

# QuantOm Collaboration

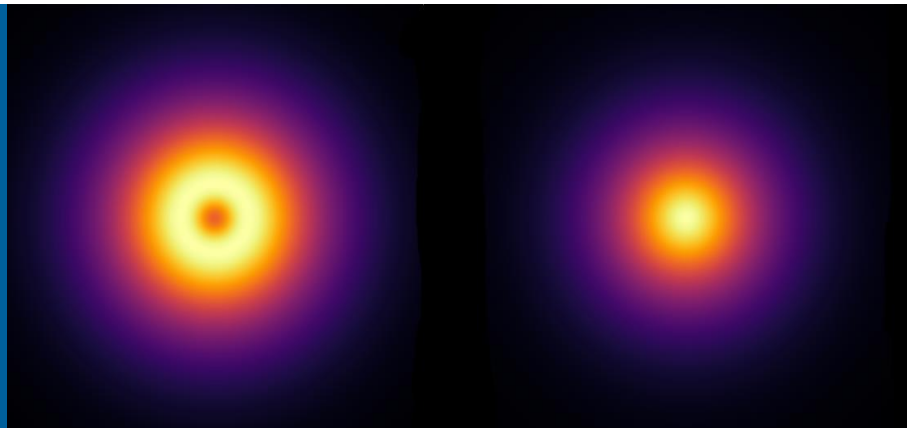


This work is supported by the U.S. Department of Energy, Office of Science, Office of Nuclear Physics, Office of Advanced Scientific Computing Research through the Scientific Discovery through Advanced Computing (SciDAC) program, under contracts DE-AC02-06CH11357, DE-AC05-06OR23177, and DE-SC0023472, in collaboration with Argonne National Laboratory, Jefferson Lab, National Renewable Energy Laboratory, Old Dominion University, Ohio State University, and Virginia Tech.

<https://quantom-collab.github.io>

## SAGIPS: A SCALABLE FRAMEWORK FOR SCIDAC QUANTOM

*SCIDAC-5: FEMTOSCALE IMAGING OF NUCLEI  
USING EXASCALE PLATFORMS*



**DANIEL LERSCH**

For the QuantOm collaboration

AI4EIC 2025, MIT, Boston

October 29, 2025

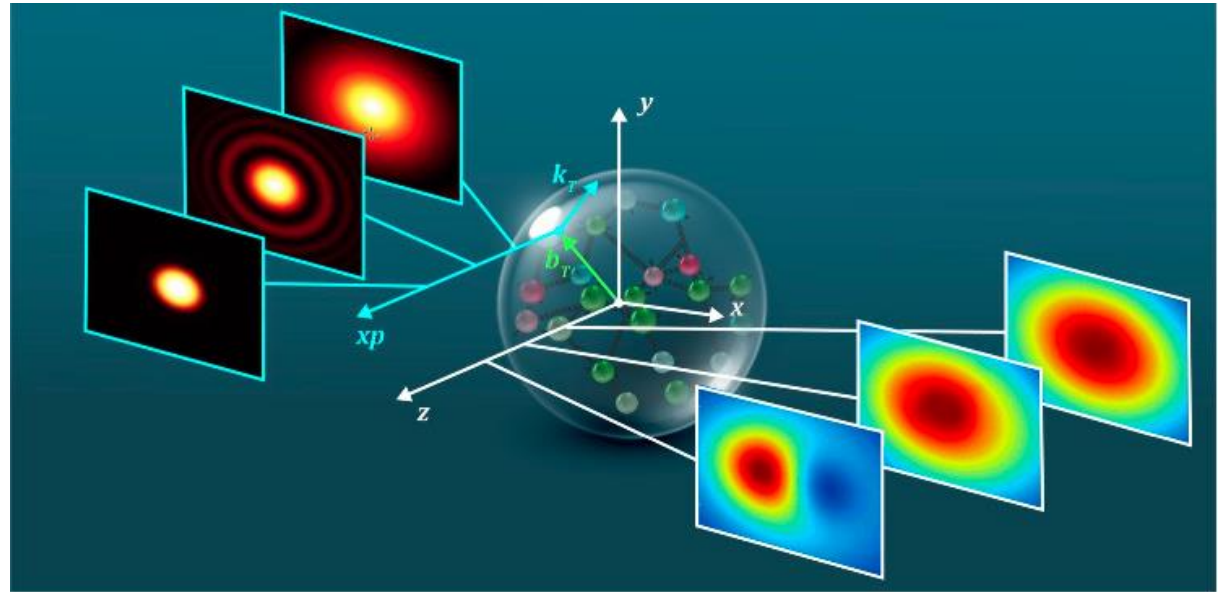


U.S. DEPARTMENT  
of ENERGY



# Exploring the Quark-Gluon Structure

- QUAntum Nuclear TOMography (QUANTOM)
- Extract Quantum Correlation Functions (QCFs) from Deep Inelastic Scattering (DIS) experiments
- Joint theoretical-experimental analysis
- 3D imaging of nuclei



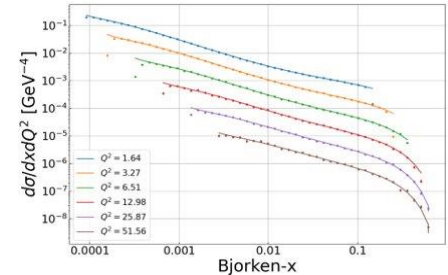
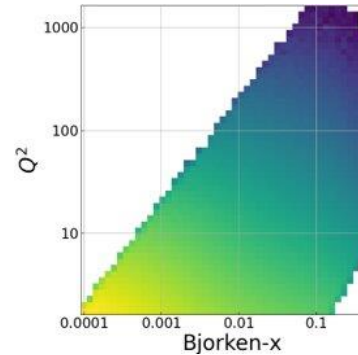
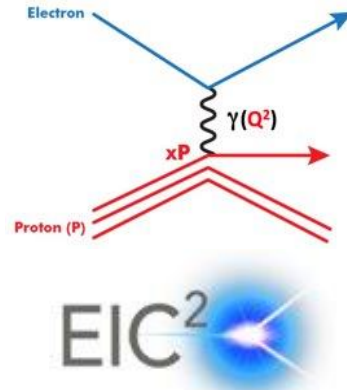
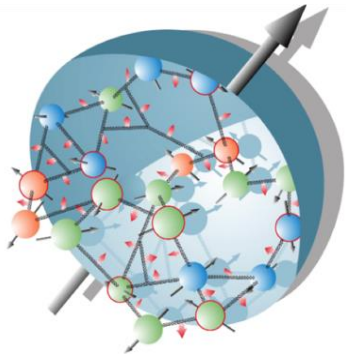
# Towards a joint Theoretical-Experimental Analysis of QCF

QCD

Experiments

Observables

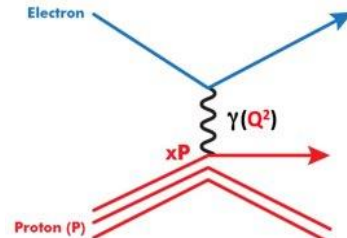
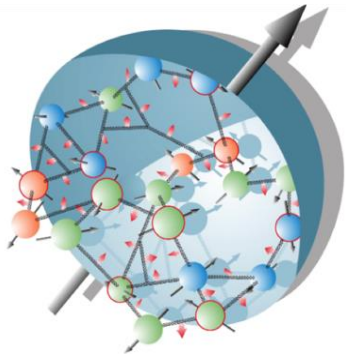
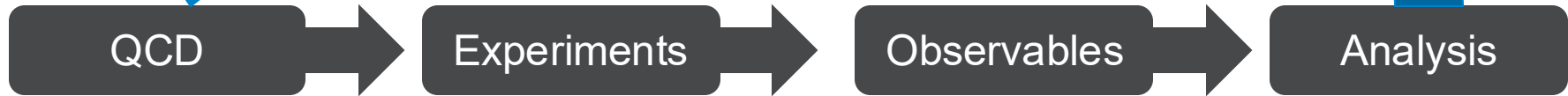
Analysis



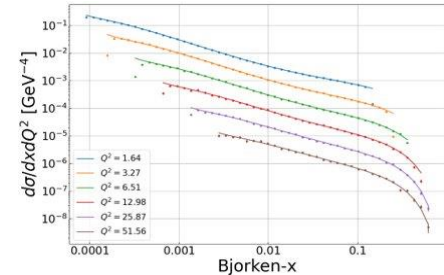
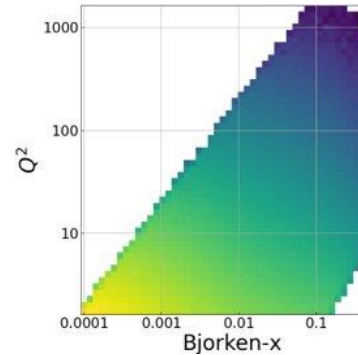
# Towards a joint Theoretical-Experimental Analysis of QCF

conventional

- Binned analysis
- Acceptance corrections
- Fit experimental data
- Extract QCFs



EIC<sup>2</sup>



# Towards a joint Theoretical-Experimental Analysis of QCF

Event-Level Fit

conventional

- Unbinned analysis
- Extract QCFs directly from experimental data
- Access to higher dimensional correlated data

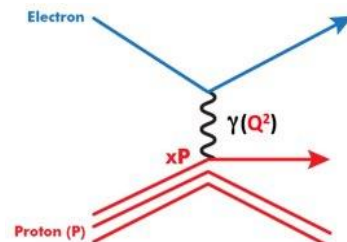
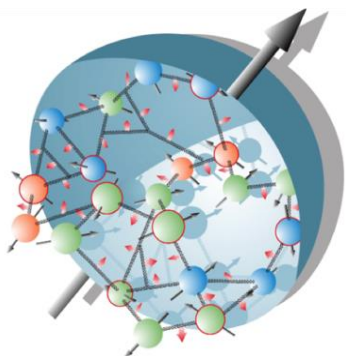
- Binned analysis
- Acceptance corrections
- Fit experimental data
- Extract QCFs

QCD

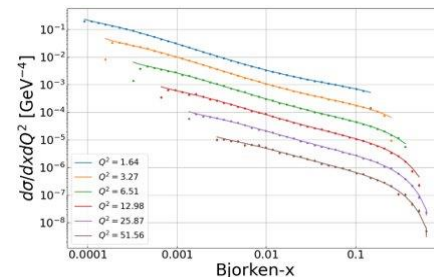
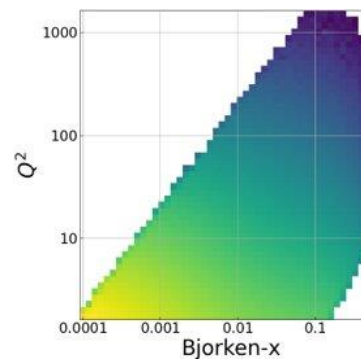
Experiments

Observables

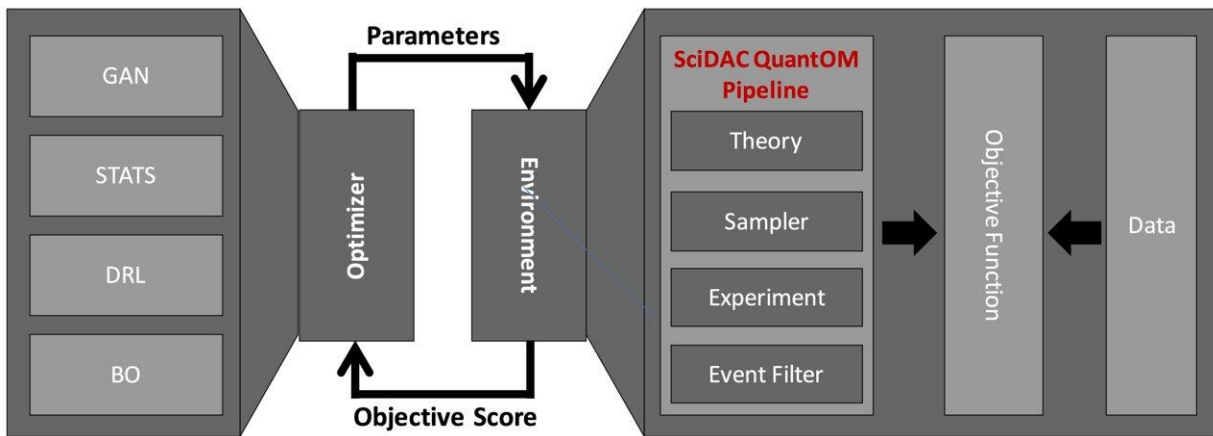
Analysis



EIC<sup>2</sup>

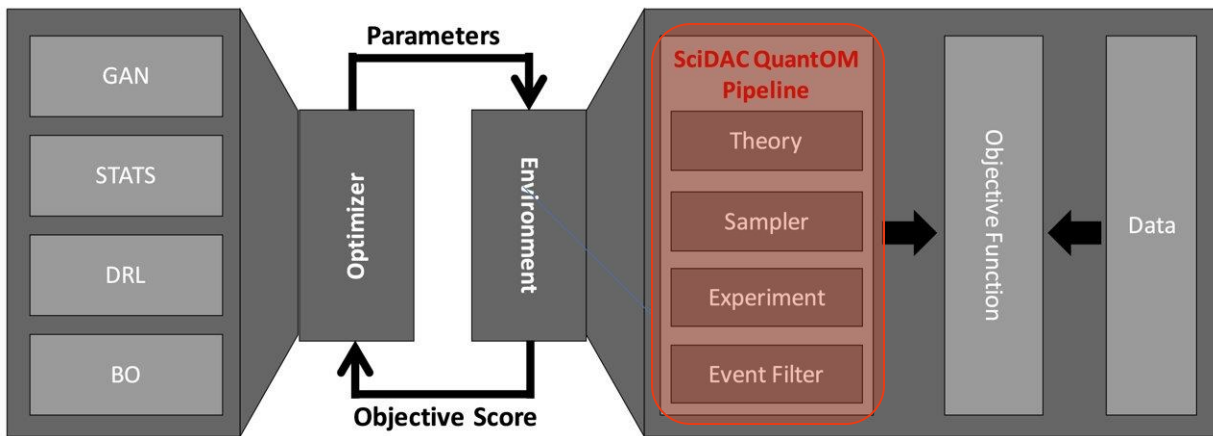


# SAGIPS: Scalable Asynchronous Generative Inverse Problem Solver



- Modular framework
- Adjustable pipeline translates predictions to physics events
- Leverage input from theory and experiment
- Currently 2 optimization approaches available
  1. Statistical workflow
  2. GAN workflow (today's focus)

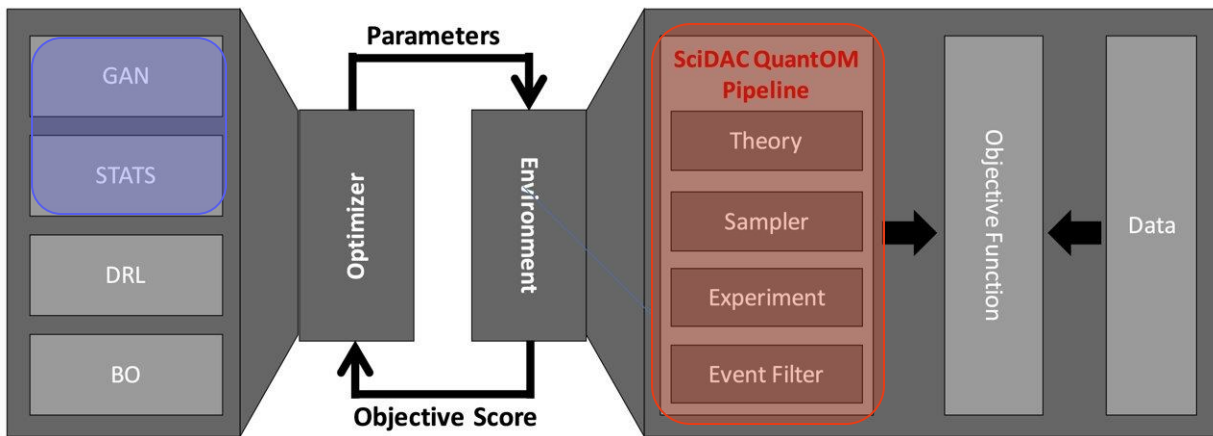
# SAGIPS: Scalable Asynchronous Generative Inverse Problem Solver



- Modular framework
- Adjustable pipeline translates predictions to physics events
- Leverage input from theory and experiment
- Currently 2 optimization approaches available
  1. Statistical workflow
  2. GAN workflow (today's focus)

**Domain Science**

# SAGIPS: Scalable Asynchronous Generative Inverse Problem Solver



- Modular framework
- Adjustable pipeline translates predictions to physics events
- Leverage input from theory and experiment
- Currently 2 optimization approaches available

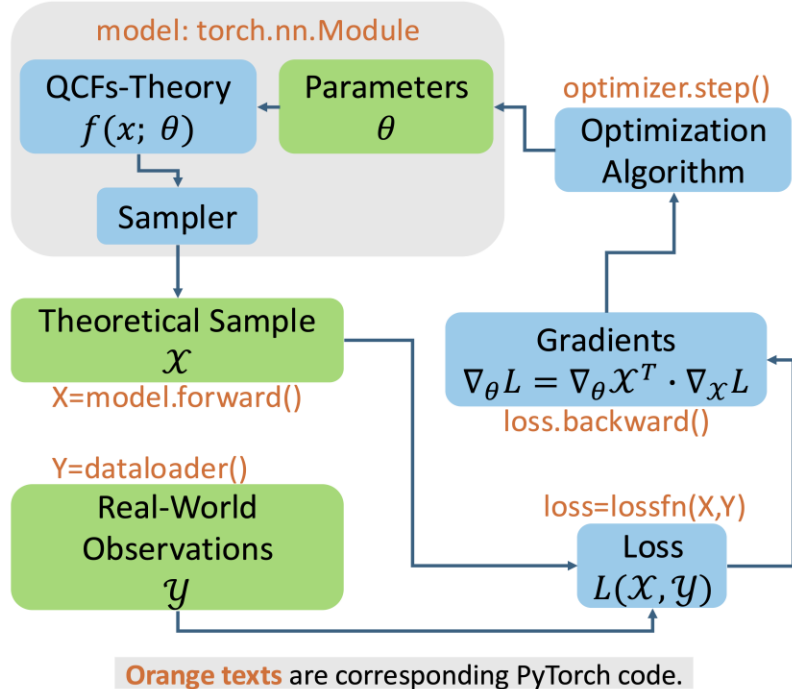
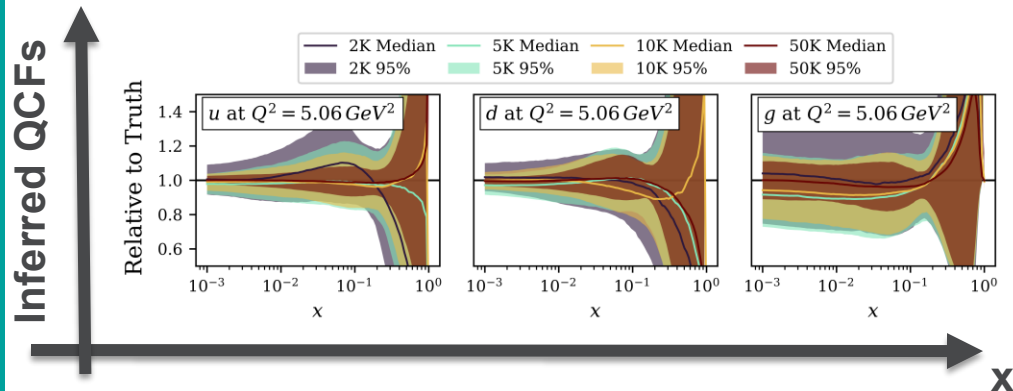
**Domain Science**

1. Statistical workflow
2. GAN workflow (today's focus)

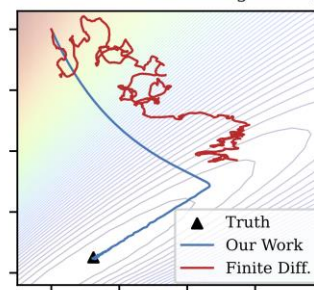
**Data Science**

# Statistical Workflow

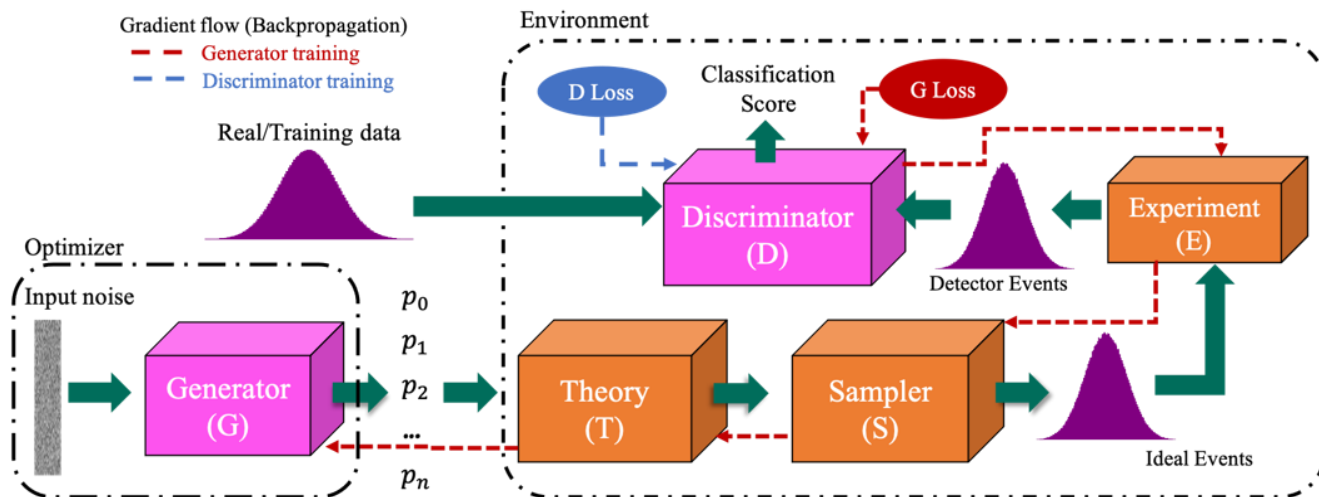
- Pi-Yueh Chuang (Argonne) et al.
- Determine gradients for non-differentiable components: **DistroSA** (Distribution Sensitivity Analysis)
- Support any event sampling routine
- First tests on proxy problem
- GPU accelerated loss function



Loss Surface & Training History



# GAN Workflow

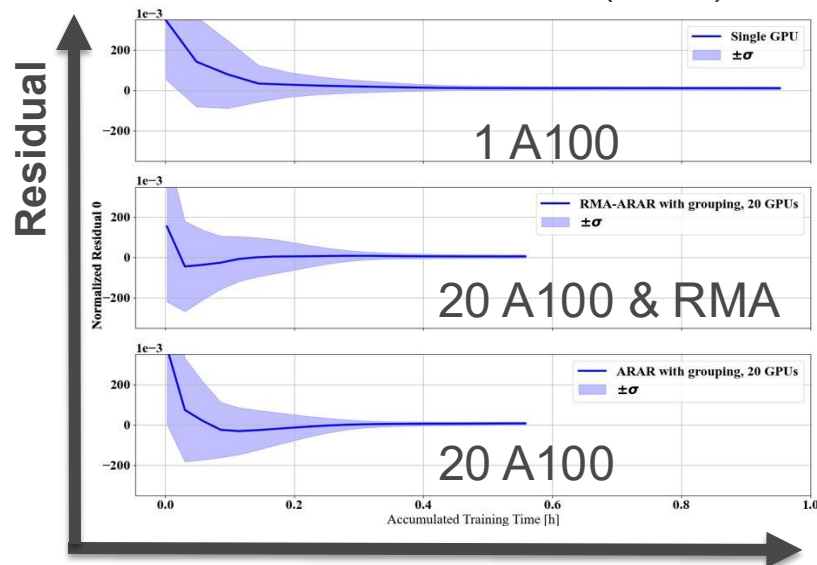
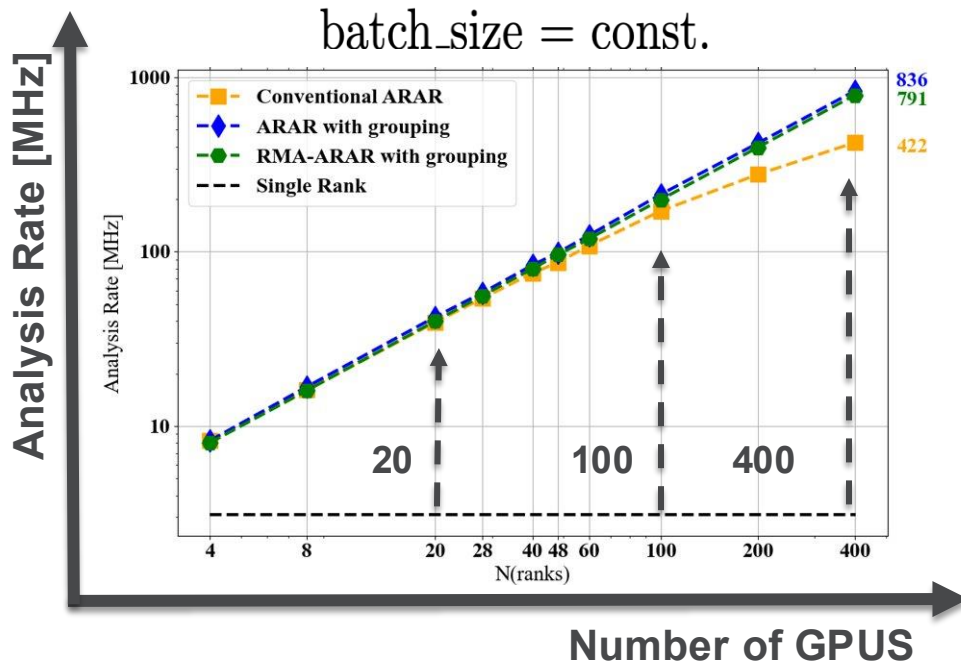


- All components need to be differentiable
- Use proxy problems to evaluate workflow
- 3 evaluation phases

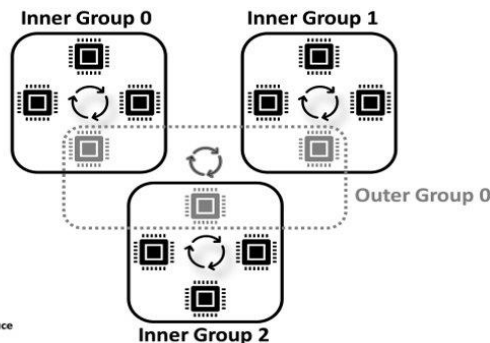
Phase	Proxy Problem Dimensionality	Complexity
1	1D	low
2	2D	mid
3	5D	high

# Phase 1: Distributed Analysis on Polaris

$$\text{batch\_size} \propto \frac{1}{N(\text{GPUs})}$$

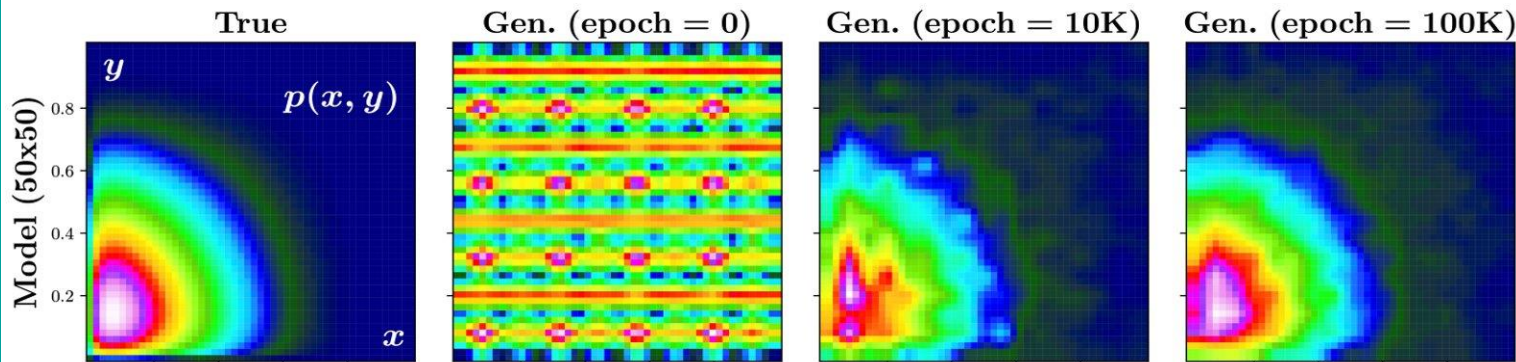
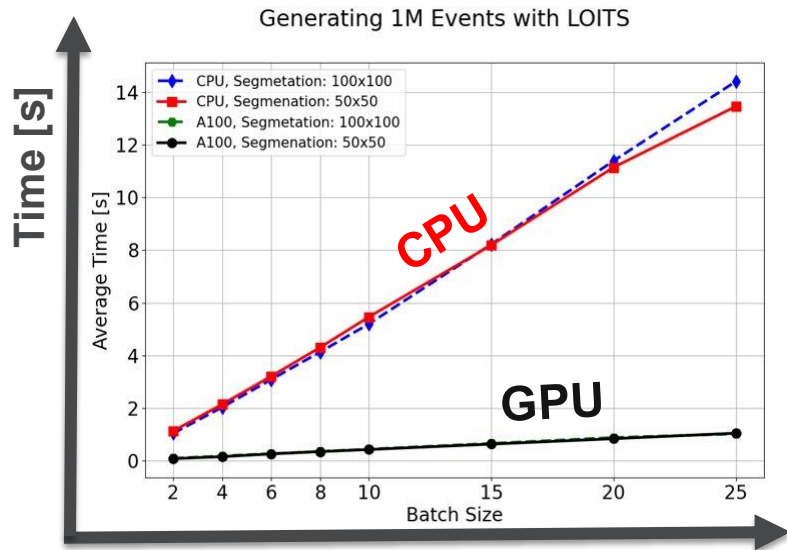
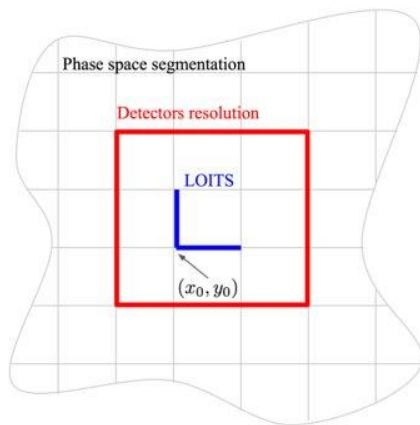


- Analyzed 1D proxy problem:  $x \sim \rho(x, p_0, \dots, p_5)$
- Share generator gradients across GPUs
- Asynchronous ring communication w/ grouping mechanism
- Optional: Remote Memory Access
- Published in: IOPScience: Machine Learning: Science and Technology



# Differential Sampling in 2D

- Local Orthogonal Inverse Transform Sampler – **LOITS**
- Provide gradients for backpropagation
- Optimal performance on GPU
- Tested on 2D proxy problem:  
 $(x, y) \sim \rho(x, y)$
- Submitted to Physics Letter B



Batch Size

Training Progress

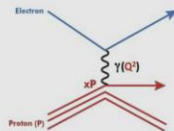
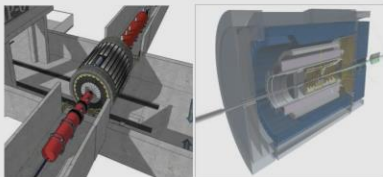
# Multi-Dataset Analysis

N Datasets from different  
DIS Experiments

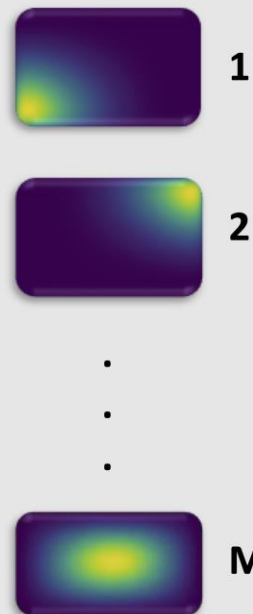
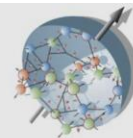
Jefferson Lab  
Thomas Jefferson National Accelerator Facility



EIC<sup>2</sup>



M ≤ N Density  
Functions

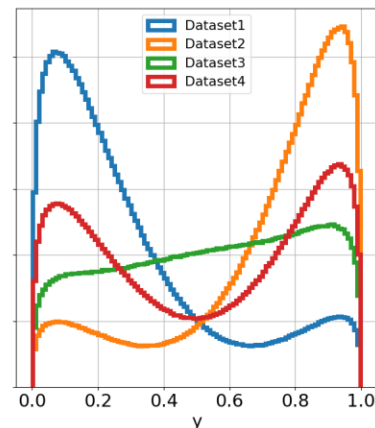
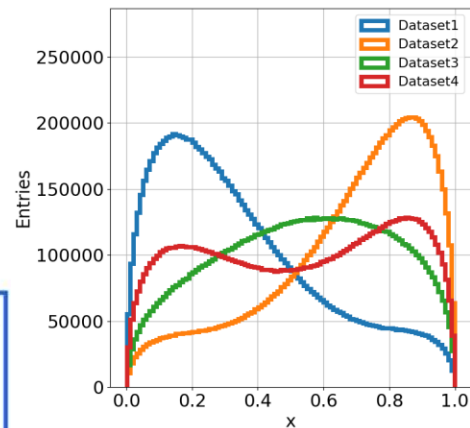
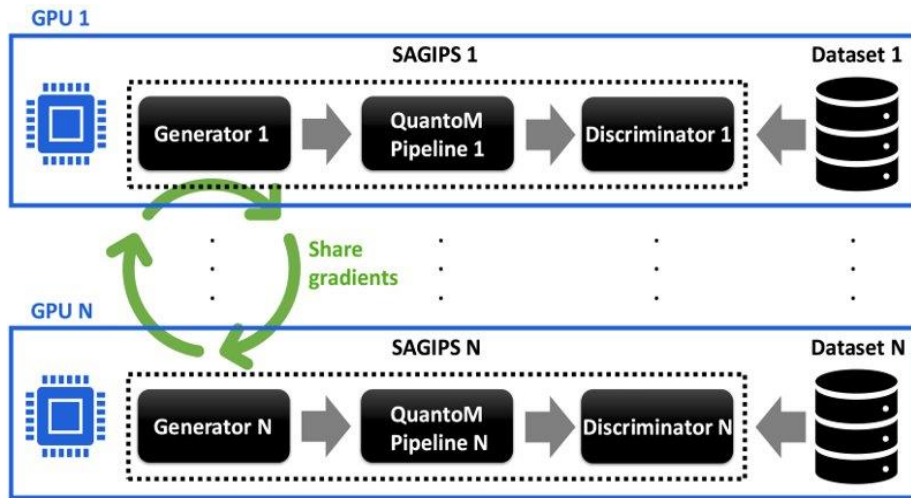


# Phase 2: Multi-Dataset Analysis on 2D Proxy Problem

- Simultaneously fit 4 datasets and extracted 3 densities

$$(x_i, y_i) \sim \sum_{j=1}^3 \left[ m_{ij} \cdot \rho_j(x, y) \right], i = 1, \dots, 4$$

- Apply distributed data parallel technique

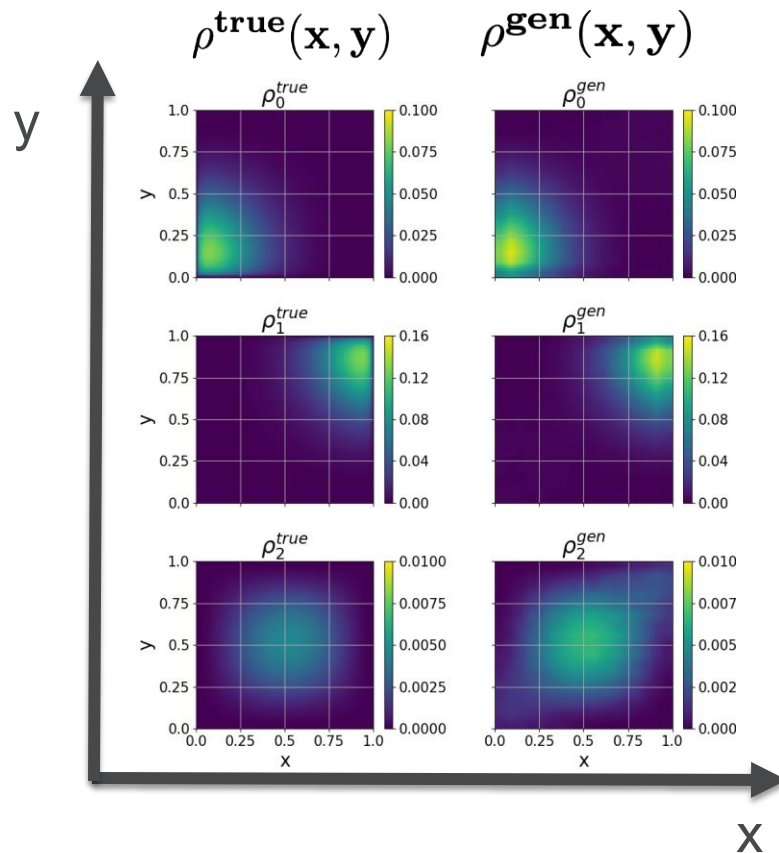
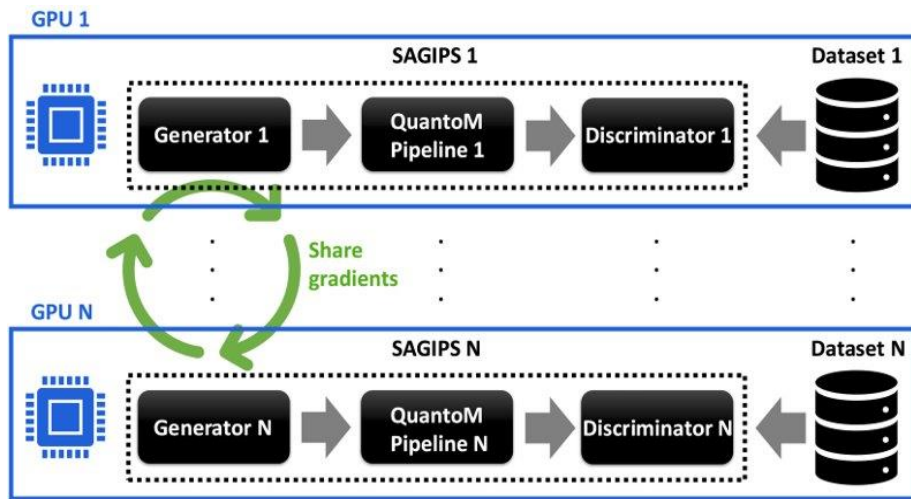


# Phase 2: Multi-Dataset Analysis on 2D Proxy Problem

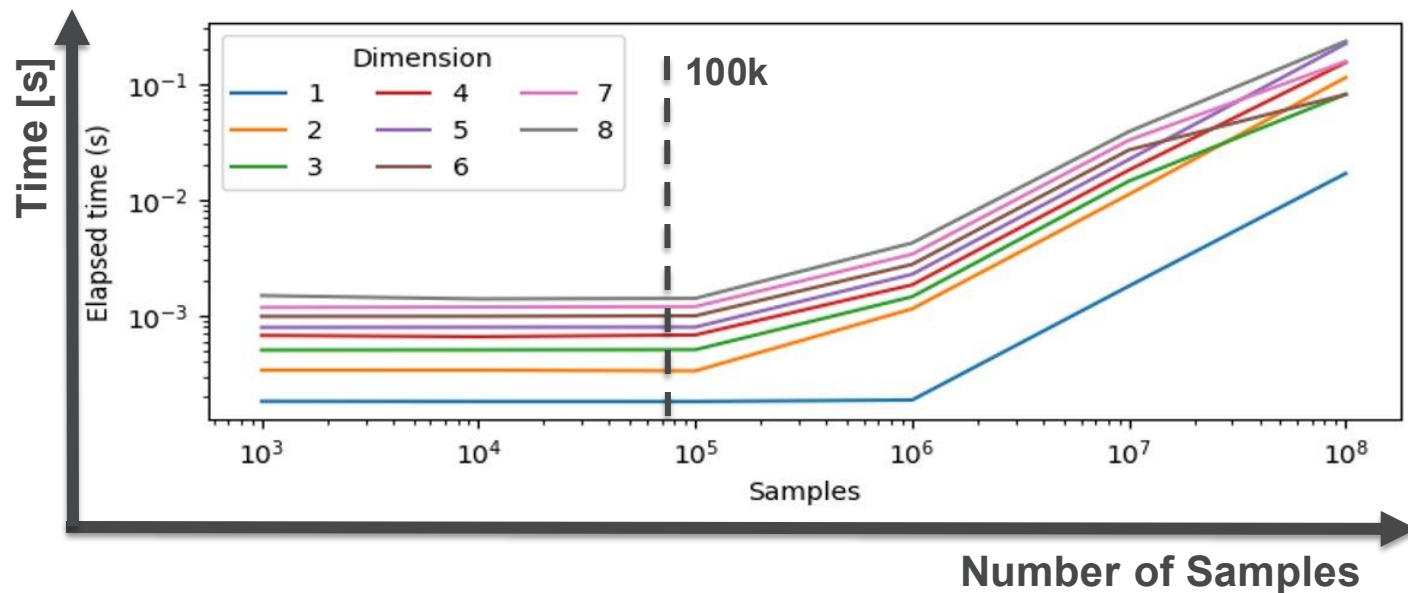
- Simultaneously fit 4 datasets and extracted 3 densities

$$(x_i, y_i) \sim \sum_{j=1}^3 \left[ m_{ij} \cdot \rho_j(x, y) \right], i = 1, \dots, 4$$

- Apply distributed data parallel technique
- Utilized 160 Polaris GPUs
- Detailed evaluation ongoing



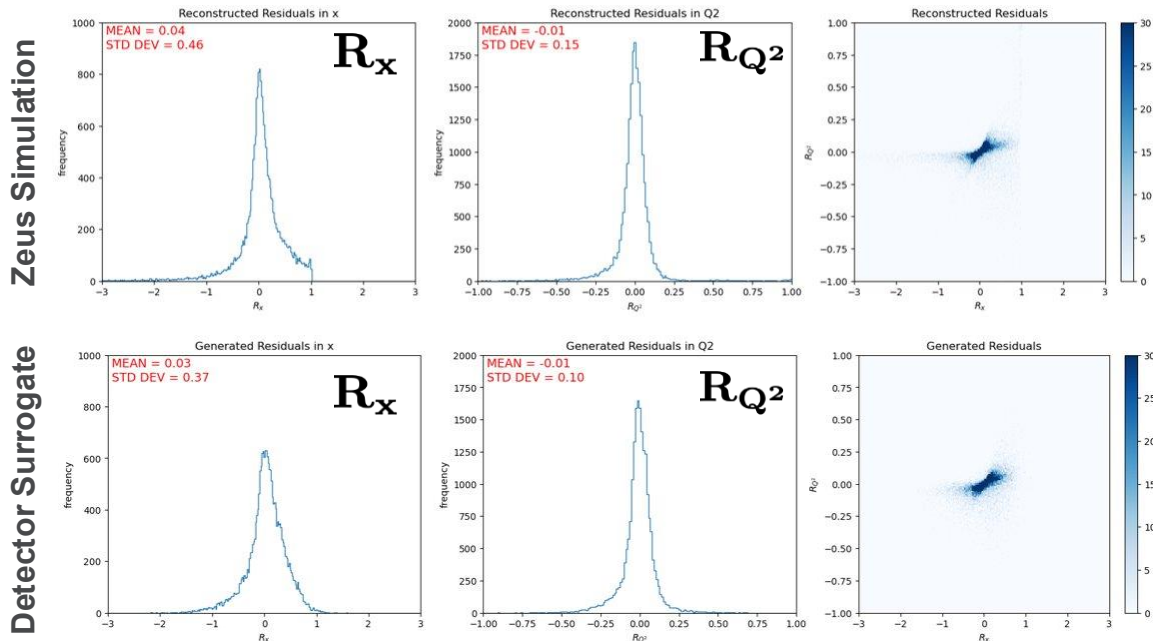
# Towards Phase 3 with N-D Sampling



- Extension of LOITS to run in 5D
- Dimensional scaling tests on A800
- Observe nearly linear scaling
- Integrate N-D sampler into workflow and run performance tests

# Experimental Effects

- Events generated by SAGIPS need to match experimental data
- Include experimental effects
  - Accelerator and physics backgrounds
  - Detector misalignment and inefficiencies
  - Acceptance and resolution effects
  - Particle misidentification
- Full simulation chain not applicable --> Use surrogate instead
- Detector modeling mostly finished
- Work on background handling ongoing



- Surrogate generates residuals:  $R_x = \frac{x_0 - x'}{x_0}$ ,  $x_0$ : ideal,  $x'$ : after detector
- Apply resolution effects (folding):  $x \mapsto x \cdot (1 - R_x)$
- First test: Trained VAE detector surrogate on Zeus simulations

# Summary & Outlook

- 3D Imaging of nuclei through simulation based inference
- Joint theoretical-experimental analysis of QCFs
- Generative AI for inverse problem solving
  - Modular workflow
  - Asynchronous ring communication
  - Deployed on Polaris HPC system
  - Differentiable sampling
  - Simultaneous fit of multiple data sources
- Integrate uncertainty quantification
- Work on detector surrogate and background modeling ongoing
- Run analysis on real experimental data

# BACKUP SLIDES



U.S. DEPARTMENT  
of ENERGY



QuantOm  
Collaboration