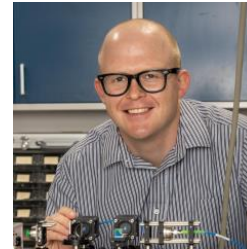


# ASCAC Early Career Virtual Panel

## Panelists



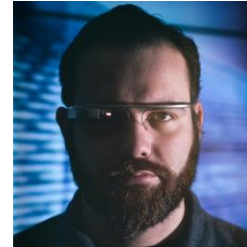
Andrey Lokhov  
Staff Scientist  
*Los Alamos National Laboratory*



Joseph M. Lukens  
Associate Professor  
*Purdue University*



Ana Gainaru  
Computer Scientist  
*Oak Ridge National Laboratory*



Victor Mateevitsi  
Computer Scientist  
*Argonne National Laboratory*

## Panel Moderator



Cindy Rubio González  
Associate Professor  
*University of California, Davis*



# ASCAC Early Career Virtual Panel

**Andrey Lokhov**

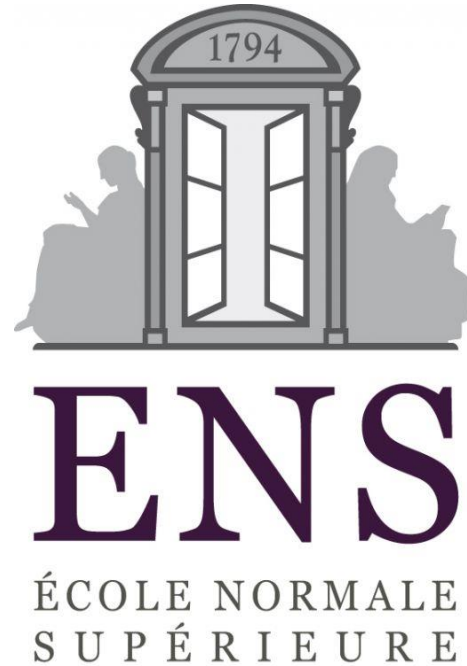
**Theoretical Division, Applied Mathematics Group**

January 2024

# Education and background



B.Sc. in **Physics** at Ecole Polytechnique, France



M.Sc. in **Theoretical Physics** at Ecole Polytechnique, France

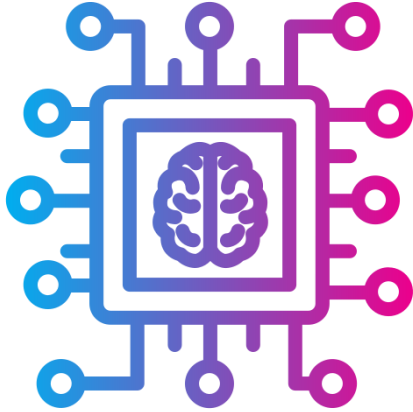


Ph.D. in **Statistical Physics** at Université Paris-Sud, France



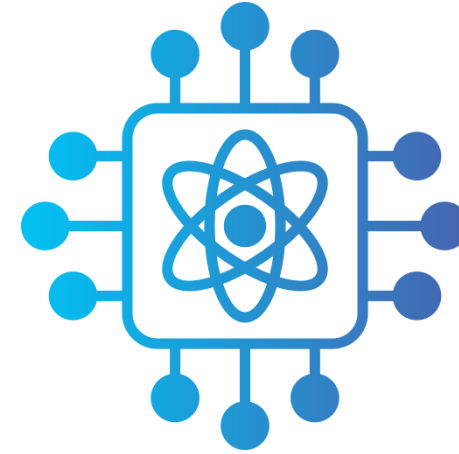
Postdoc → Scientist in **Applied Mathematics** at Los Alamos National Laboratory, USA

# Main areas of my work and funding



Statistical Learning

**Funding:** DOE Office of Science ASCR Early Career



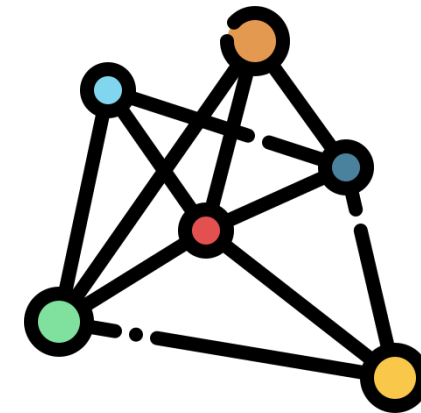
Quantum Computing

**Funding:** Defense Advanced Research Projects Agency (DARPA)



Power Grid

**Funding:** DOE Office of Electricity



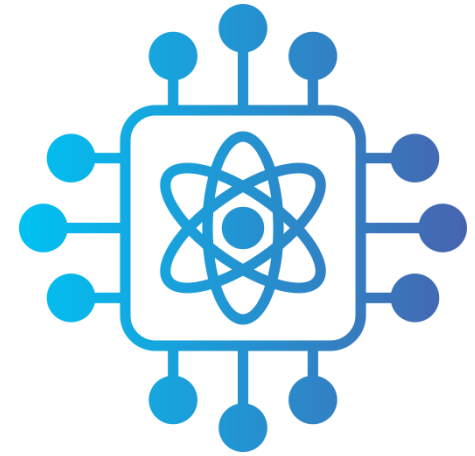
Graph Algorithms

**Funding:** Laboratory Directed Research and Development (LDRD)

# A Few Words on... Quantum Computing



Quantum  
Benchmarking  
Initiative

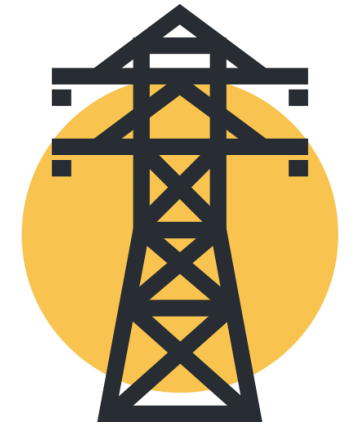


DARPA's **Quantum Benchmarking Initiative (QBI)** aims to determine if it's possible to build an industrially useful quantum computer much faster than conventional predictions.

Specifically, QBI is designed to rigorously **verify and validate** if any quantum computing approach can achieve utility-scale operation — meaning its computational value exceeds its cost — by the year 2033.

**Funding:** Defense Advanced Research Projects Agency (DARPA)

# A Few Words on... Power Grid



**Grid Controls and  
Communications**

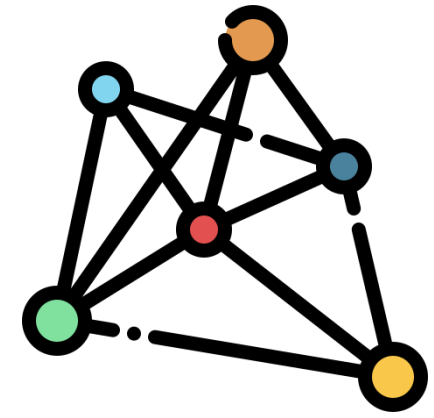
## **Real-time Valuation of Sensor Reliability:**

Establish and benchmark a high-performance online sensor valuation method based on a novel idea for reliability assessment of measurements;

Use the developed metric to inform optimal sensor placement tools for enhanced grid reliability.

**Funding:** DOE Office of Electricity

# A Few Words on... Graph Algorithms



## **Novel Solutions for Out-of-Equilibrium Dynamics of Disordered Systems:**

Is it possible to solve out-of-equilibrium reversible dynamics of disordered systems (epidemic spreading, dynamics of magnetic materials, *etc.*) and beat traditional Markov Chain Monte Carlo (MCMC) methods?

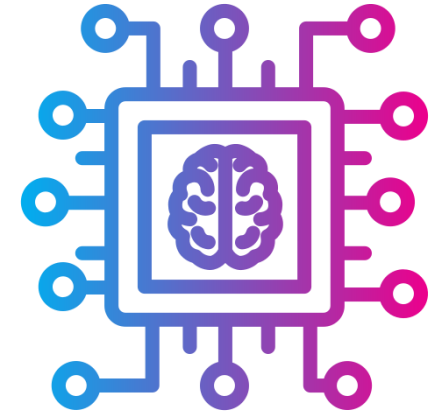
Making progress using a class of algorithms called dynamic message-passing (DMP)

**Funding:** Laboratory Directed Research and Development (LDRD)

# A Few Words on... Statistical Learning



**Advanced Scientific  
Computing Research**



## **Resurgence of Markov Random Fields for Scientific Machine Learning: New Mathematics for an Old Framework**

Advance mathematics behind machine learning methods for Markov Random Fields (MRFs);

Explore potential of MRFs to become a dominant paradigm for Scientific Machine Learning, and make impact in applications: many-body quantum physics, power grids, statistical physics, and field theories

**Funding:** DOE Office of Science ASCR Early Career Program



# A Few Words on... Statistical Learning

**Boltzmann distribution** (Physics)

**Gibbs measure** (Mathematics)

**Exponential family** (Statistics)

**Undirected graphical models, Markov random fields** (Computer Science)

**Log-linear model, Energy-based model, Boltzmann machine** (Machine Learning)

$$\mu_{\theta}(\mathbf{x}) = \frac{1}{Z} \exp(-H_{\theta}[\mathbf{x}]) = \frac{1}{Z} \exp\left(-\sum_{k \in K} \theta_k q_k[\mathbf{x}]\right)$$

Gibbs distribution

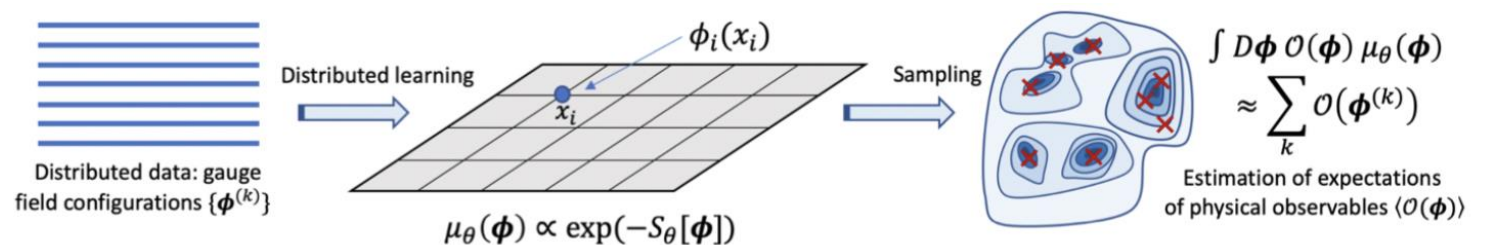
“Hamiltonian” energy function

Parameters

Sufficient statistics

## Applications:

- many-body quantum states
- power grids
- statistical physics
- field theories



**Funding:** DOE Office of Science ASCR Early Career Program

# Team



**Dr. Abhijith Jayakumar**

Theory of MRF learning,  
Quantum Computing



**Dr. Minh Vu**

Power grid dynamics



**Dr. Shreya Shukla**

Applications in field theories



**Dr. Karthik Elamvazhuthi**

Algorithms for MRF learning



**Dr. Shiba Biswal**

Algorithms for MRF sampling

# ASCAC Early Career Virtual Panel

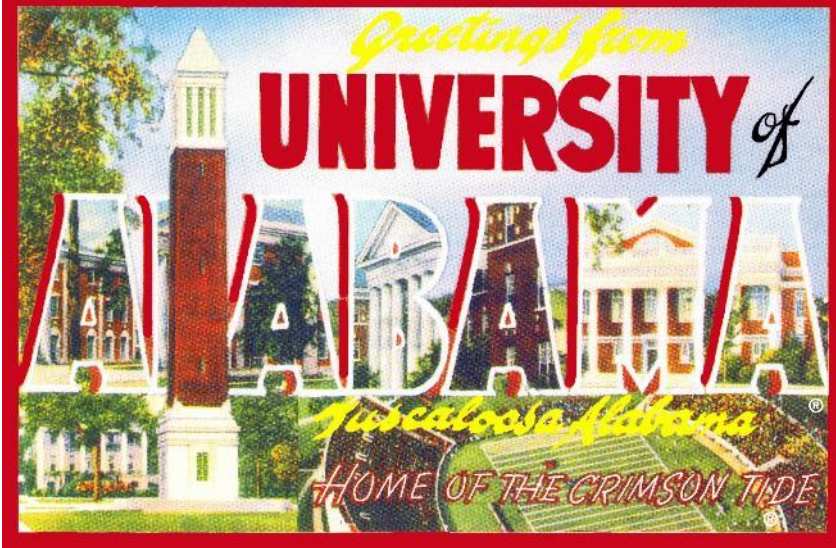
Joseph M. Lukens

Purdue University

16 Jan. 2025

# About myself: education

University of Alabama (2007–2011)



**B.S. Electrical Engineering & Physics**

Undergraduate research with  
Profs. Margaret Kim & Patrick  
Kung



Purdue University (2011–2015)



**Ph.D. Electrical Engineering**

Graduate research  
with Andy Weiner

**Dissertation**

*Novel applications of  
photonic signal processing:  
temporal cloaking and  
biphoton pulse shaping*



In memoriam  
(1958–2024)

# About myself: career

## Oak Ridge National Lab

- 2015–present
  - Eugene P. Wigner Fellow (2015–2018)
  - Research Scientist (2018–2022)
  - Joint Faculty Appointment (2022–present)
- Won Early Career Award from DOE ASCR in 2019
  - *Scalable Architectures for Hybrid Quantum/Classical Networking*



## Arizona State University

- 2022–present
  - Sr. Director of Quantum Networking & Research Professor (2022–2024)
  - Adjunct Professor (2025–present)



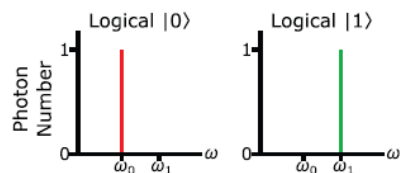
## Purdue University

- 2025–present
  - Associate Professor of Electrical & Computer Engineering

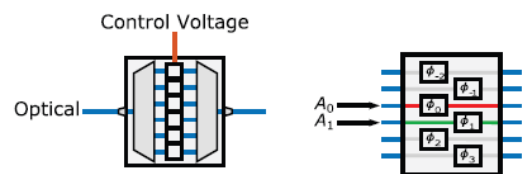


# Research interests

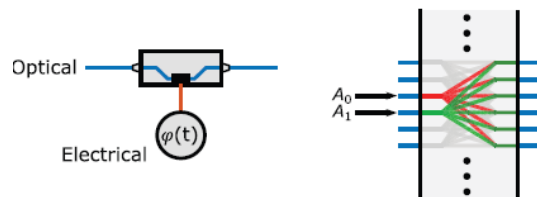
## Frequency-bin quantum information



### PULSE SHAPER

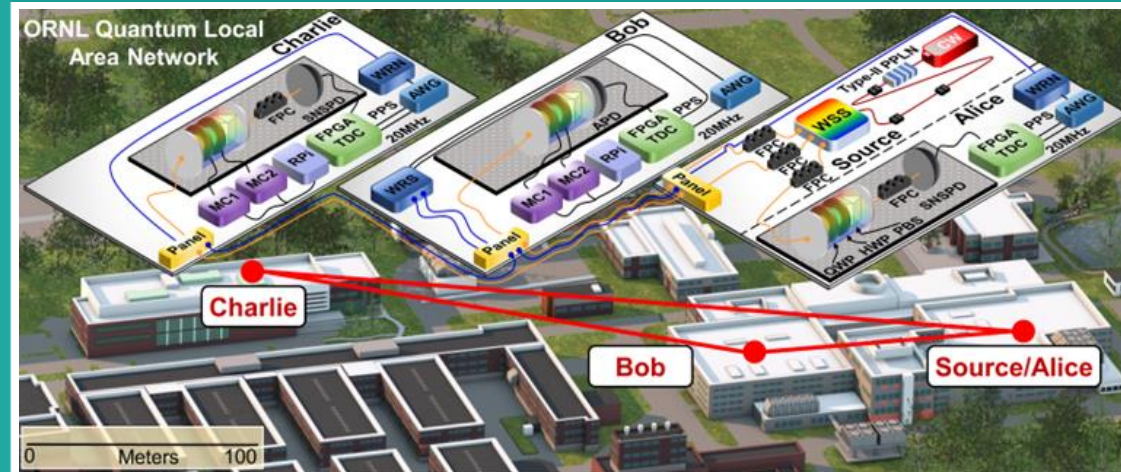


### PHASE MODULATOR



J. M. Lukens & P. Lougovski, *Optica* **4**, 8 (2017).

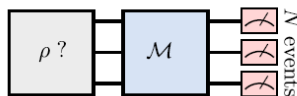
## Quantum networking



M. Alshowkan, B. P. Williams, P. G. Evans, N. S. V. Rao, E. M. Simmerman, H.-H. Lu, N. B. Lingaraju, A. M. Weiner, C. E. Marvinney, Y.-Y. Pai, B. J. Lawrie, N. A. Peters, & J. M. Lukens, *PRX Quantum* **2**, 040304 (2021).

## Bayesian quantum state tomography

(a) Experiment



(b) LS projection

$$\rho_{LS} = [P_{\mathcal{M}}(\rho)]_{LS}$$

(c) Parameterization

$$\mathbf{x} = \{y_1, \dots, y_D, z_1, \dots, z_D\}$$

$$\rho(\mathbf{x}) = \sum_{k=1}^D \left( \frac{y_k}{\sum_l y_l} \right) \frac{\mathbf{z}_k \mathbf{z}_k^\dagger}{|\mathbf{z}_k|^2}$$

(d) Prior

$$Y_k \sim \Gamma(\alpha, 1) \quad \mathbf{Z}_k \sim \mathcal{CN}(0, I_D)$$

$$\pi_0(\mathbf{x}) \propto \prod_{k=1}^D y_k^{\alpha-1} e^{-y_k} e^{-\frac{1}{2} \mathbf{z}_k^\dagger \mathbf{z}_k}$$

(e) Pseudo-likelihood

$$L_{\mathcal{D}}(\mathbf{x}) = \exp\left(-\frac{N}{2} \|\rho_{\mathcal{M}}(\rho(\mathbf{x})) - \rho_{LS}\|_F^2\right)$$

(f) Sample from posterior

$$\pi(\mathbf{x}) = \frac{1}{Z} L_{\mathcal{D}}(\mathbf{x}) \pi_0(\mathbf{x})$$

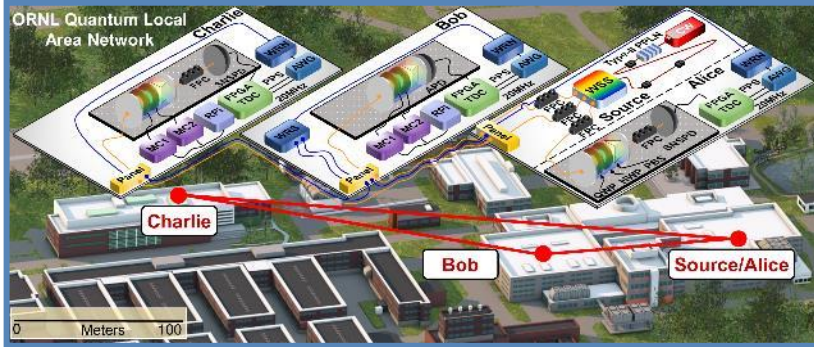
(g) Estimate expectation values

$$\langle \phi(\rho) \rangle = \int d\mathbf{x} \pi(\mathbf{x}) \phi(\rho(\mathbf{x}))$$

$$\approx \frac{1}{R} \sum_{r=1}^R \phi(\rho(\mathbf{x}^{(r)}))$$

J. M. Lukens, K. J. H. Law, A. Jasra, & P. Lougovski, *New J. Phys.* **22**, 063038 (2020).

## QLAN testbed



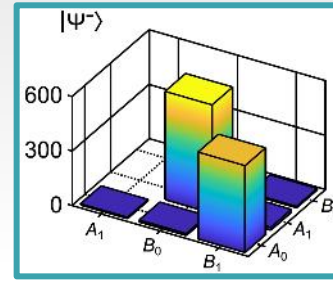
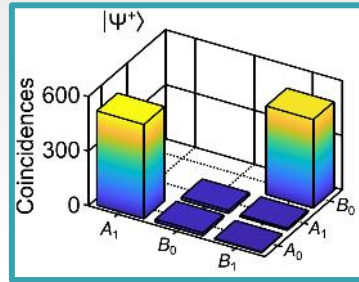
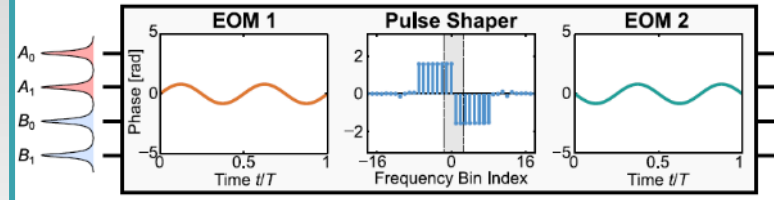
- First deployed **flex-grid** quantum network.
- Picosecond timing with **White Rabbit**.
- Secure classical comms with **QKD**.

M. Alshowkan *et al.*, *PRX Quantum* **2**, 040304 (2021).  
M. Alshowkan *et al.*, *J. Opt. Commun. Netw.* **14**, 493 (2022).



**Artifacts**  
35 journal articles  
47 conference papers  
1 book chapter  
1 patent

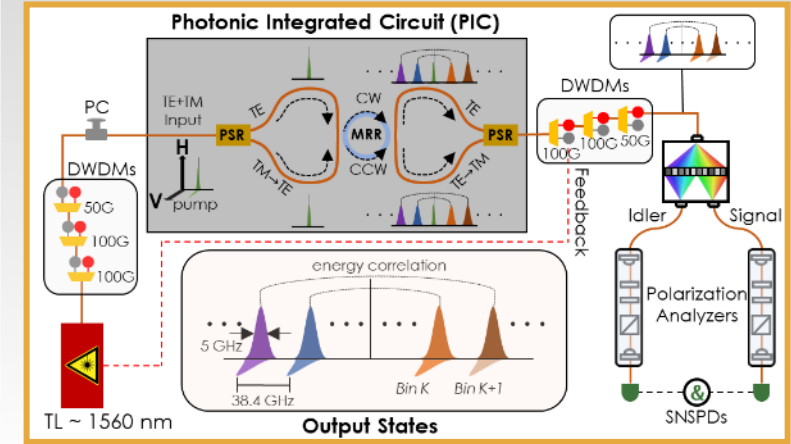
## Frequency-bin networking



- First **Bell state analyzer** for frequency-bin qubits.
- Building block for **frequency-based quantum repeaters**.

N. B. Lingaraju *et al.*, *Optica* **9**, 280 (2022).

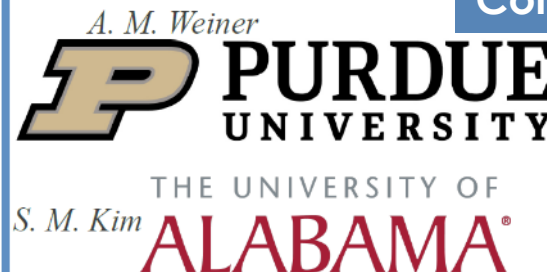
## Silicon photonics



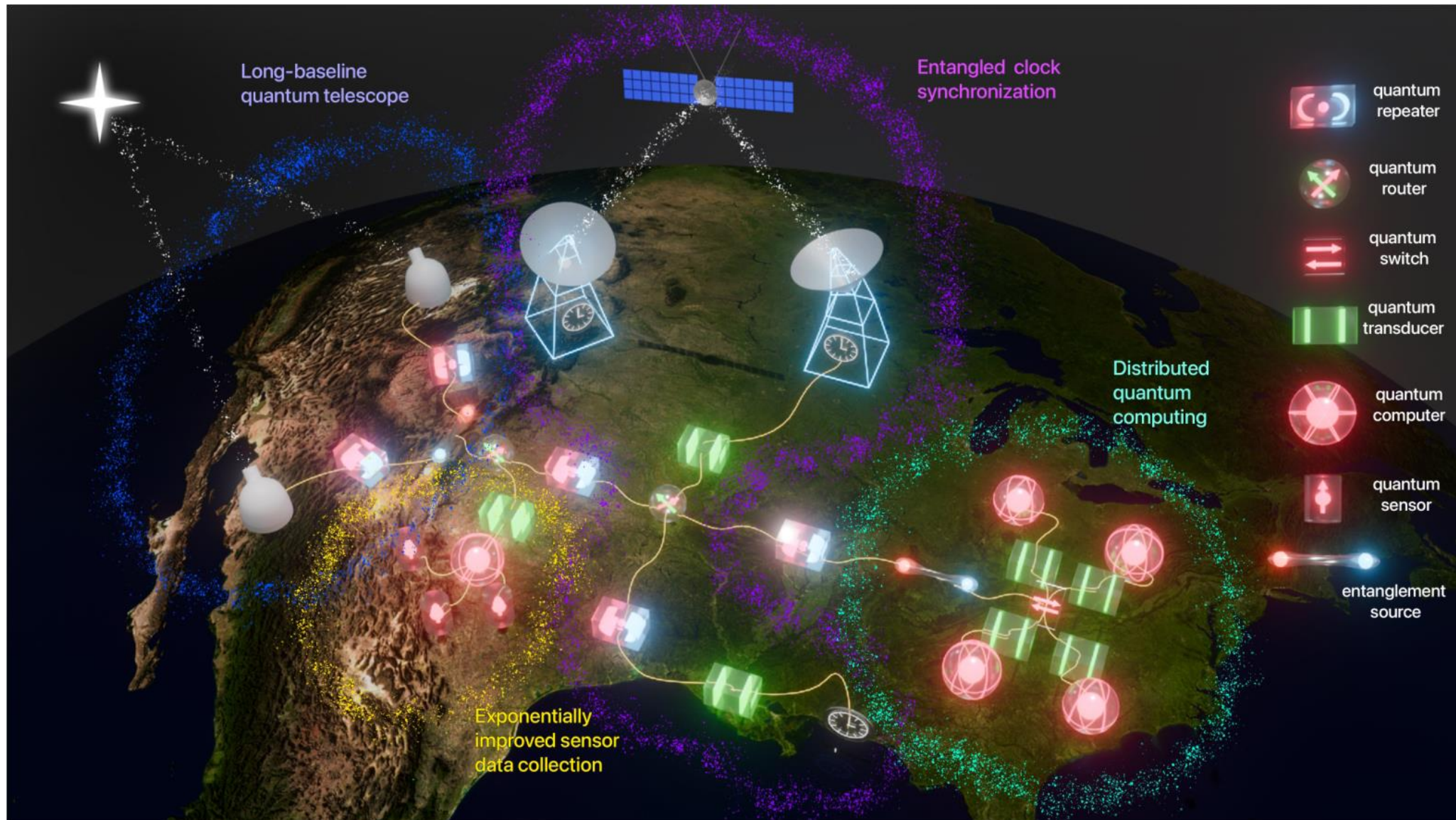
- **All-CMOS** design for frequency-polarization entanglement.
- Over **100 channels** of high-fidelity polarization states.

A. Miloshevsky *et al.*, *Optica Quantum* **2**, 254 (2024).

## Collaborators



# A possible future...



from the *Quantum Information Science Applications Roadmap for the U.S. Department of Energy* (December 2024)



# DOE ASCAC Early Career Virtual Panel

**Ana Gainaru**

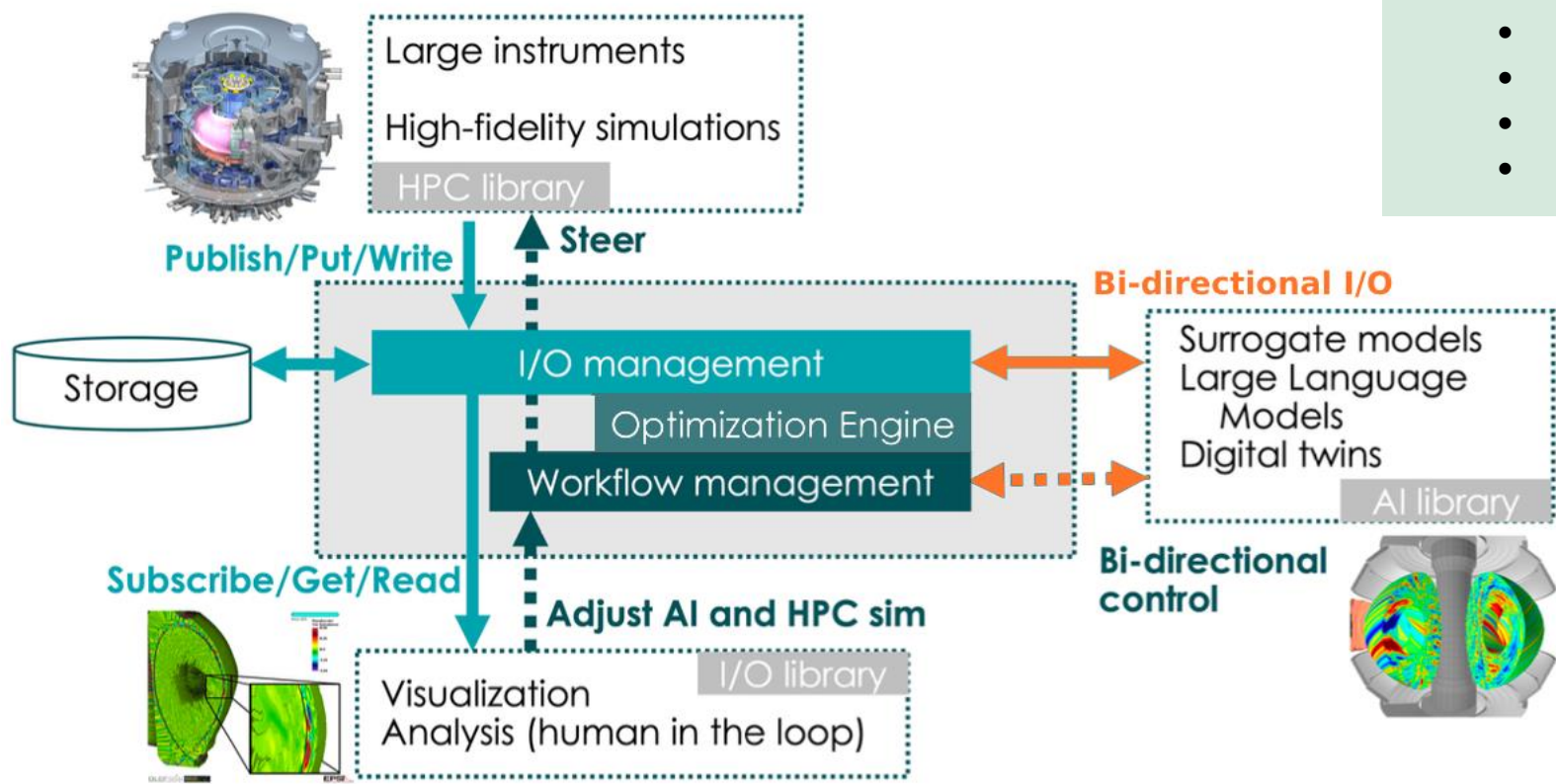
Workflow System Group, Data and AI systems Section  
CSMD, Oak Ridge National Laboratory

Data and workflow management for complex scientific application

ORNL is managed by UT-Battelle LLC for the US Department of Energy

# HPC application landscape

Applications are becoming proxy for experiments



Workflows combine

- High fidelity simulations
- Instruments
- Visualization services
- Analysis
- AI surrogate models
- Digital twins

SKA (Square Kilometer Array)  
- Generating 1 TB/s data

Cancer research  
- Ensemble runs for 10k+ images

Fusion  
- Simulations training digital twins  
- Bi-directional feedback  
- Human in the loop

# Background

- Interests
    - **Data fusion** with extreme and small data
    - **Coupling** between traditional HPC and AI
    - Extract **knowledge** from data
    - Smart **querying** capabilities
      - For data or for performance monitoring
- ASCR projects
    - SciDAC FES
    - Data management
    - RAPIDS

- PhD at **UIUC**
  - The Blue Waters project, **performance optimization and resilience**
  - Advisor: Marc Snir



- **Mellanox**
  - **Data transfer** optimization for MPI collective communication using **networking** features

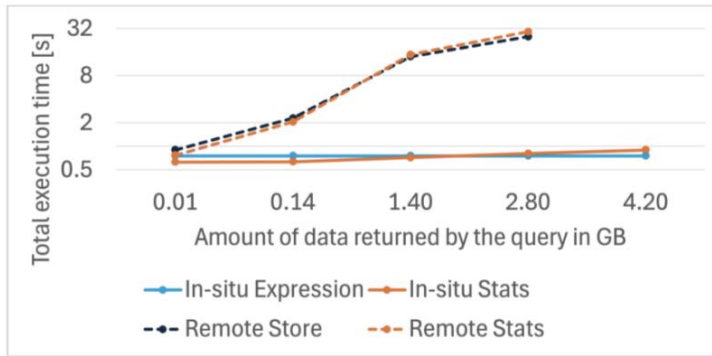
- **Vanderbilt** University
  - **Scheduling** stochastic applications of HPC systems
  - Medical **AI applications**

- **ORNL** (since 2020)
  - **I/O** optimizations and **query** engine
  - **Workflow** modeling and scheduling



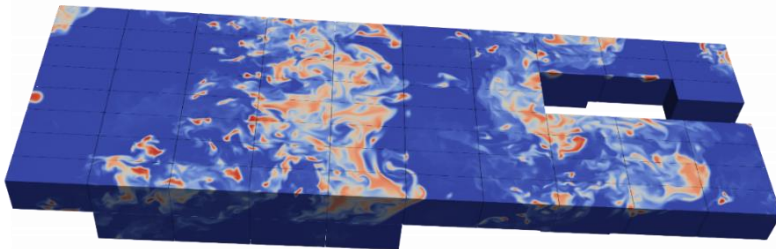
Research focusing on optimizing the performance of applications using HPC

## Data management



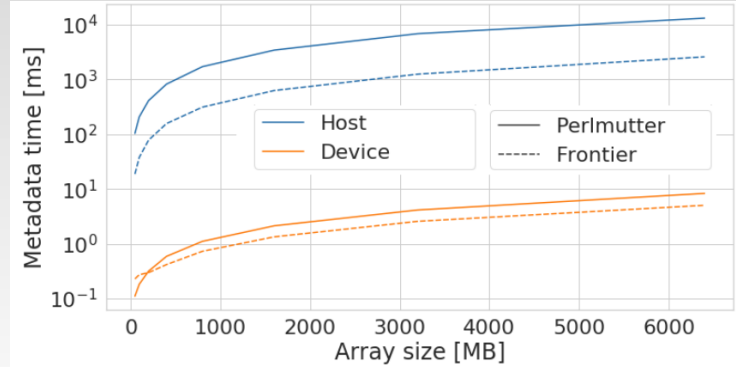
In-situ and remote analysis of S3D data

- Complex querying
  - Using quantities of interest
  - User-defined accuracy
  - On remote data



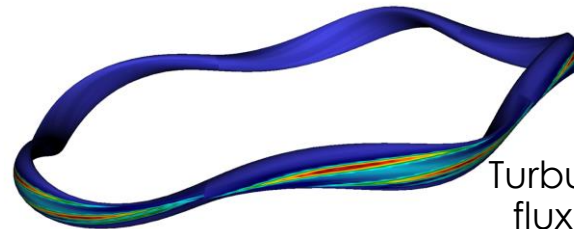
Local visualization in a region of interest with 0.1 accuracy

## Workflow management



Metadata time between the CPU and GPU backend used by I/O libraries

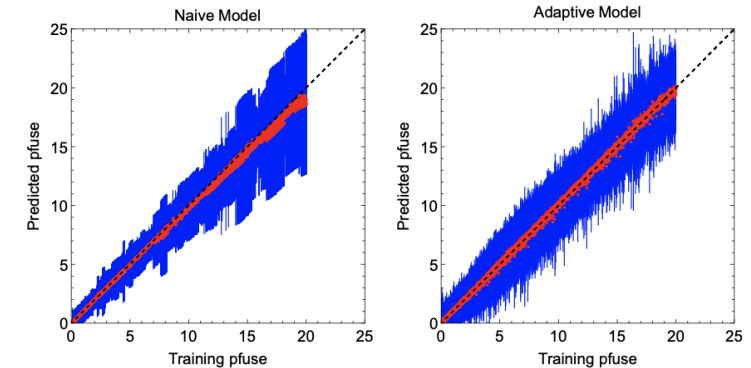
- Code coupling
  - The output time on Frontier is 17 times faster than the previous I/O interface
  - The cost of code coupling is less than 1%



Turbulence intensity on one of the islands flux surfaces obtained running GX with Trinity and Genray

## Application impact

- Adaptive training workflow optimization
  - Streaming improved the I/O time by 75%
  - Having a GPU backend available in ADIOS improves I/O by 15%

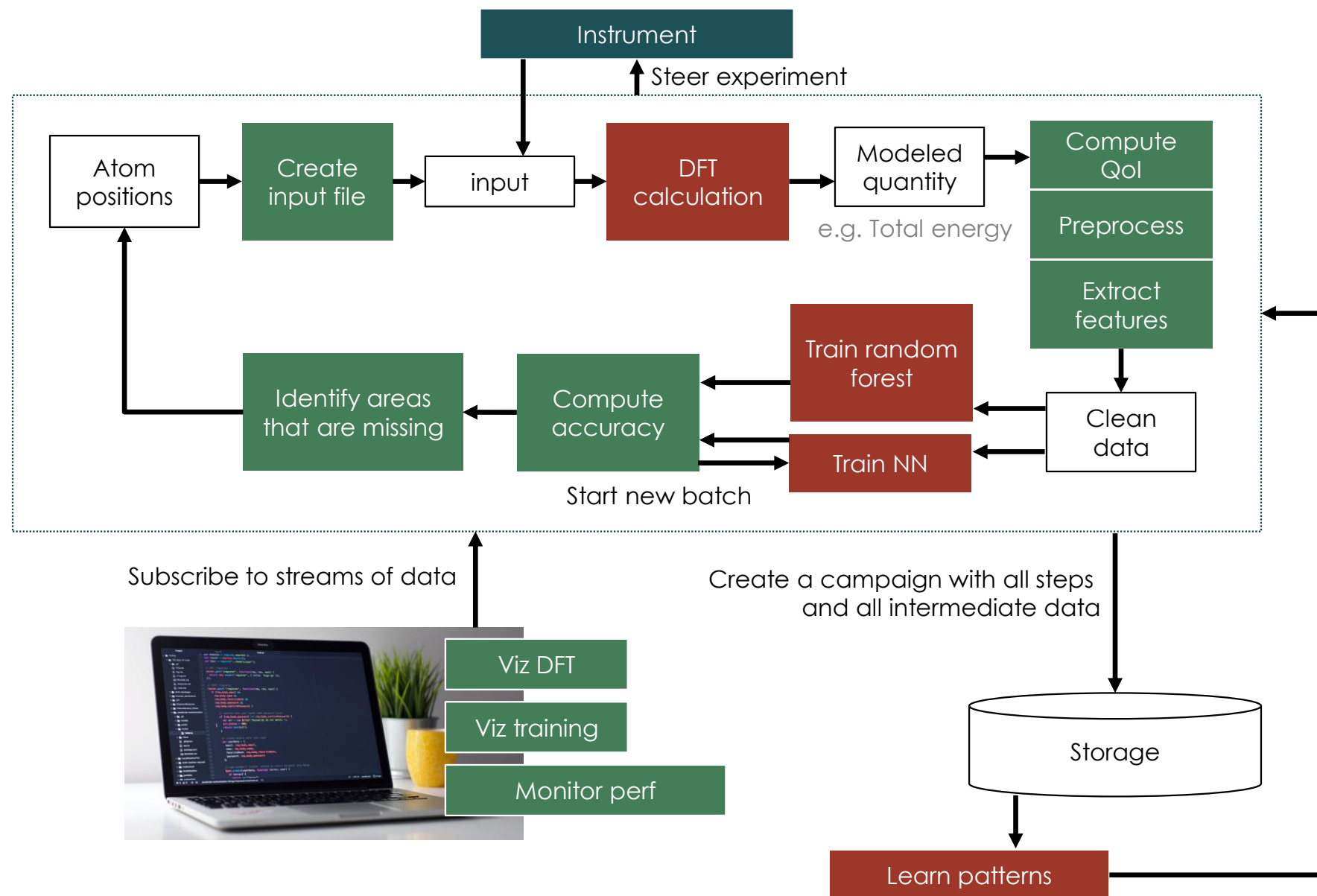


An adaptively trained model provides more accurate prediction at high electron fusion power output.

# Future: AI driven workflows

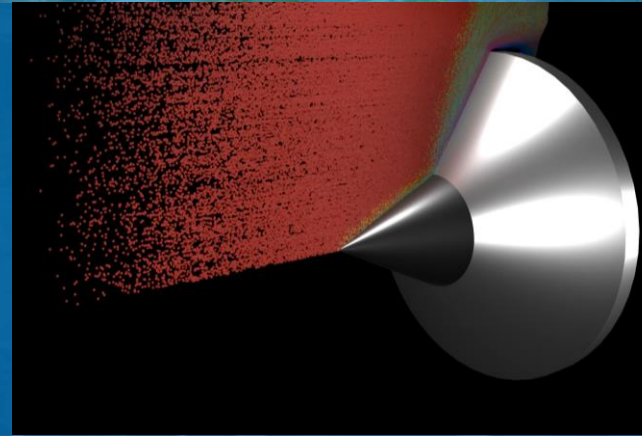
## Challenges

- Digital twin creation and updating
- Query campaigns of runs
- Schedule all intermediate analysis codes
  - Agent based
- Keep track of intermediate representation of data
- Learn from access patterns
  - Prefetch and pre-compute



JANUARY 16, 2025

# DOE ASCAC EARLY CAREER VIRTUAL PANEL



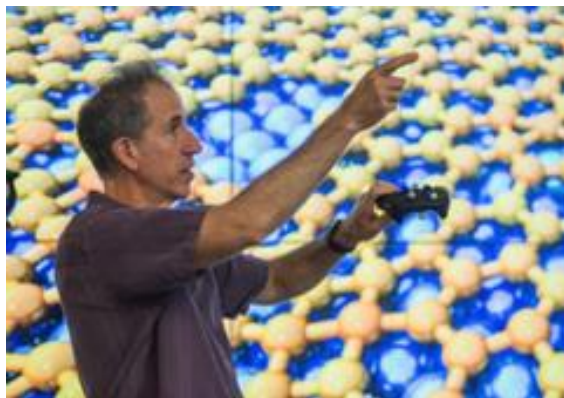
**VICTOR MATEEVITSI**  
Assistant Computer Scientist



Argonne National Laboratory is a  
U.S. Department of Energy laboratory  
managed by UChicago Argonne, LLC.



# ALCF VISUALIZATION AND DATA ANALYTICS GROUP



Joe Insley  
*Team Lead*



Silvio Rizzi  
*Computer Scientist*

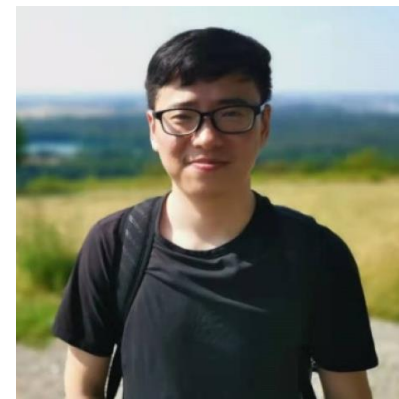


Victor Mateevitsi  
*Assistant Computer Scientist*

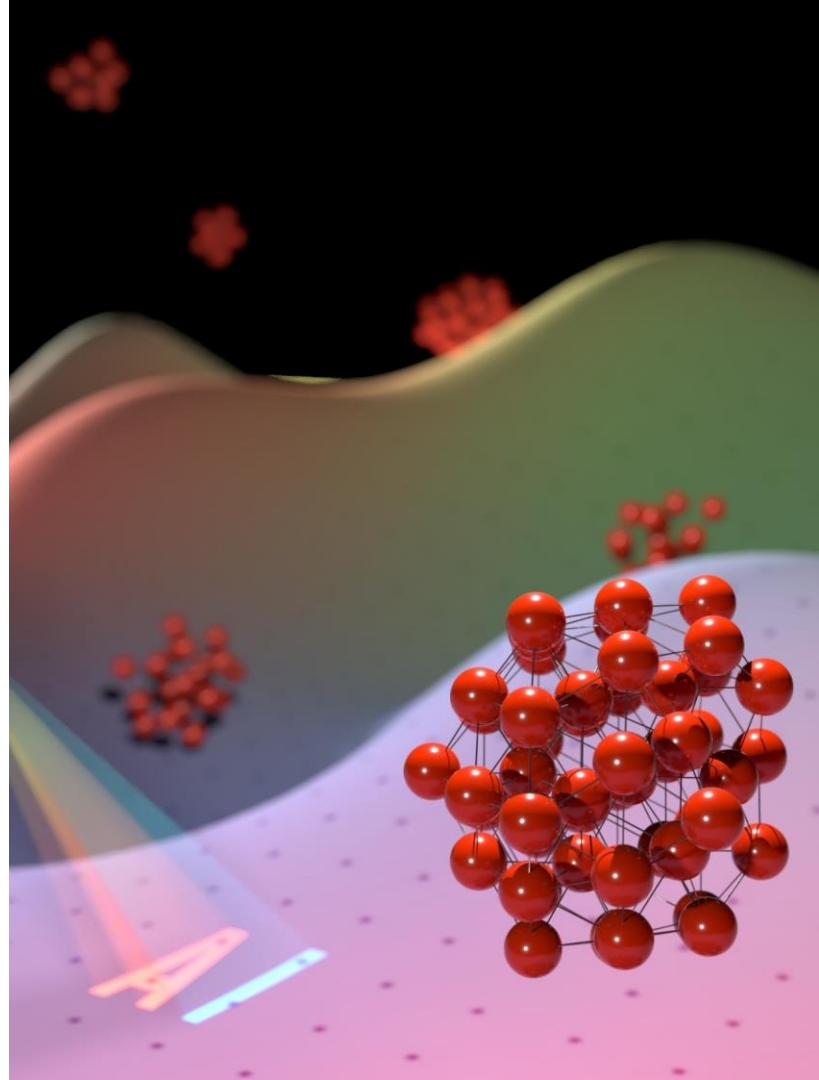
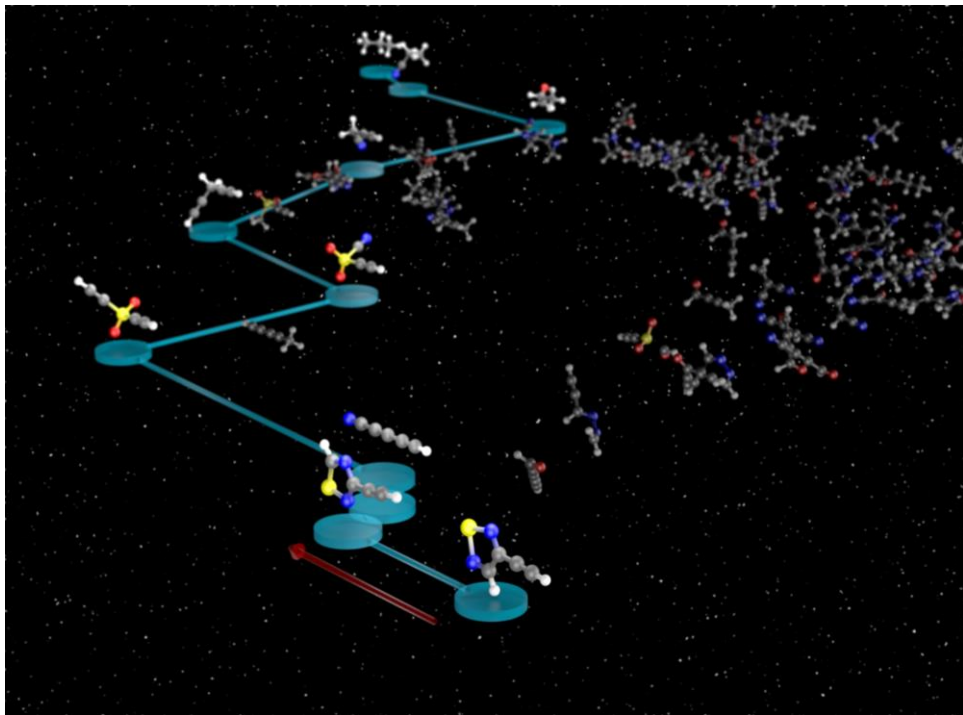


Janet Knowles  
*Principal Software  
Engineering Specialist*

Geng Liu  
*Postdoctoral Appointee*

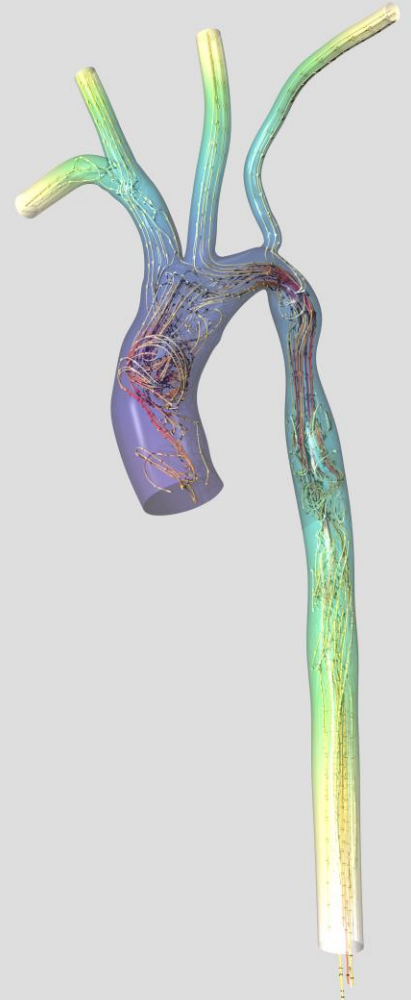
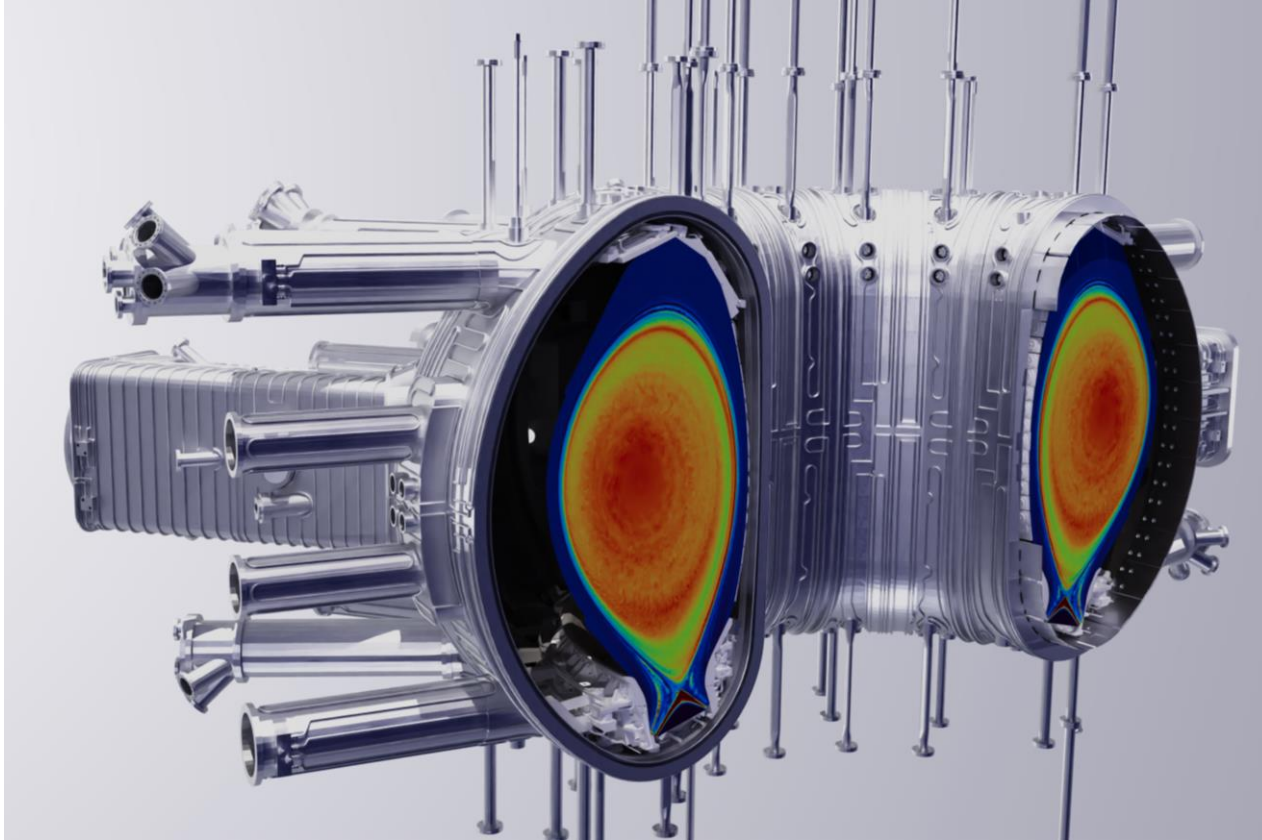


# PRODUCTION VIS

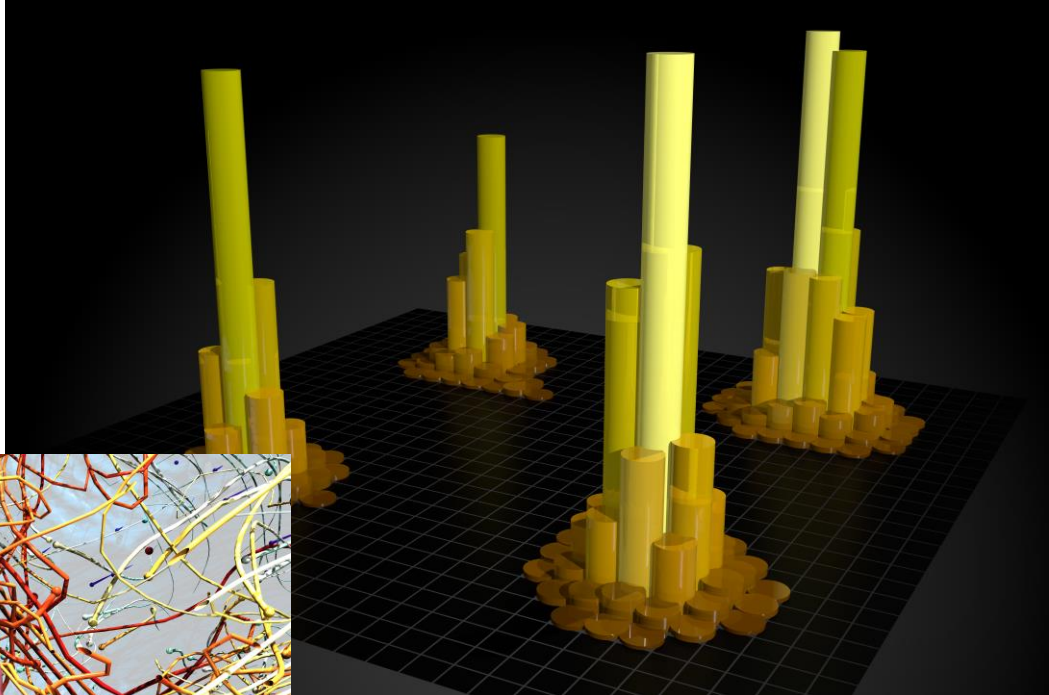
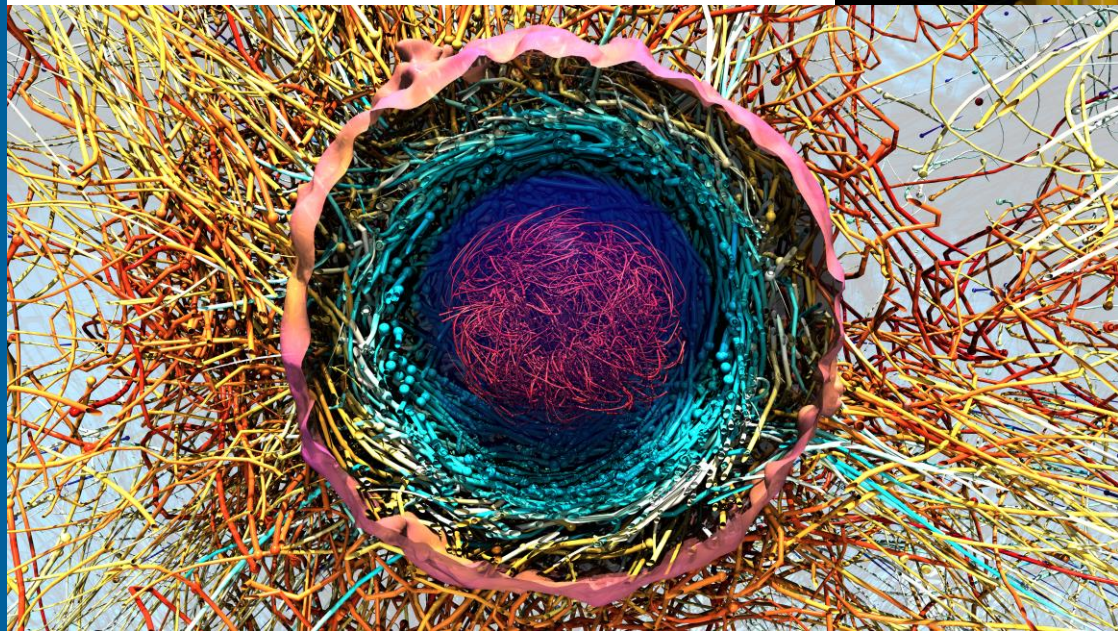




# PRODUCTION VIS







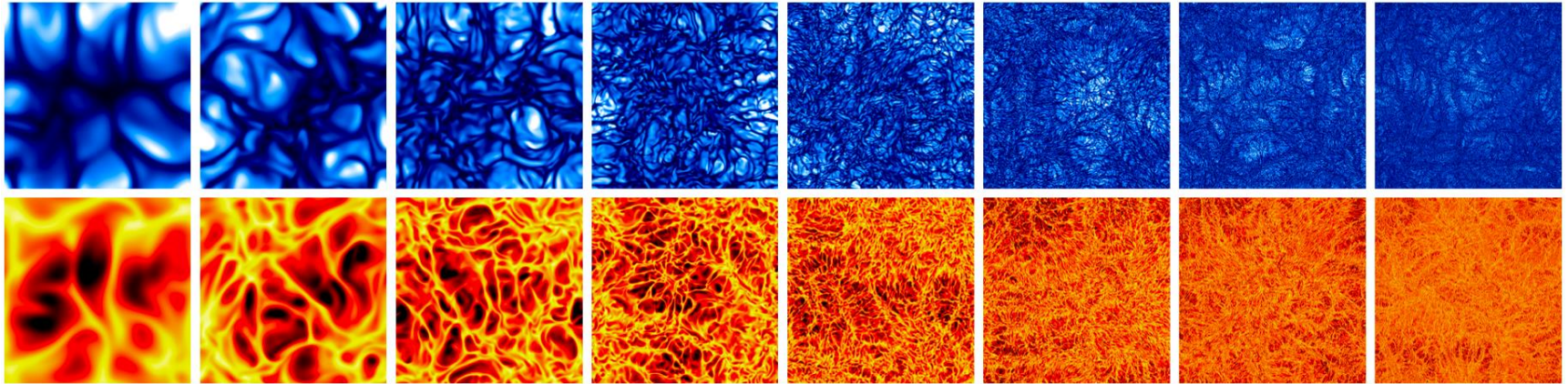
# PRODUCTION VIS



# INSITU VIS

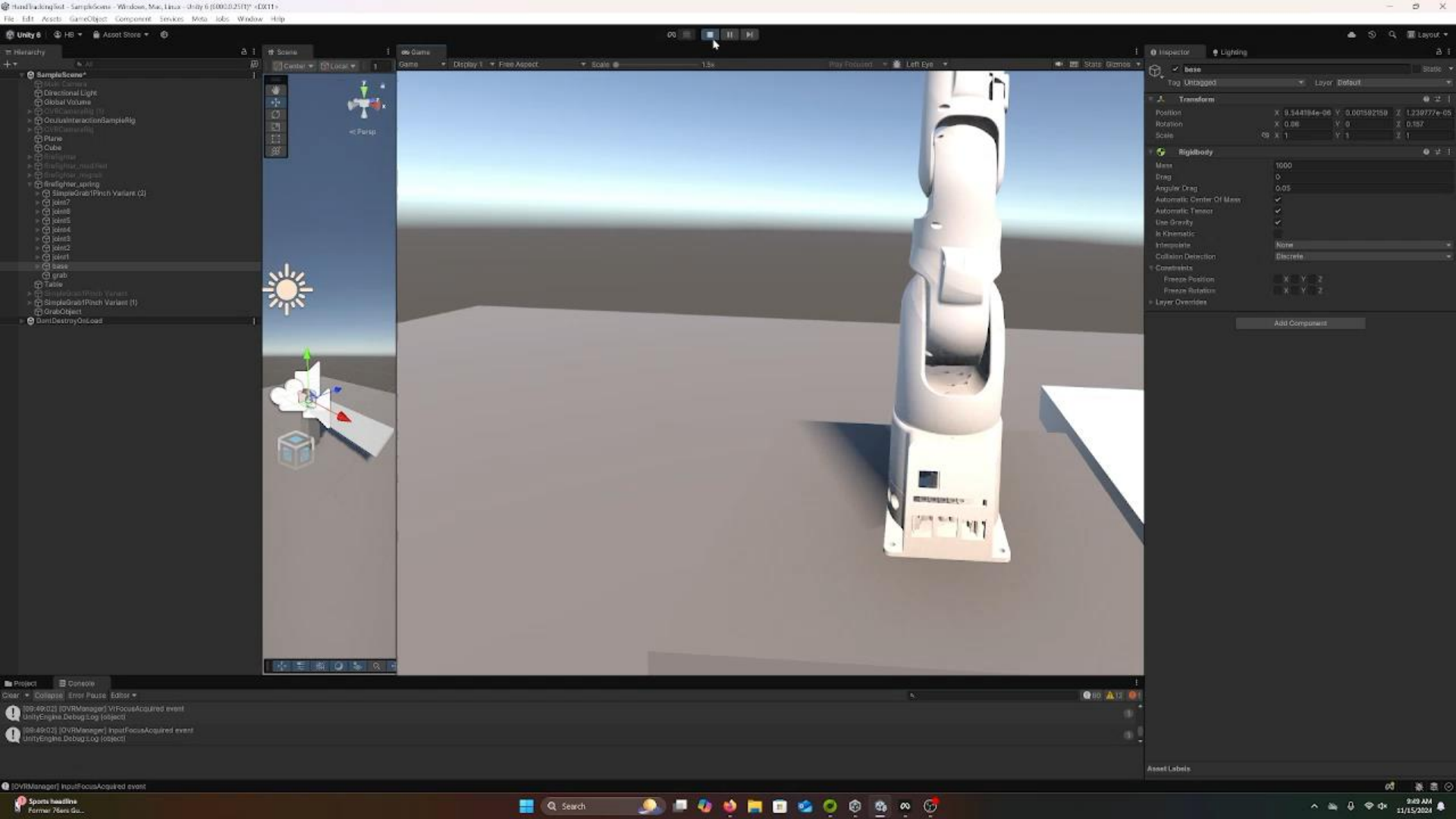


- In-situ visualization of NekRS (CFD code)  
- Bi-directional steering  
- Concurrent CPU (LDAV poster)



# TELE-ROBOTICS TO TELE-AUTONOMOUS ROBOTICS FOR ISOTOPE PRODUCTION





# DIGITAL TWINS AND AI-ENABLED & IMMERSIVE ENVIRONMENTS FOR AUTOMATED SCIENTIFIC LABORATORIES (DAIMSL)

Goal: create **automated, self-driving laboratories** by integrating robotics, augmented/virtual reality (AR/VR), and digital twin (DT) technologies.

- AI driven-process automation: Record human-operated robotic tasks to **train AI models** that automate these tasks over time
- Digital Twins: model and simulate lab processes in real-time to predict and prevent risks, ensure protocols are followed, and optimize experimental outcomes.



# THANK YOU!

- Questions?

