling

Data and AI Needs

- Characterizing the dynamic fire environment : Variation of wind, smoke, moisture, fuels, fire perimeter, ...
- Detection of fire ignitions
- Decision support for fire management
- Prediction of potential fire ignitions

Heterogenous Computing Needs

- Edge
- AI processors
- Cloud
- HPC



Deep Learning-Based Smoke Detection

CITE: <https://www.mdpi.com/2072-4292/14/4/1007>



Data: Camera images of wildland fires

- **System Architecture:** Several deep learning models to extract spatio-temporal information from camera imagery to detect smoke plumes.

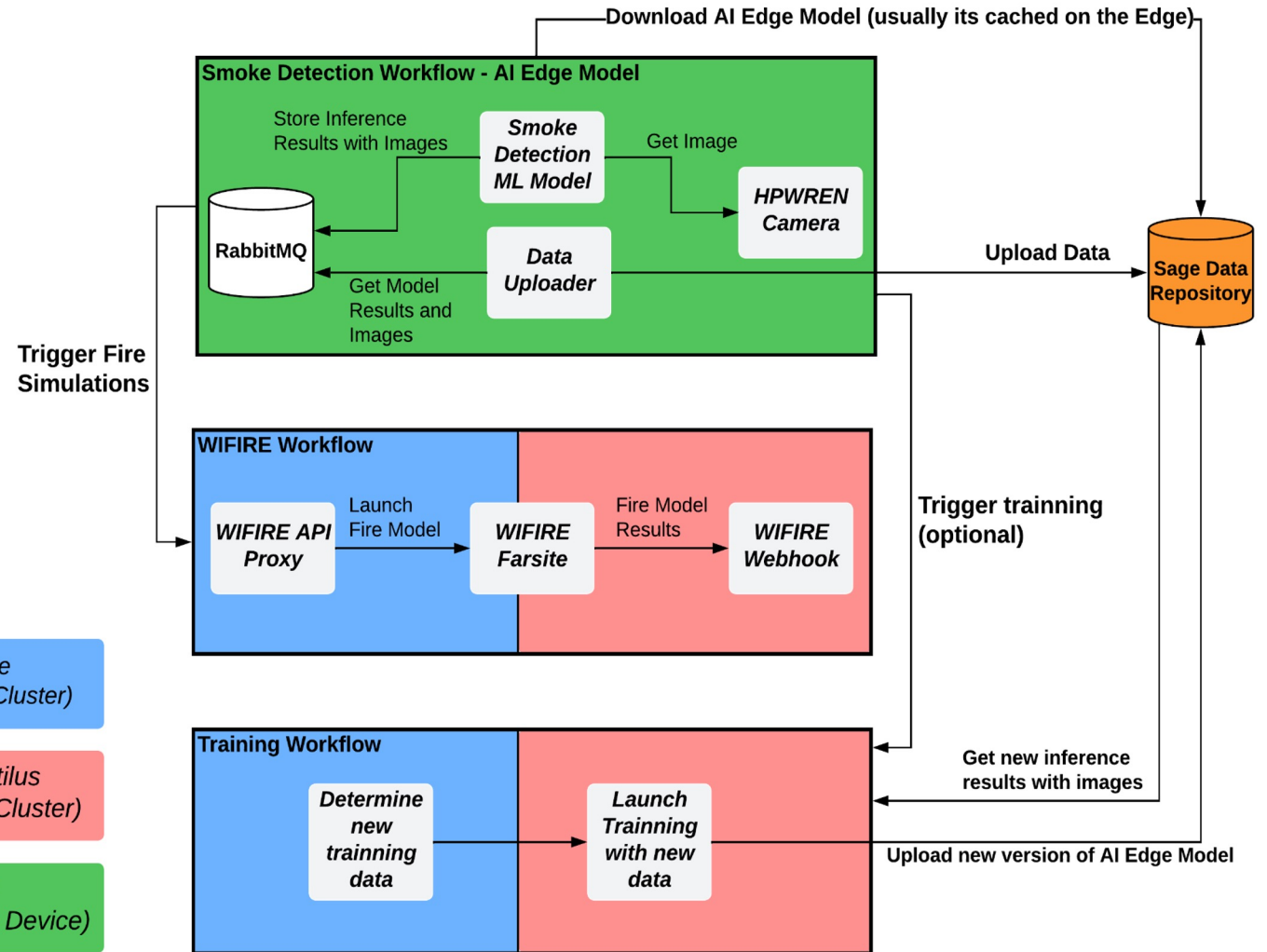
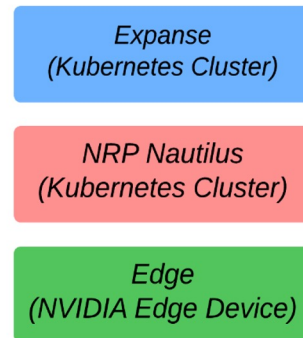
Training Data: Labeled camera images

Computing: Edge and composable systems



Fire Simulations using Composable Systems and Edge Smoke Detection

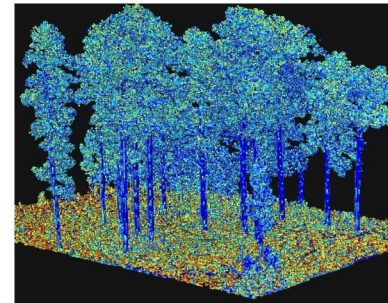
- Three workflows
 - Smoke – Sage Edge App
 - Fire simulator
 - AI Training
- Both the fire simulator and training workflows are can be run on Expanse or Nautilus through the federation layer



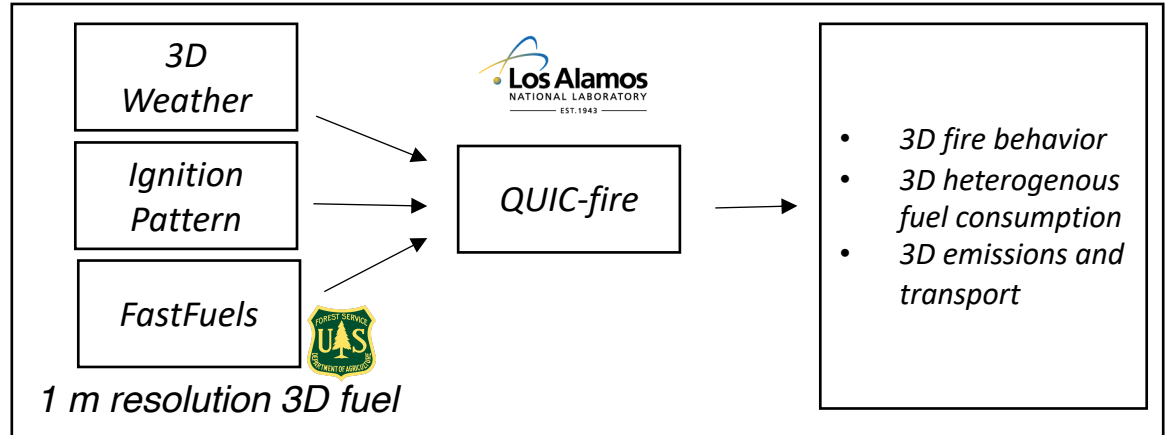
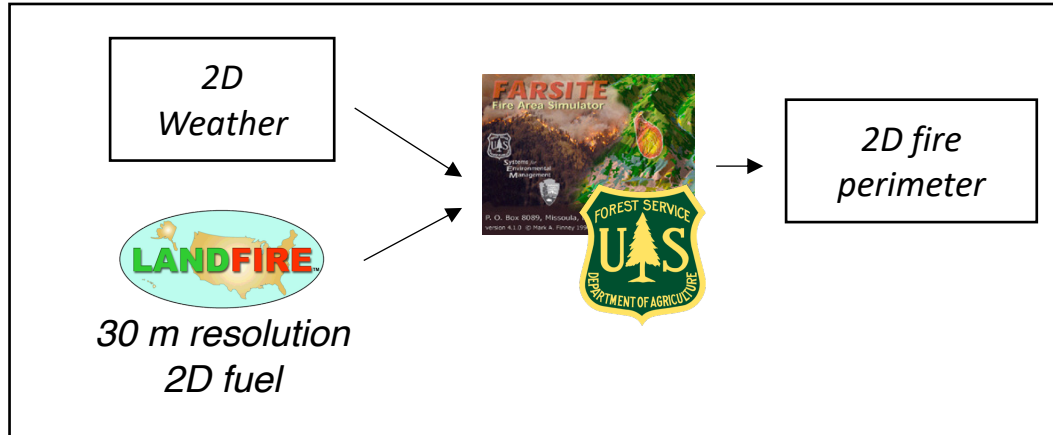
Prescribed Burns

Transformational Science

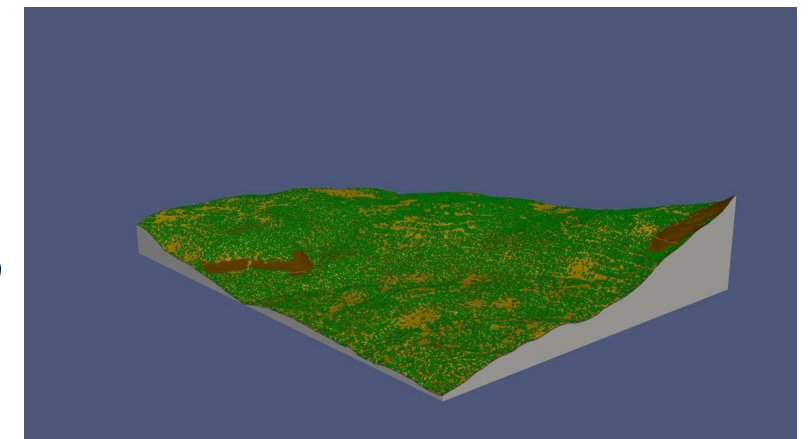
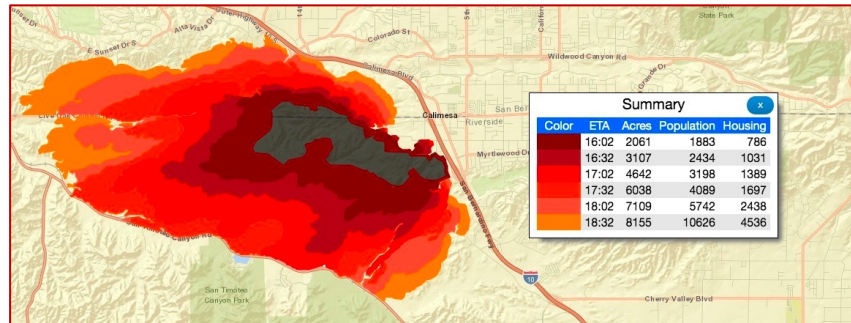
Fast 3D is better than 2D!



Next-generation fire models provide the right science basis for proactive mitigation and management decisions.



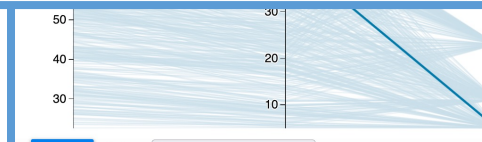
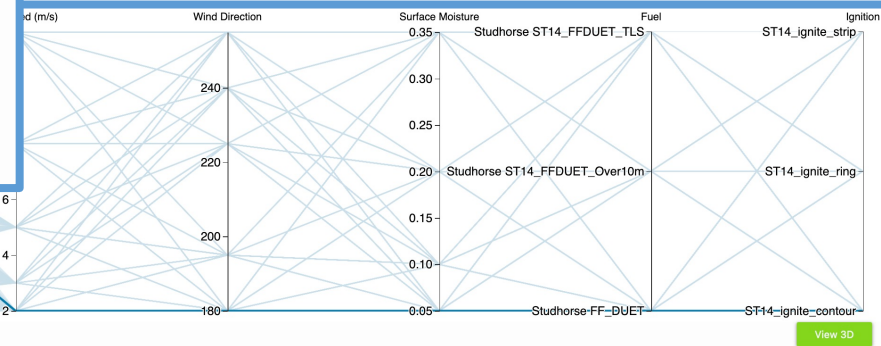
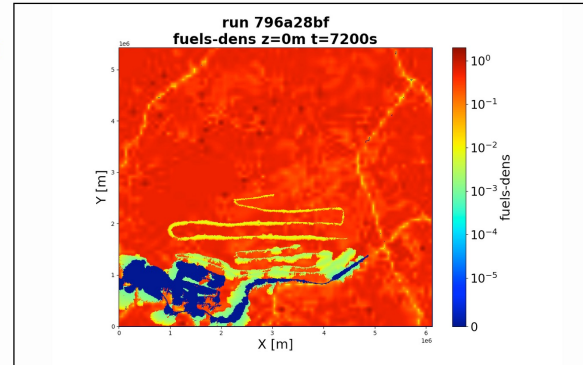
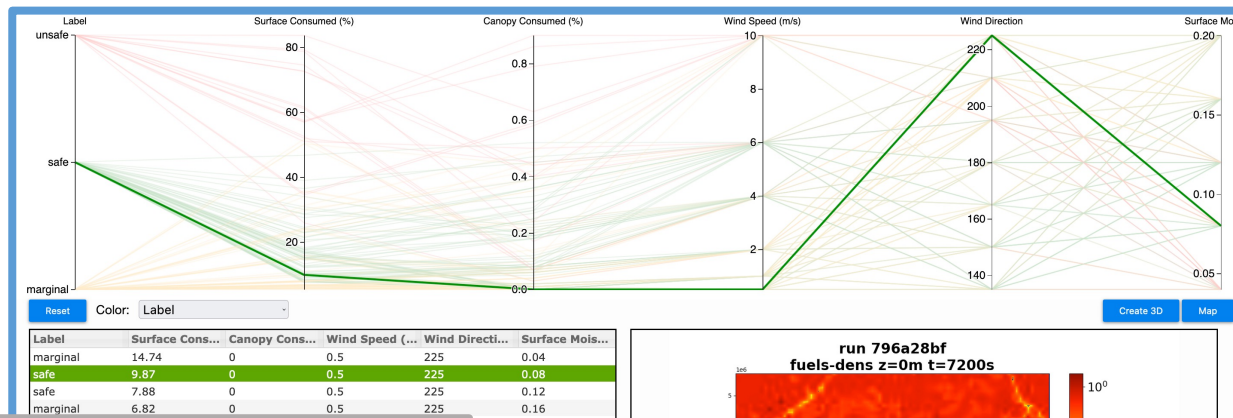
Collaboration with Rod Linn (LANL), Kevin Hiers (TTRS) and Russ Parsons (USFS)



From 30m resolution in 2D ... in 3D

To 1m resolution

Ensemble Run Analysis



Ensembles

- Idealized Grass
- Piedmont-Hitchiti
- Yosemite Studhorse BU 1-4

Runs: 900

Fuel: FastFuels, 1.8 km x 1.1 km

Sim Time: 53 min

Ignitions: 3: Contour, Ring, Strip

Surface Moisture: 5%, 10%, 20%, 35%

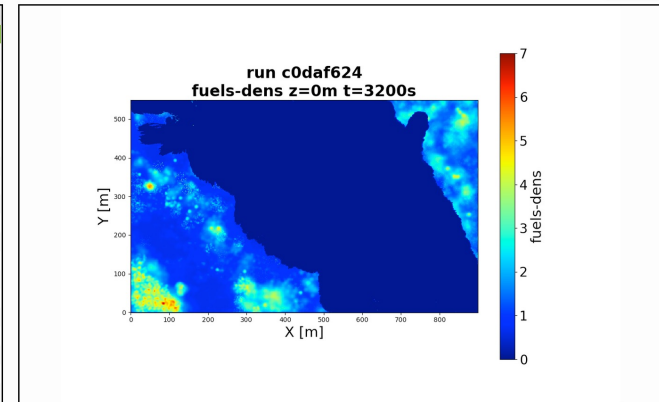
Wind Direction: 180, 195, 225, 240, 255

Wind Speed: 2, 3, 5, 8, 12

Outputs: Fuel Density, Surface Energy, Wind

ANALYZE

Yosemite Studhorse BU 5



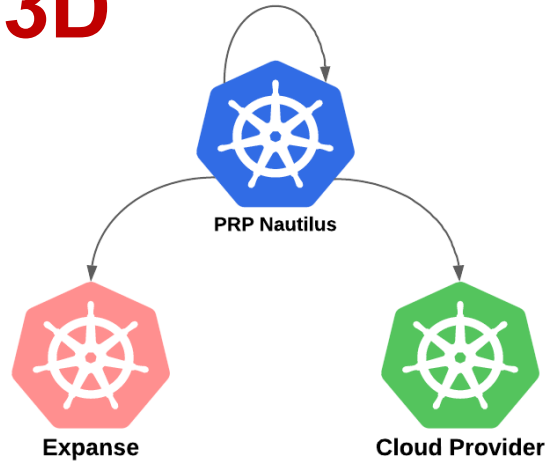
Emerging AI Needs for Next Generation Fire Modeling

- High-res, 3D and dynamic fire environment data
- Speed, flexibility, and interpretability of fire models

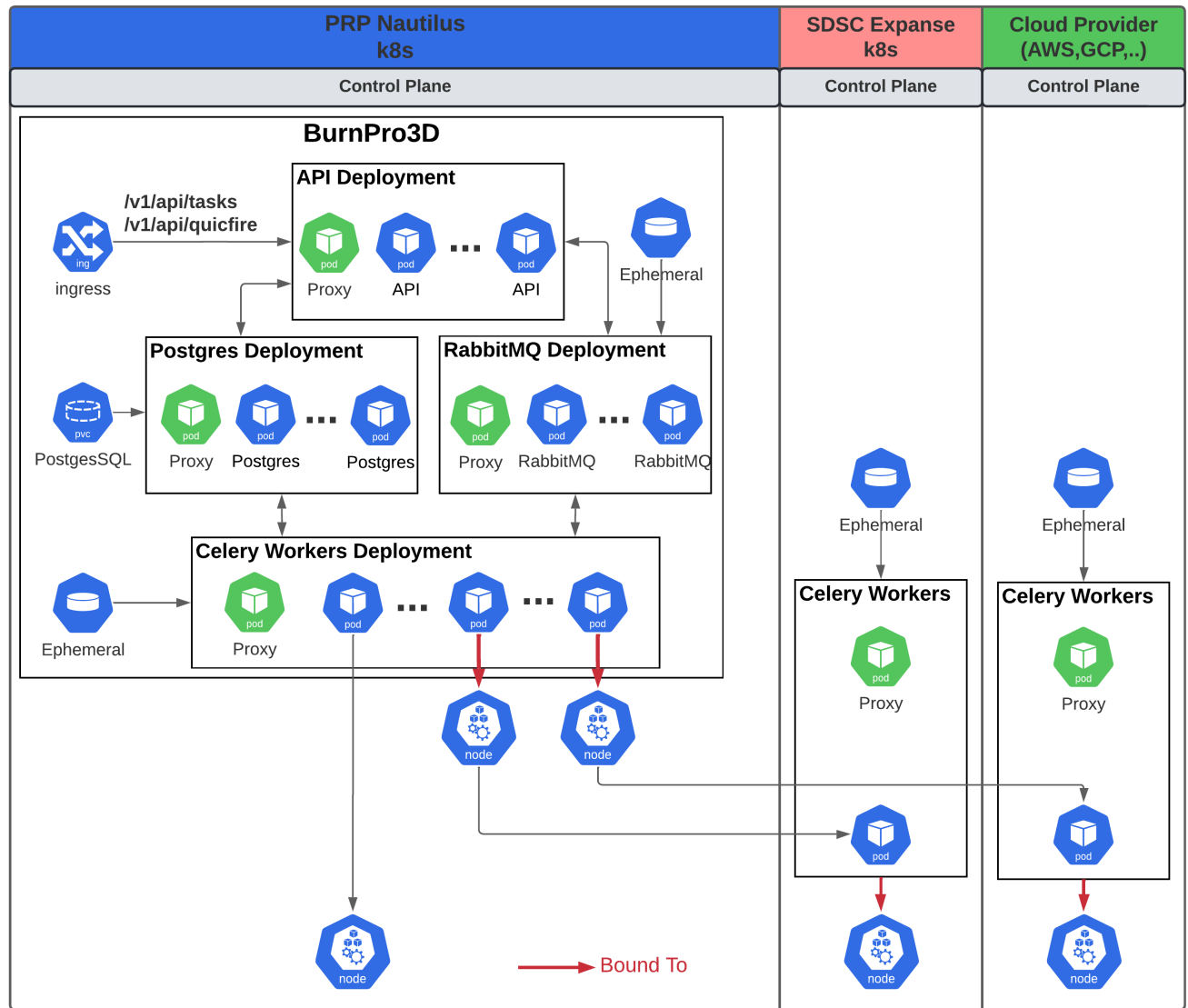


The **BurnPro^{3D}** platform gives our public sector partners **next-generation fire science** using **data** and **AI** to optimize prescribed burns at an unprecedented scale.

3D Fire Simulation Ensembles using Composable Systems in BurnPro3D



- Instead of deploying all the services on each cluster we only federate over the pods that are performing the fire simulation

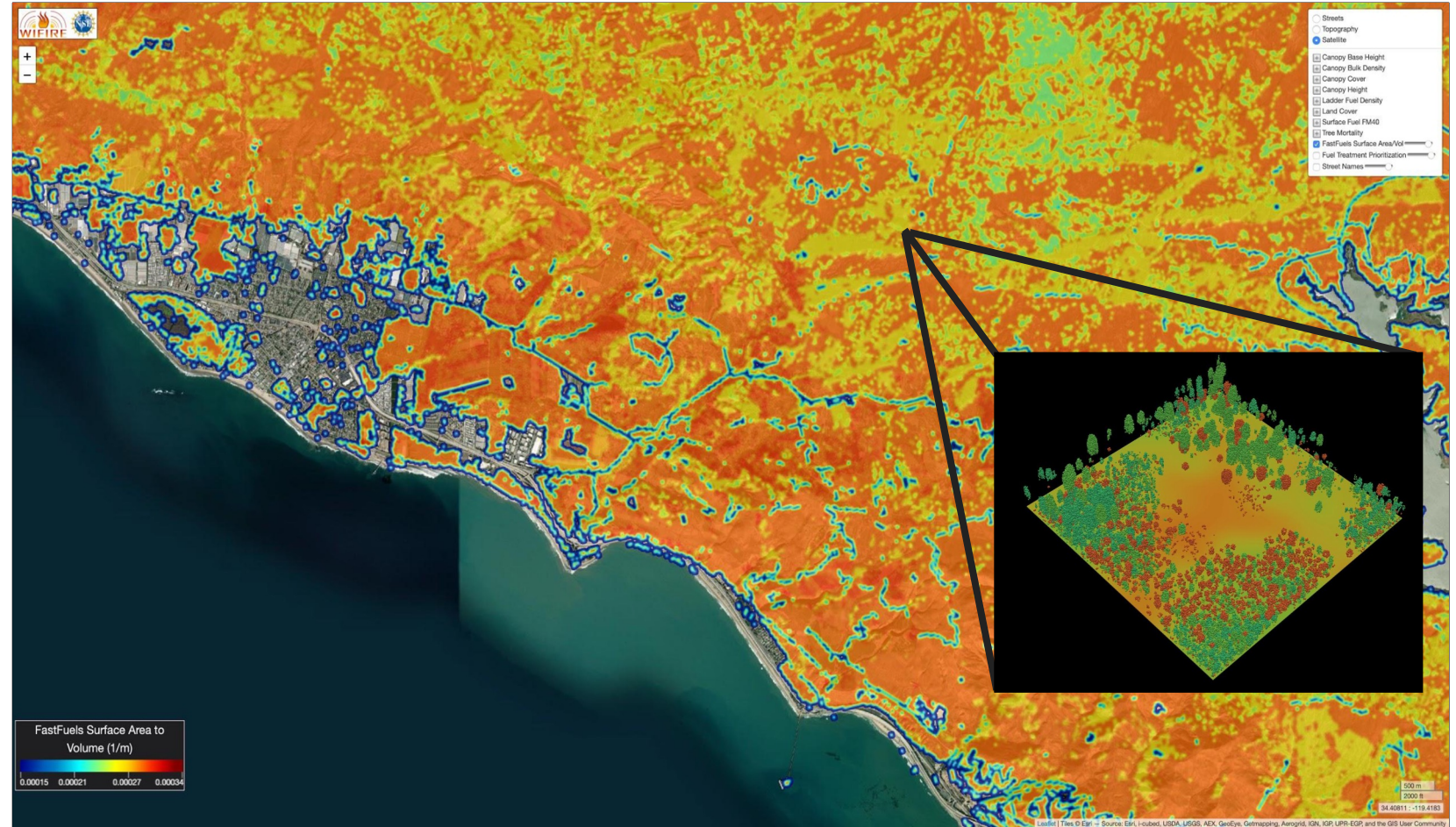


AI Techniques to Condition Data and Improve Model Accuracy

3D vs 2D

900x more detailed

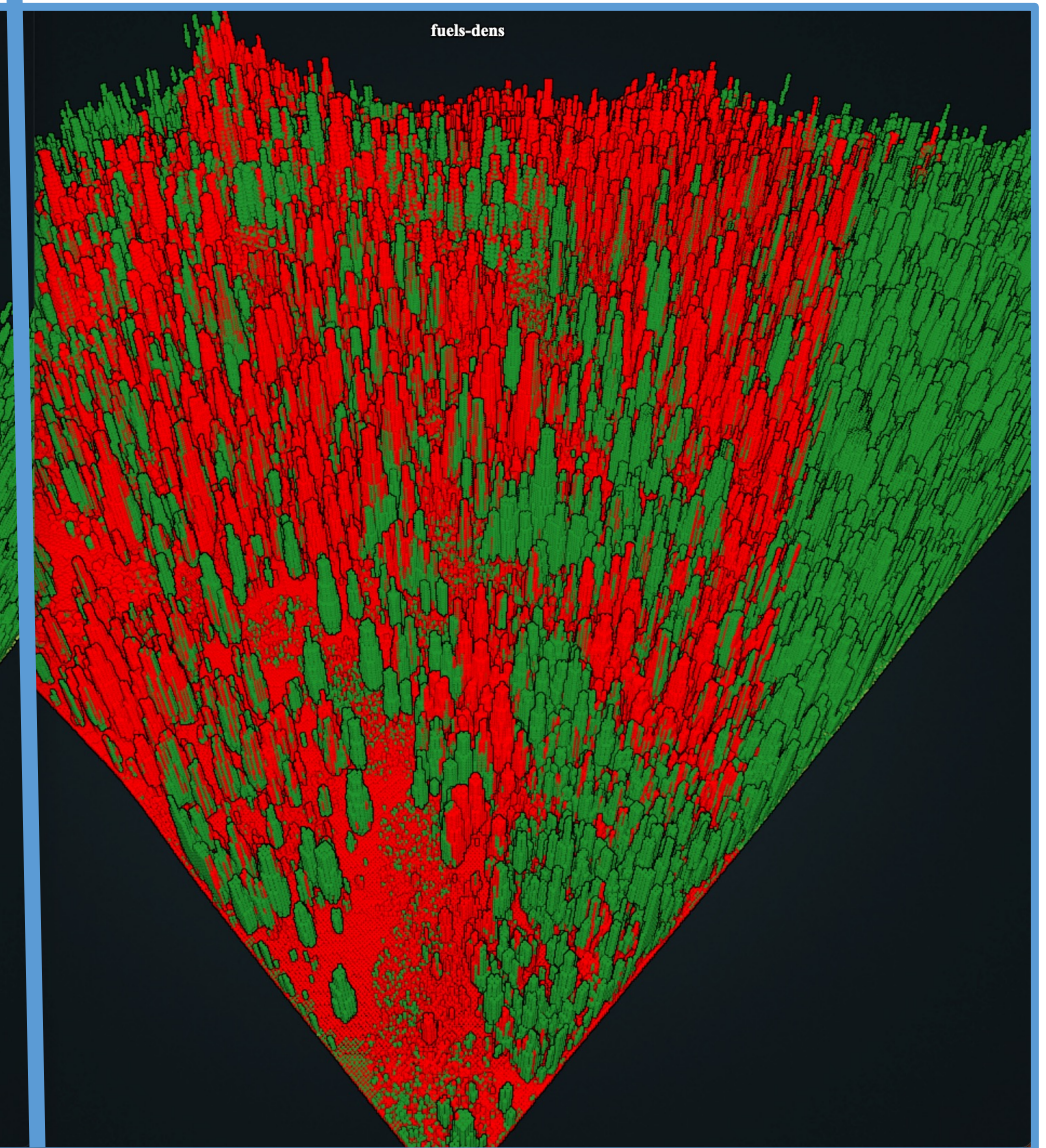
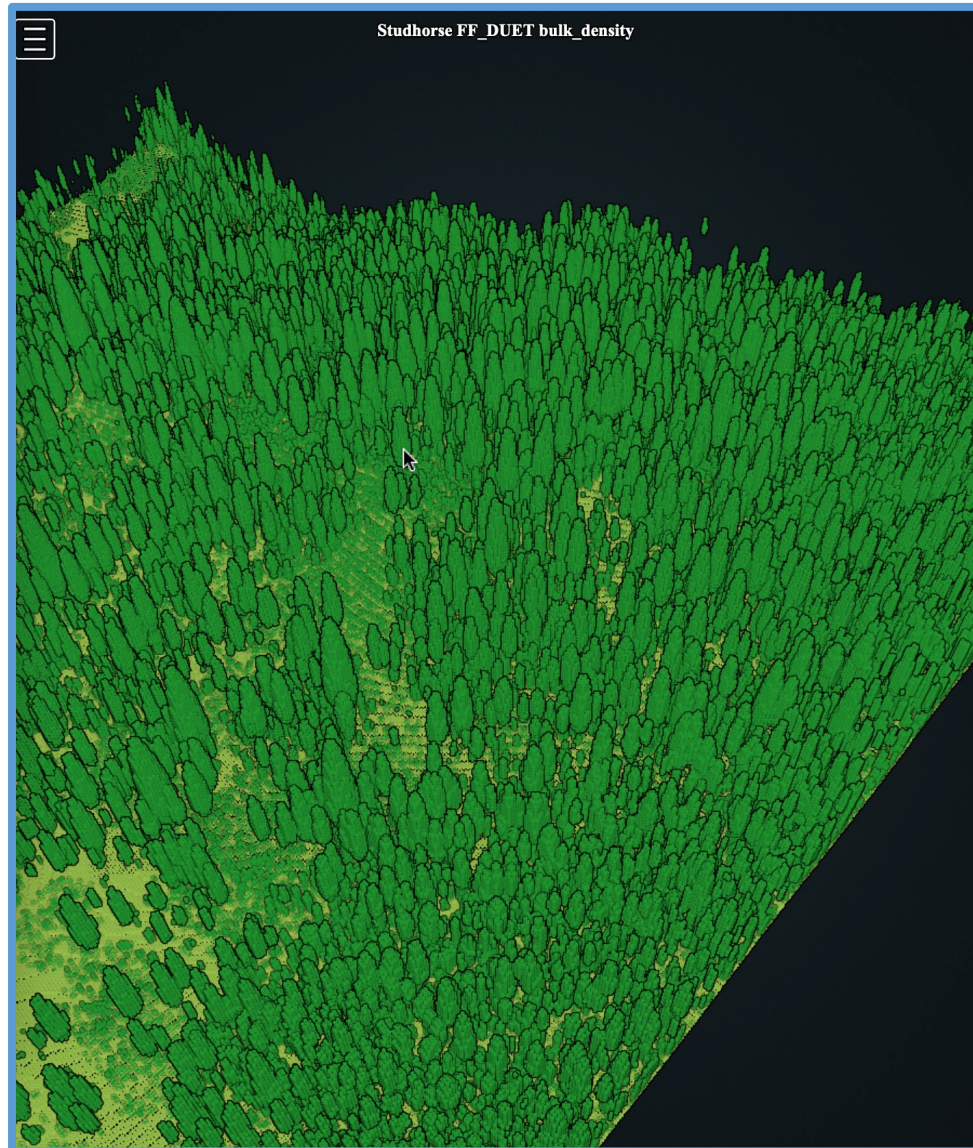
Collaboration with Rod Linn (LANL), Kevin Hiers (TTRS) and Russ Parsons (USFS)



BEFORE

AFTER

Yosemite National Park



AI Techniques to Improve Decision Making

Weather

*Ignition
Patterns*

Smoke

PHYSICS-GUIDED MACHINE LEARNING

To improve predictive fire behavior models

OPTIMIZATION

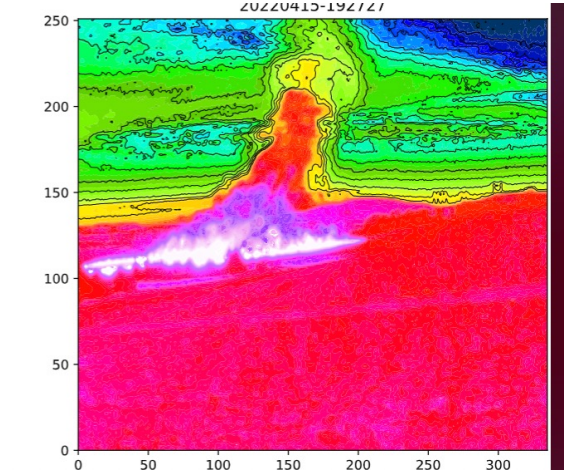
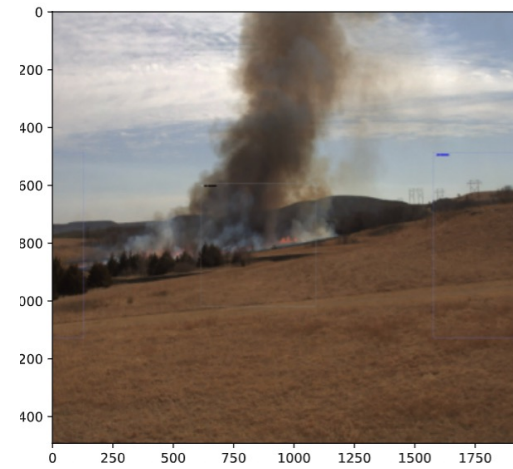
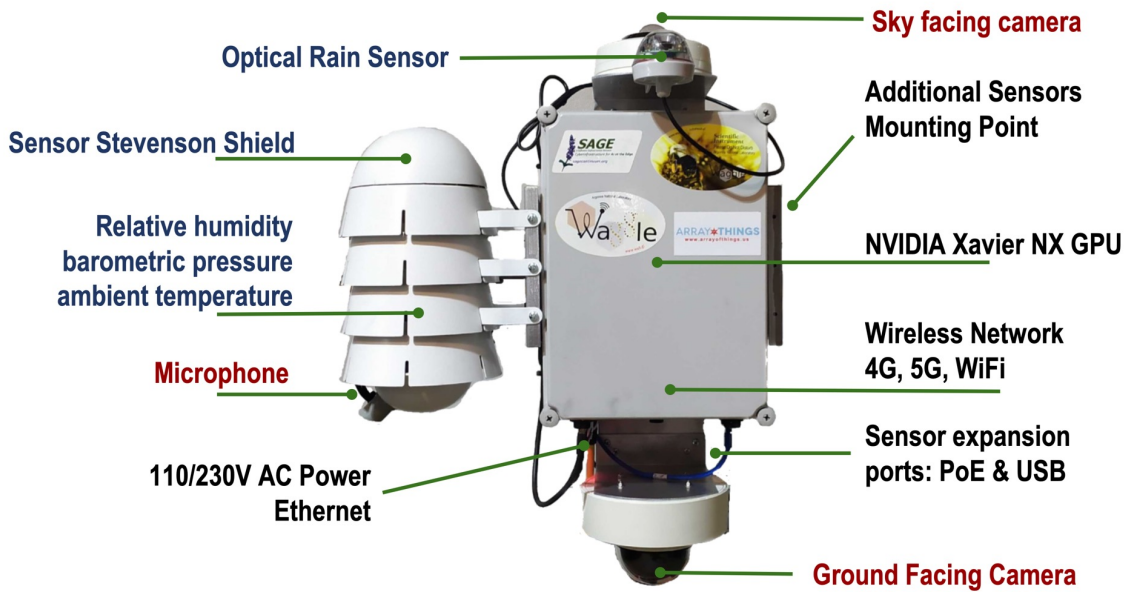
To address complex tradeoffs and prioritization

EXPLAINABLE AI

To increase scientific understanding and interpretability all along the decision-making chain

High resolution data from Sage is needed for understanding of the prescribed fire environment at the resolutions required.

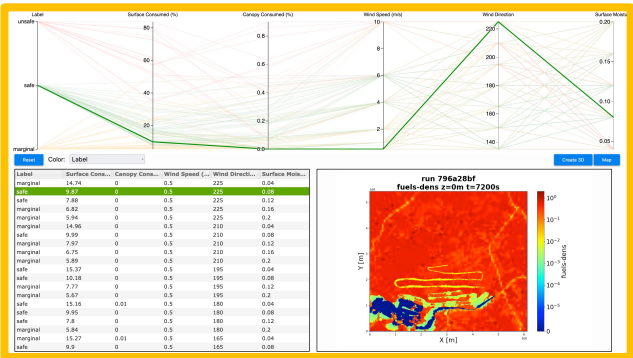
NEON Mobile Deployment Platform (MPD) with SAGE Konza Prairie for controlled burn: April 2022.



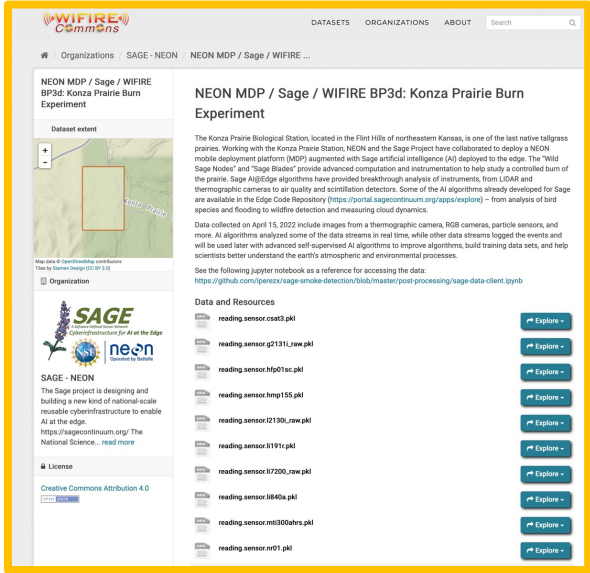
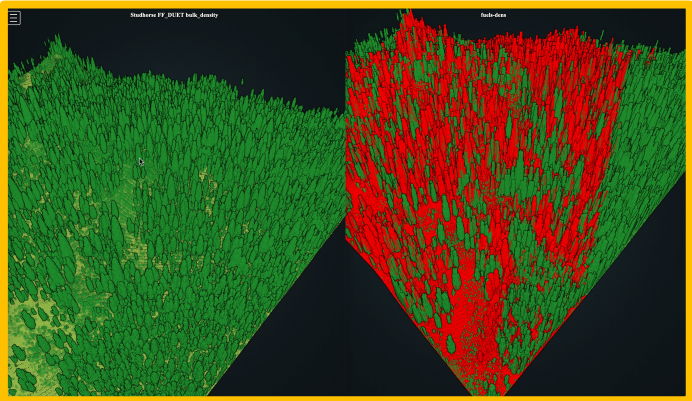
Konza Site Prescribed Fire Scenario

AFTER

BEFORE



DAY OF

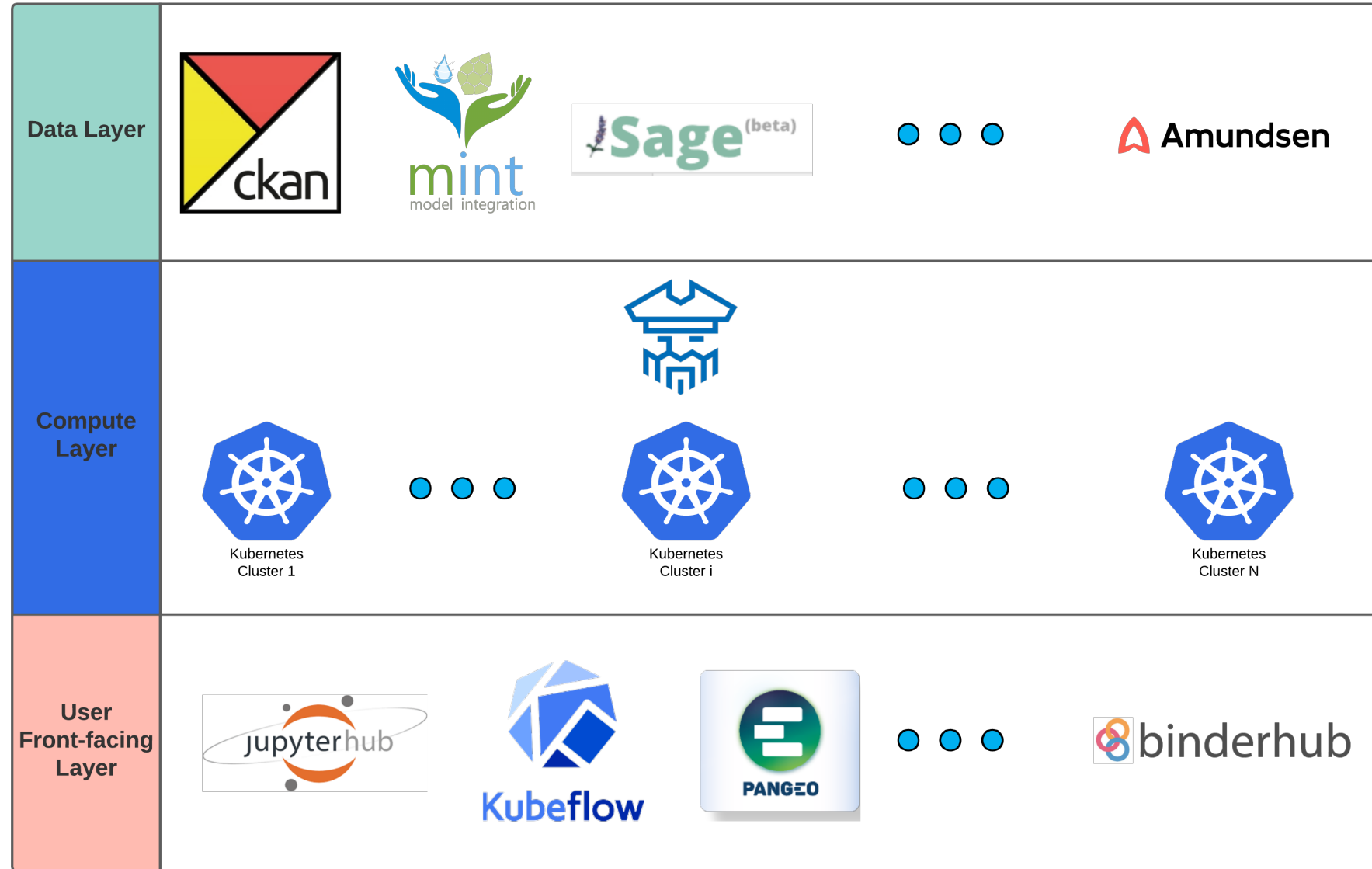


- **Run ensemble models of the burn**
- **Decide on burn prescription**
- **Make Sage NEON mobile unit request**

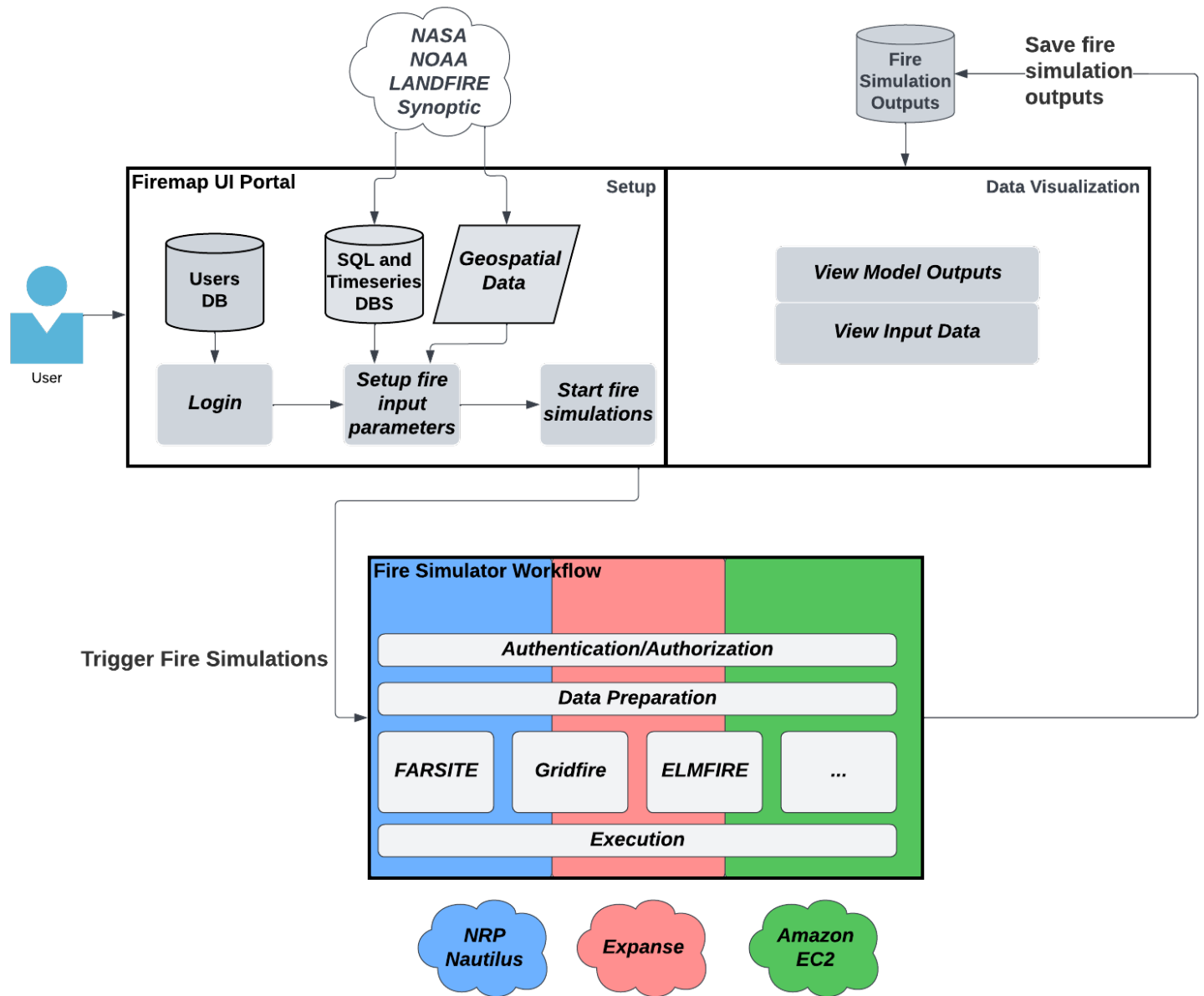
- **Make Go/No Go decision**
- **Deploy edge AI for smoke detection and environment monitoring**
- **Set up Sage dashboard**

- **Publish data in WIFIRE Commons**
- **Conduct post-fire model evaluation**
- **Create adaptive prescribed burn strategies**

Use Case: Generalized Application Development



Use-Case: Many Fire Simulators Workflow



powered by WIFIRE Commons

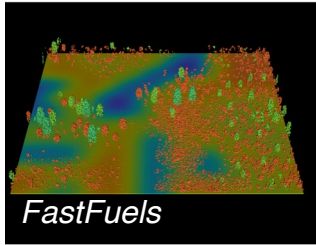
A Data and AI Framework for Convergence in Wildland Fires

Data Sharing and AI as a bridge.

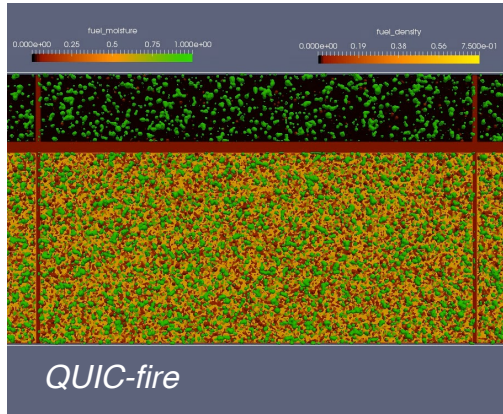
Fire Science

- Knowledge Management
- Physics-Guided ML
- Optimization
- Interpretability

Application to Preparedness, Mitigation, Response, Recovery



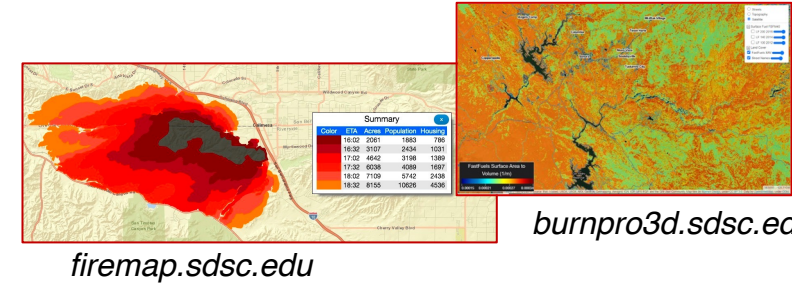
FastFuels



QUIC-fire

Data Commons
wifire-data.sdsc.edu

Model Commons
wifire.models.mint.isi.edu

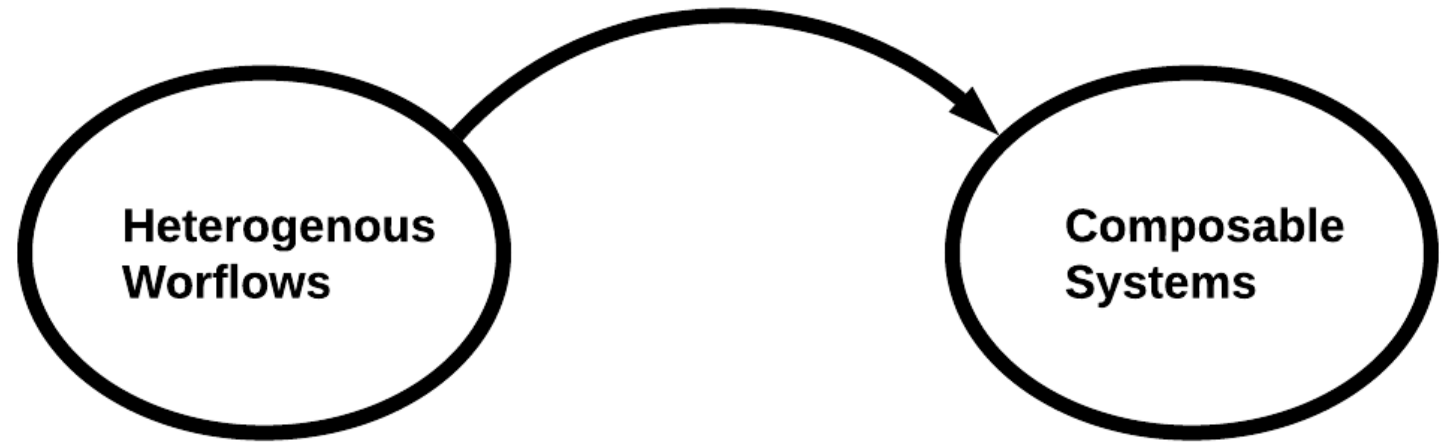


Use-Inspired Interfaces

3D Next Generation Fuel and Fire Models

Mapping between Heterogenous Workflows to Composable Systems

- Burnpro3D: Prescribed Fire Planning Platform
- JupyterHub: Sage Edge App Development and WIFIRE AI Gateway
- Other projects
 - Neurokube: AI Neuroscience imaging
 - TemPredict for IoT use in COVID detection



M. Madany, K. Marcus, S. Peltier, M. H. Ellisman and I. Altintas, "NeuroKube: An Automated and Autoscaling Neuroimaging Reconstruction Framework using Cloud Native Computing and A.I.," 2020 IEEE International Conference on Big Data (Big Data), Atlanta, GA, USA, 2020, pp. 320-330, doi: 10.1109/BigData50022.2020.9378053.

S. Purawat et al., "TemPredict: A Big Data Analytical Platform for Scalable Exploration and Monitoring of Personalized Multimodal Data for COVID-19," 2021 IEEE International Conference on Big Data (Big Data), 2021, pp. 4411-4420, doi: 10.1109/BigData52589.2021.9671441.

Artwork: **Jen Stark, Cosmographic, 2014**, acid-free paper, holographic paper, glue, wood, acrylic paint, 34 x 37 x 4 in.

To sum up...

Emerging new applications require integrated AI in dynamically composed workflows.



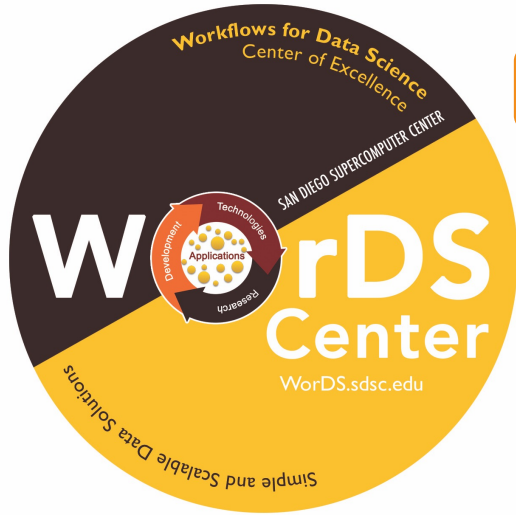
Embrace Complexity!

Complexity comes at a cost!

- Composable systems is not a turnkey functionality
- Requires collaboration with and between infrastructure providers
- End-to-end data pipelines need to be defined for each application along with microservice execution

Contact: *Ilkay Altintas, Ph.D.*

Email: ialtintas@ucsd.edu



<https://words.sdsc.edu/>

<https://wifire.ucsd.edu/>



We are hiring! -- <https://words.sdsc.edu/careers>

Questions?



U.S. DEPARTMENT OF
ENERGY

Office of
Science

The presented work is collaborative work with many wonderful individuals, and parts of it are funded by NSF, DOE, NIH, UC San Diego and various industry, government and foundation partners.