



Jefferson
Lab

AID(2)E: AI-Assisted Detector Design for the EIC

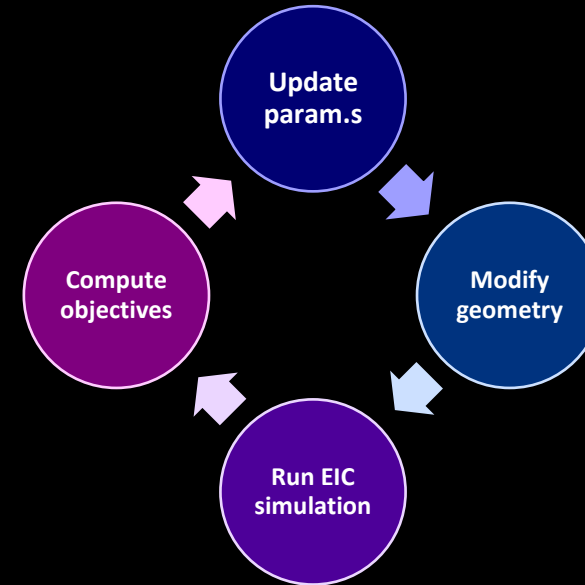
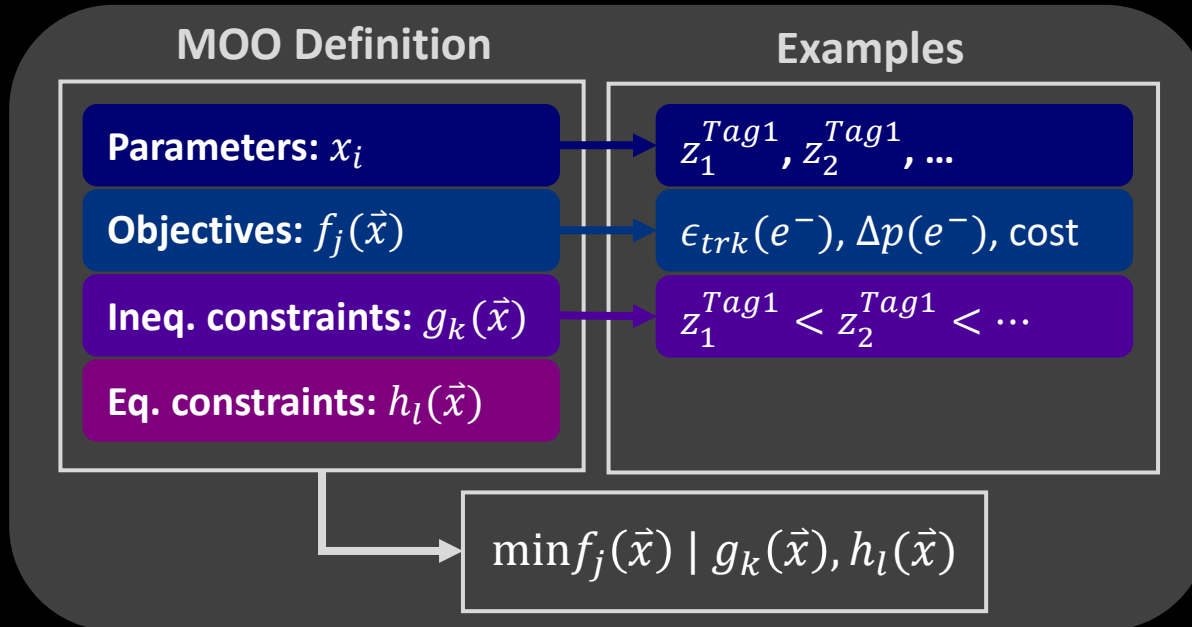
2026 EIC UGM, AI Town Hall

Derek Anderson

07.16.2026



Method | Detector Optimization as MOO



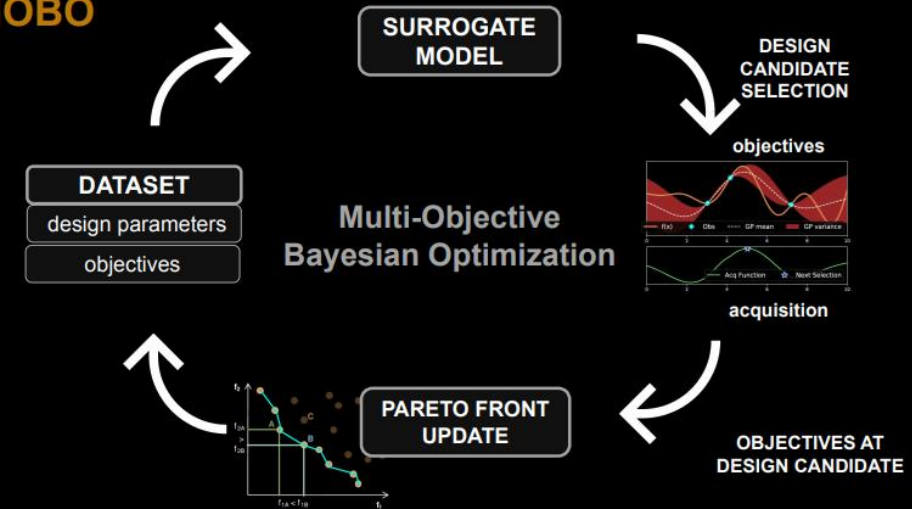
- Detector design can be formalized as a **Multi-Objective Optimization (MOO)** problem:
 - Have a set of objectives f_j that are functions of parameters $\vec{x}$ subject to constraints g_k and h_l
 - Map onto **physically-motivated observables (f_j)** and **parameters of detector ($\vec{x}$)**
- Goal is to minimize (maximize) objectives
 - ☞ **But how do you do that *effectively*?**
- For (very) small no. of $\vec{x}$, could do grid search... But:
 - 1) EIC simulations can be *very* heavy;
 - 2) And what if you have *many* $\vec{x}$ and/or complex trade-offs b/n f_j ?

AI/ML | AID(2)E Project

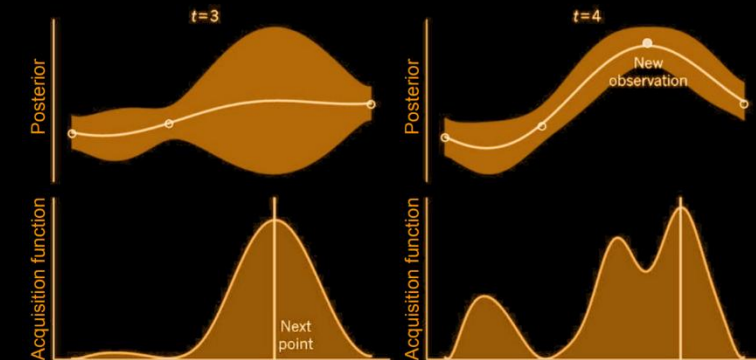


- **AI/ML can help!** Specifically *Bayesian* optimization techniques (MOBO)
 - MOBO guides optimization by identifying most-promising areas of design space to focus on
 - 👉 **Right:** illustration of a typical MOBO algorithm
 - Frameworks like [Ax](#) and [Pymoo](#) offer robust libraries for MOBO and Genetic Algorithms (MOGA)
- **AID(2)E: AI-Assisted Detector Design for the EIC**
 - Collaboration focused on applying and spreading AI-assisted MOO techniques to the EIC (and beyond)
 - See 2024 whitepaper [\[JINST 19 C07001 \(2024\)\]](#)

MOBO



With a single objective, for illustration



1. Select a Sample by Optimizing the Acquisition Function.
2. Evaluate the Sample With the Objective Function.
3. Update the Data and, in turn, the Surrogate Function.
4. Go To 1.

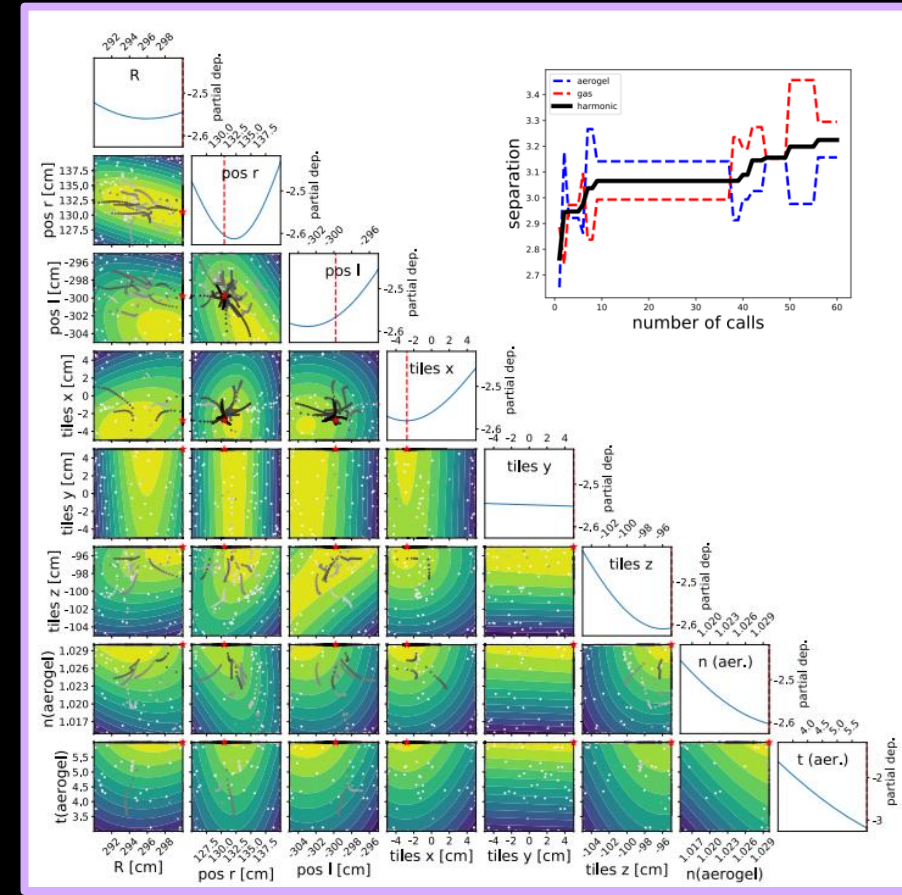
[AID\(2\)E, JINST 19 C07001 \(2024\)](#)

[C Fanelli, RHIC/AGS AUM \(2024\)](#)

Examples | dRICH, B0, and more MOBOs

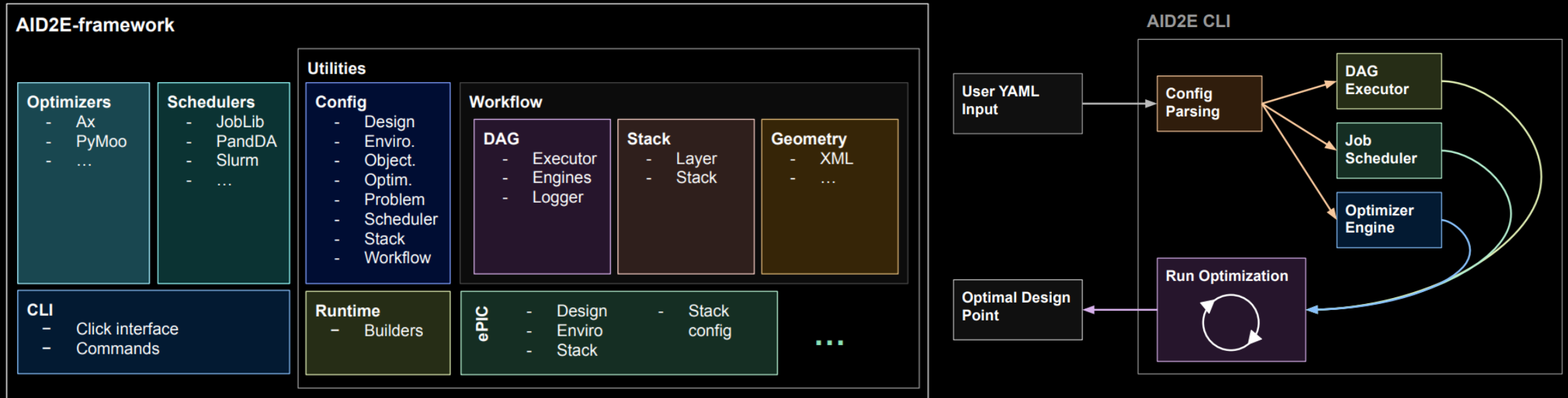


- **dRICH-MOBO** [[JINST 15 P05009 \(2020\)](#)]
 - Simultaneous optimization of dRICH mirror radius/position, tile positions, and aerogel properties
 - Used harmonic mean $\pi/K/p$ separation in aerogel and gas
 - **Right:** posterior distributions of parameters (bottom left) and objectives vs. trial no. (upper right)
 - ☞ Points indicate values selected during trials
- **B0 geometry currently being optimized by Baptiste Fraisse et al**
 - Simultaneously optimizing tracker disc positions and active region for θ RMSE
 - ☞ See [the recent technical note](#) on Zenodo!
- **And:** Simon and I are exploring using MOBO techniques for optimizing the Low-Q² tagger



E. Cisbani et al, JINST 15 P05009 (2020)

Current Targets | Framework and Paperwork



- **A key AID(2)E deliverable:** *turnkey* framework to do optimizations
 - Provide a simple, unified interface to HPC/HTC resources, optimization engines, and experiment software
 - **Above:** resource diagram for framework (**left**) and intended workflow (**right**)
- **Two primary targets for this year:**
 - a) 1st implementation of framework
 - b) Paper illustrating computational performance of framework
- **Note:** subsequent framework release will *also* support optimization of *reconstruction & analysis* parameters

Outlook

- Effectiveness of detector design as MOO demonstrated in several examples, closure tests
- **Still room for optimization at the EIC!**
 - › Some designs are still pending, some will change
 - › Reconstruction will continue to evolve
 - › Detector 2 under active development
- Working towards 1st release of AID(2)E framework and accompanying computation paper
- Ongoing effort to explore use of LLMs and agentic techniques in AID(2)E context
 - › [\[arXiv:2603.30014\]](https://arxiv.org/abs/2603.30014) Discussion on using agents with PanDA-iDDS for AID(2)E

AID(2)E

- › Amit Bashyal (BNL)
- › Markus Diefenthaler (JLab)
- › Cristiano Fanelli (W&M)
- › Baptiste Fraisse (CUA)
- › James Giroux (W&M)
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- › Karthik Suresh (W&M)
- › Anselm Vossen (Duke)
- › Fang-Ying Tsai (Duke)
- › Tianle Wang (BNL)
- › Torre Weanus (BNL)