

LUME Ecosystem Updates: Standardized Virtual Accelerators and Digital Twin Infrastructure

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July 16, 2026

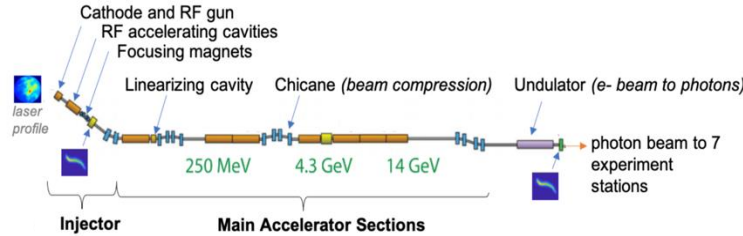
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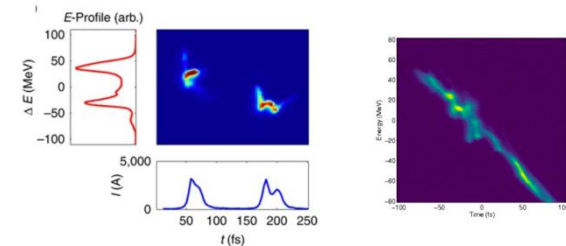
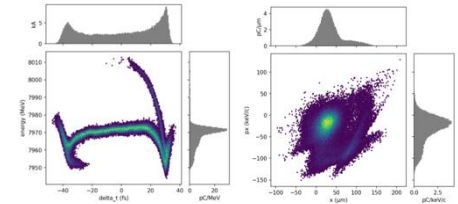
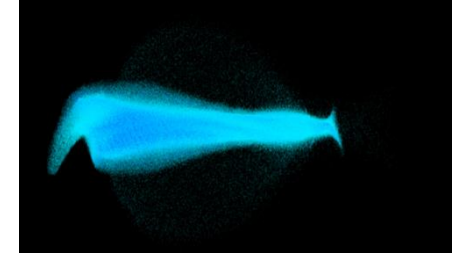
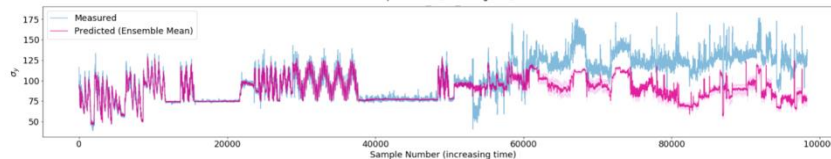
Overview

We Need Detailed & Robust System Models

Many tuning problems require detailed beam phase space customization for different experiments



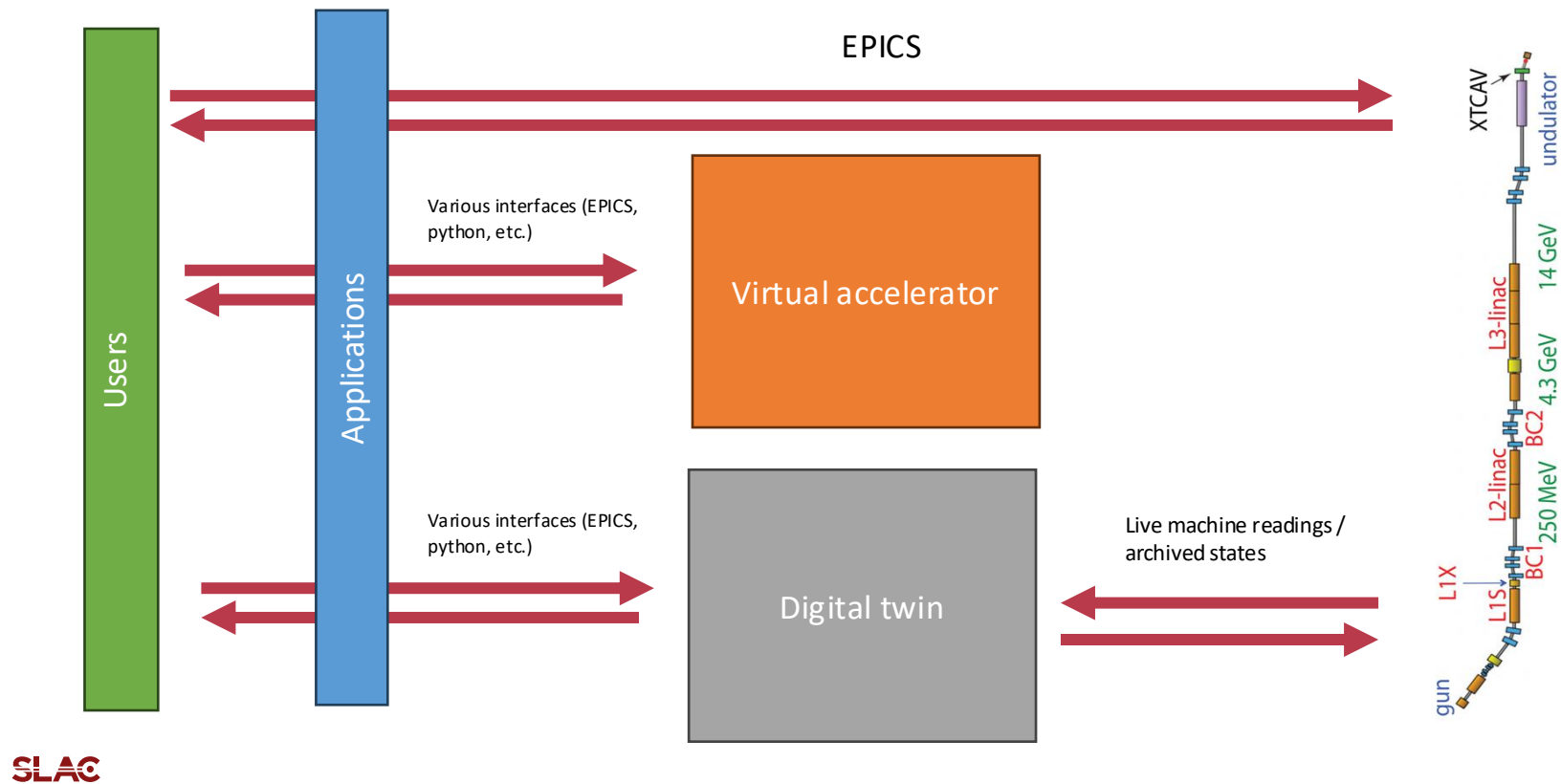
- Nonlinear, high-dimensional optimization/control problem with incomplete information
- Digital twins and virtual accelerators can provide insight
- Infrastructure is a challenge
- Reliable model adaptation methods are key for putting online models to use operationally



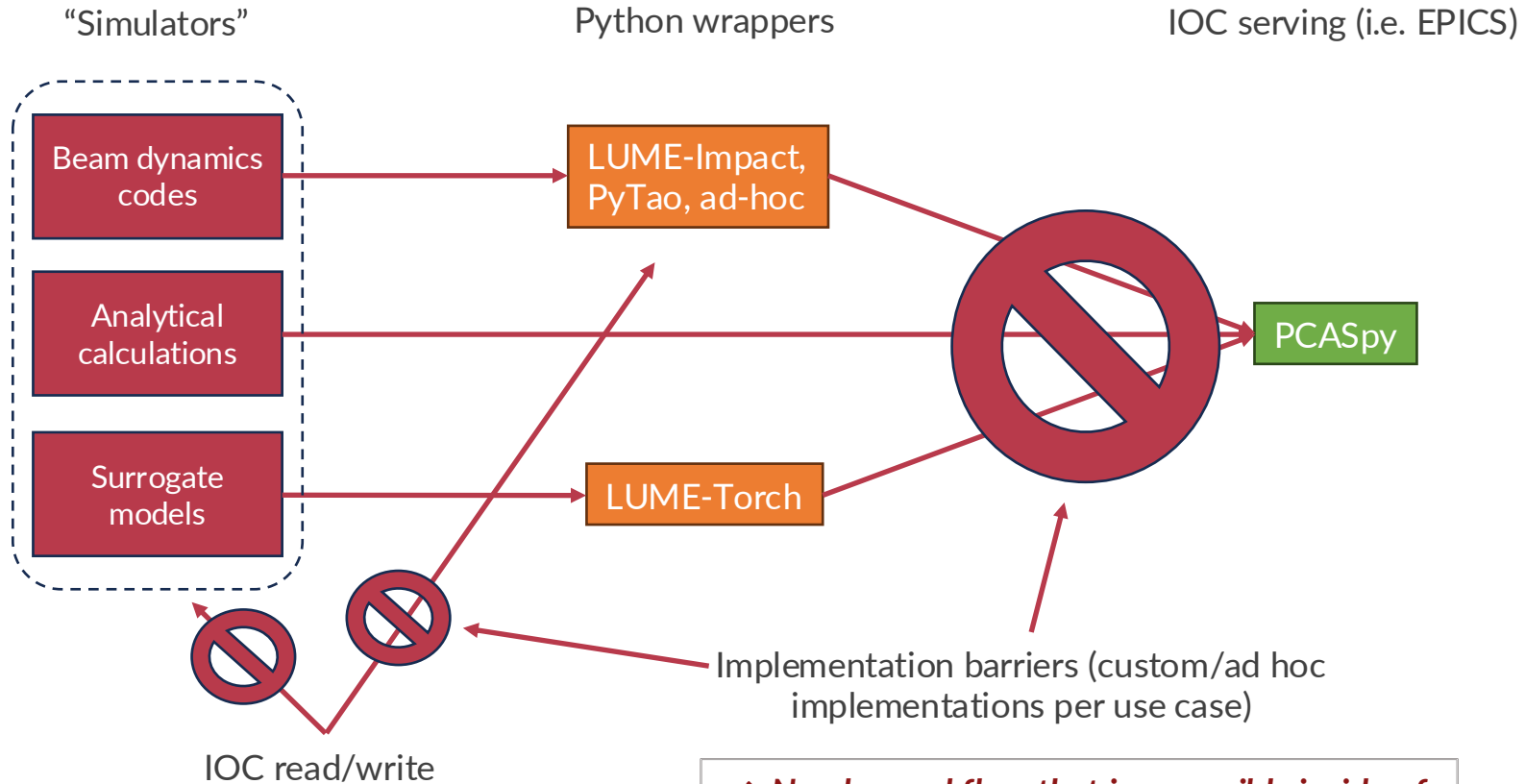
A. Marinelli, et al., Nat. Commun. 6, 63 69 (2015)

A. Marinelli, IPAC'18

Nomenclature

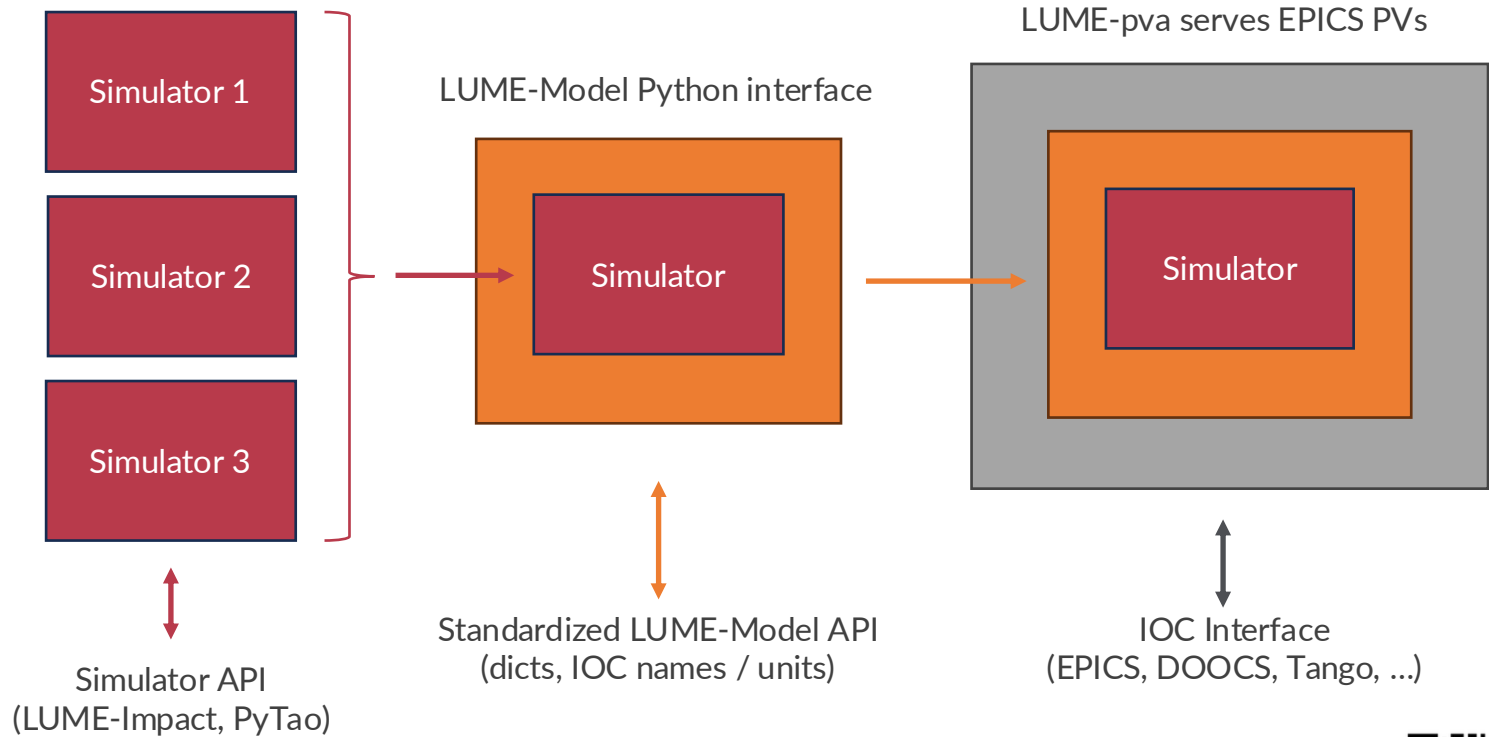


Combining and Deploying Models is Challenging



LUME for Virtual Accelerator and Digital Twin Implementations

A Unified and Standardized API with LUME



Standard Interface for VA models and Digital Twins

LUME-Base
- LUMEModel

set

Set control variables by name and run the sim/model, then update the model's state

get

Read back model state by variable name

reset

Restore model to initial state

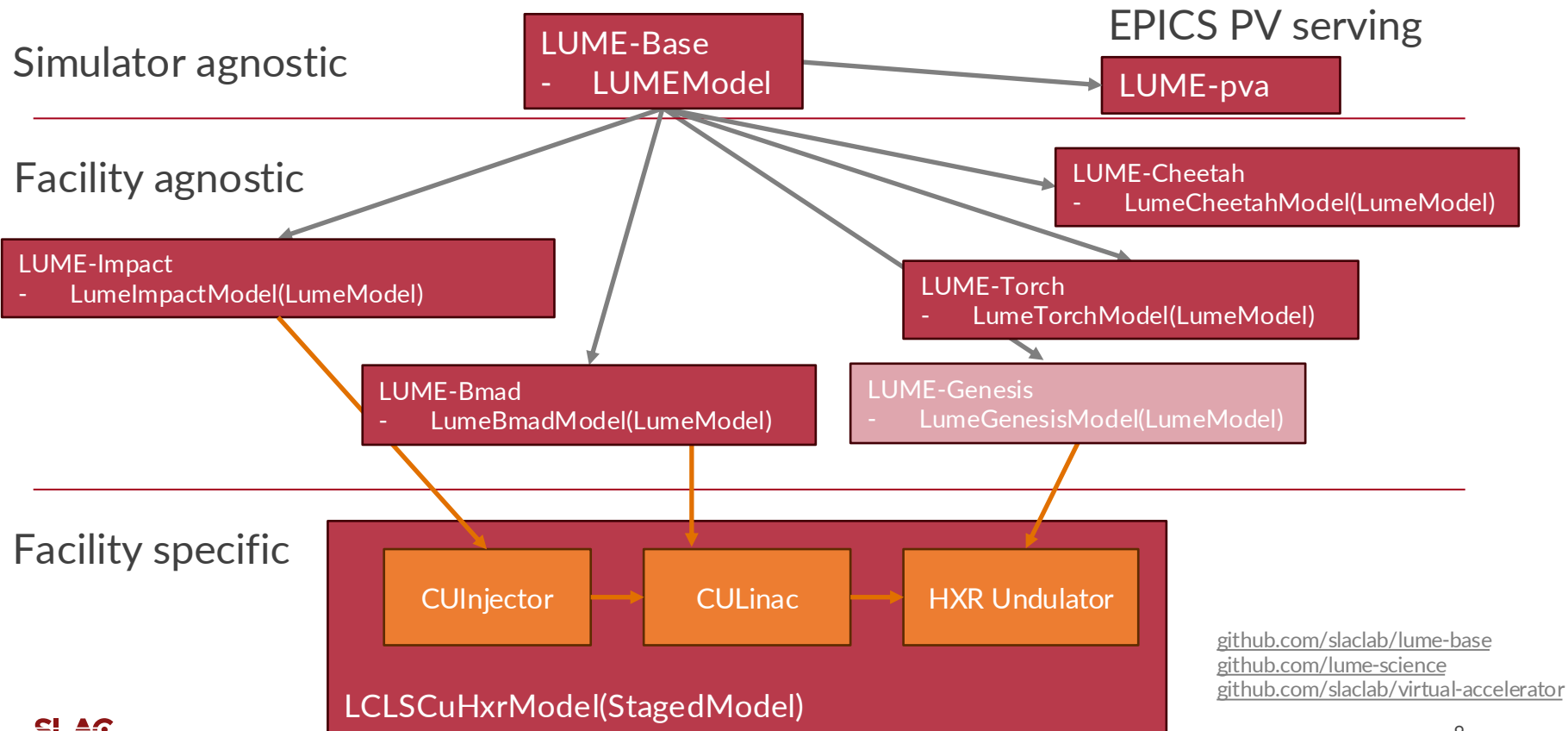
supported_variables

Dictionary of named variables that the model supports

*any simulator
can implement
this contract*

*Utilized by IOC server (e.g. LUME-pva) to create PVs –
can be extended to other
control system types*

LUME Package Structure



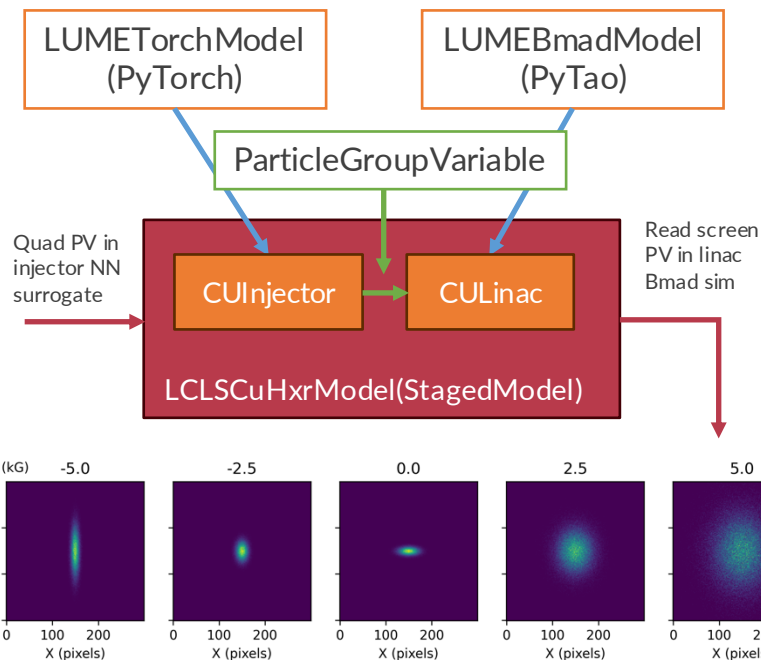
Staged Model Example: ML Surrogate with Physics Simulation

```
injector_surrogate = InjectorSurrogate()  
cu_hxr_bmad_model = get_cu_hxr_bmad_model()  
  
staged_model = StagedModel([injector_surrogate, cu_hxr_bmad_model])
```

Set PV in injector NN surrogate

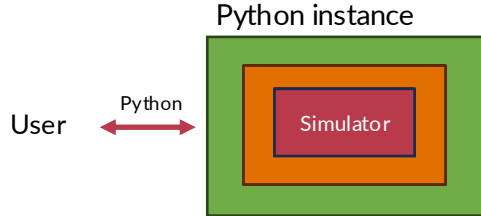
```
staged_model.set({SCAN_QUAD_PV: quad_value})  
image = staged_model.get(OTR_IMAGE_PV)
```

Read screen PV in linac Bmad sim

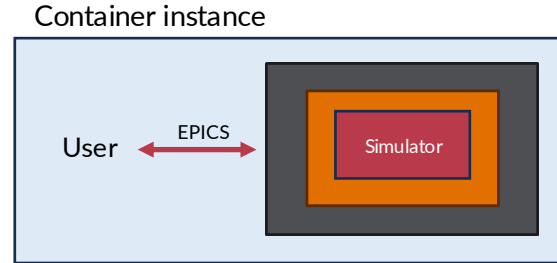


Virtual Accelerator and Digital Twin Implementations

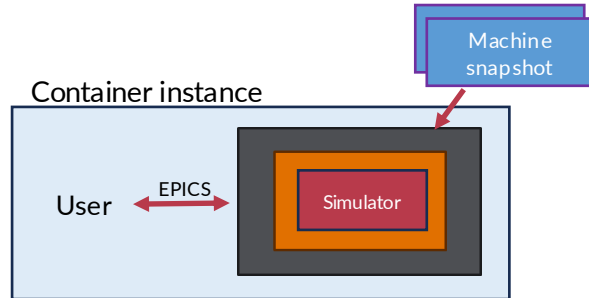
Virtual accelerator in Python



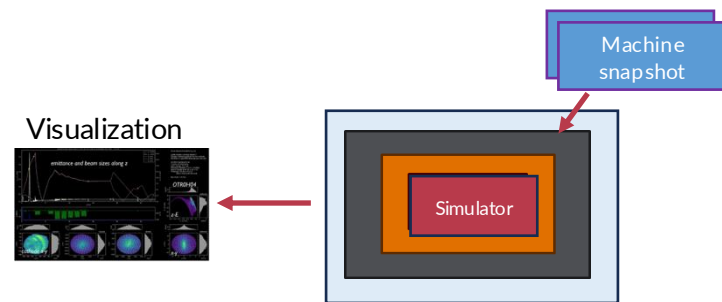
Virtual accelerator in container



Digital twin in container



Digital twin in global server



Badger Environment Testing / Continuous Integration

Use VA to perform nightly testing / validation of code used to interact with the machine

Badger GUI



LCLS Badger Environment class



Class instance

Nightly testing workflow



Testing report



EPICS

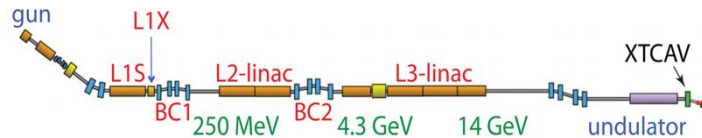


not available for software testing

EPICS

Simulator

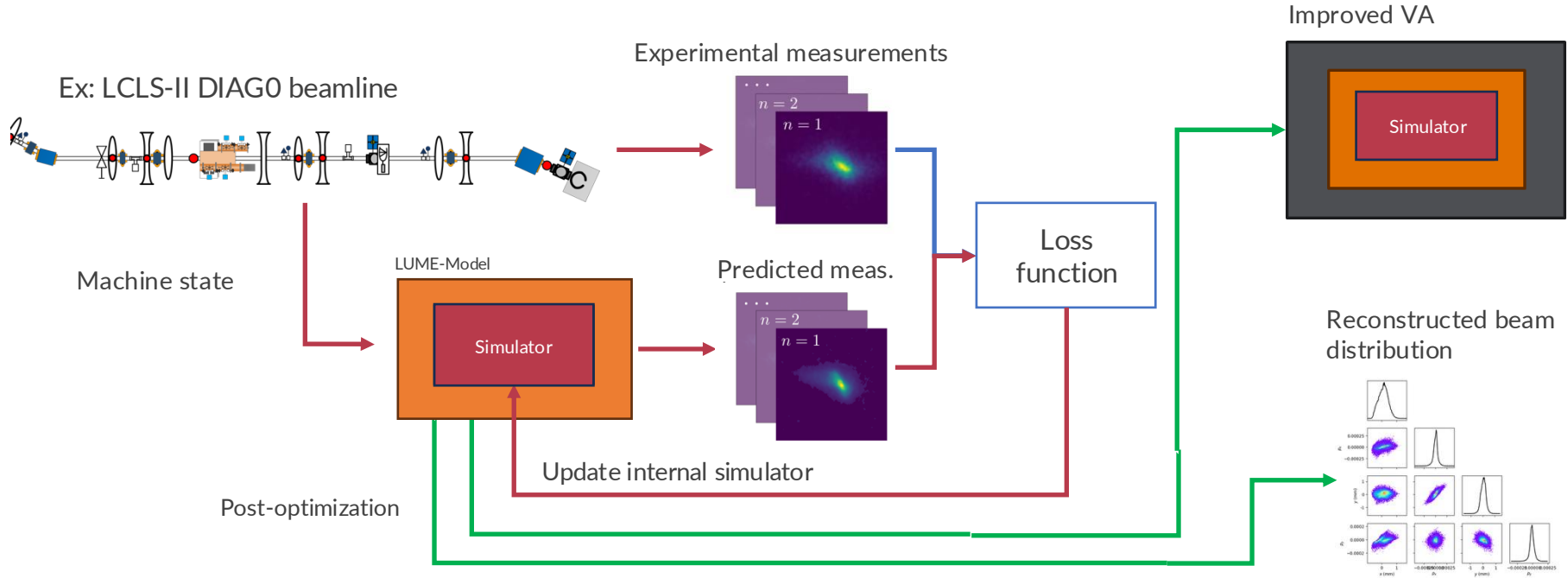
LCLS Virtual Accelerator



LCLS Machine

VA for Model Calibration / Phase Space Measurement

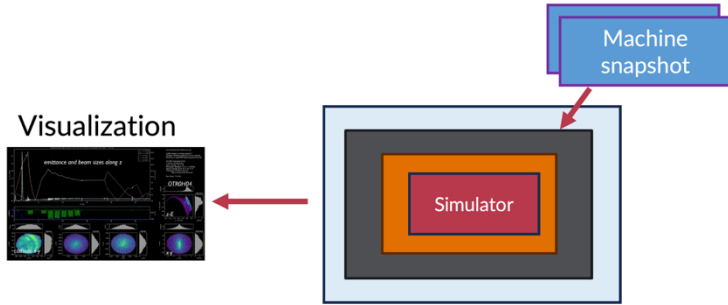
Black-box (Impact / Bmad) and differentiable simulators (Cheetah, surrogates) can be used inside VAs to perform model calibration, phase space reconstruction



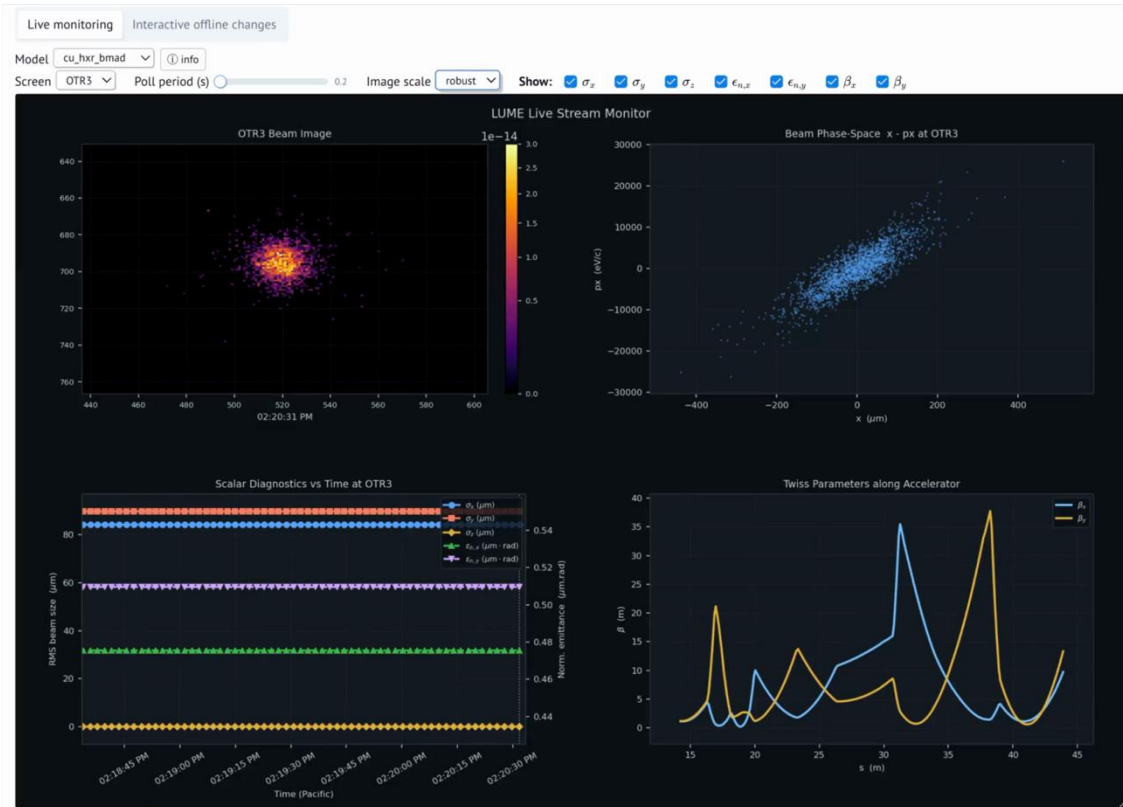
Digital Twin Live Deployment

Digital twin in global server

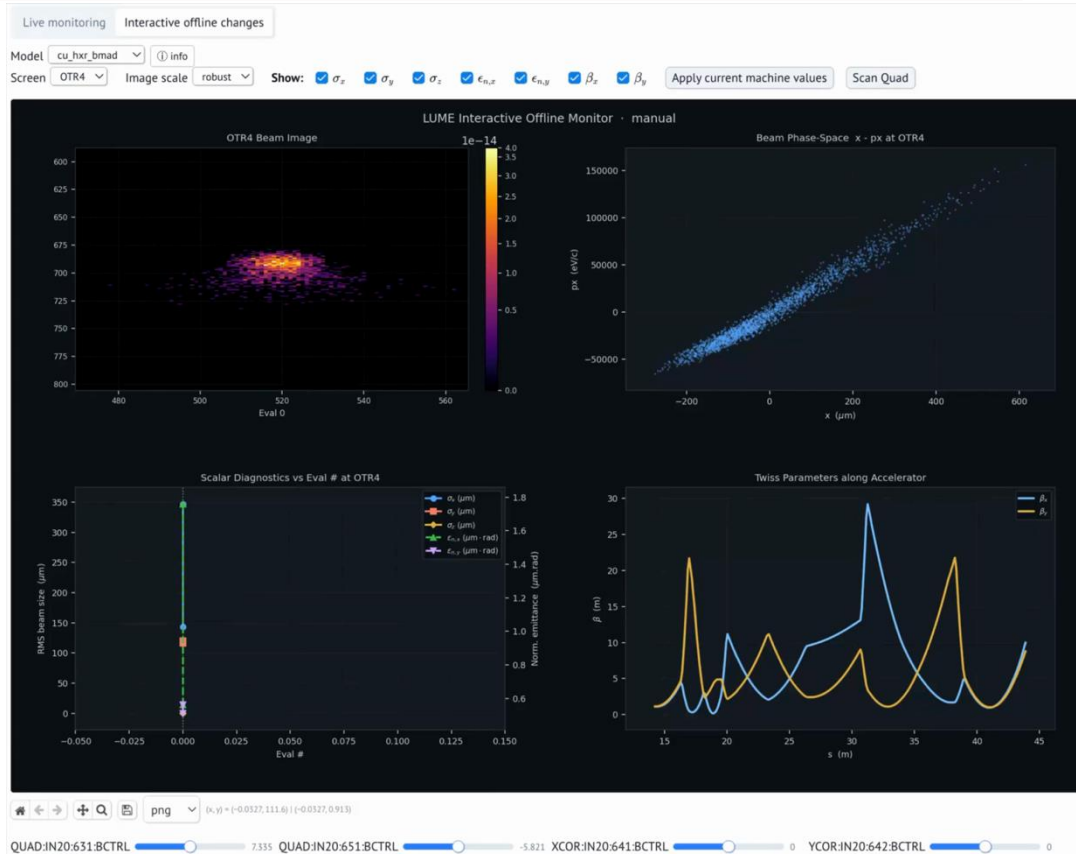
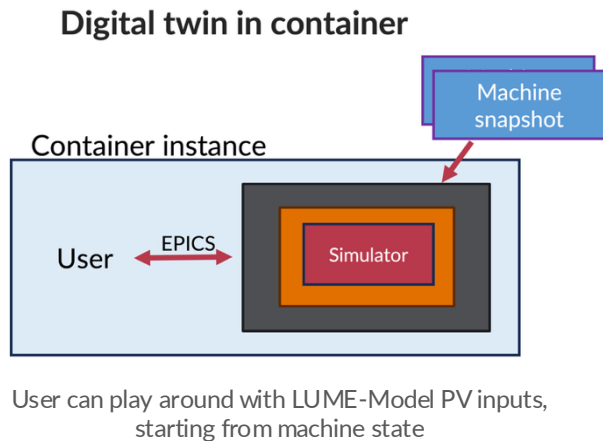
Visualization



Continuously updating LUME-Model state from live PV snapshots

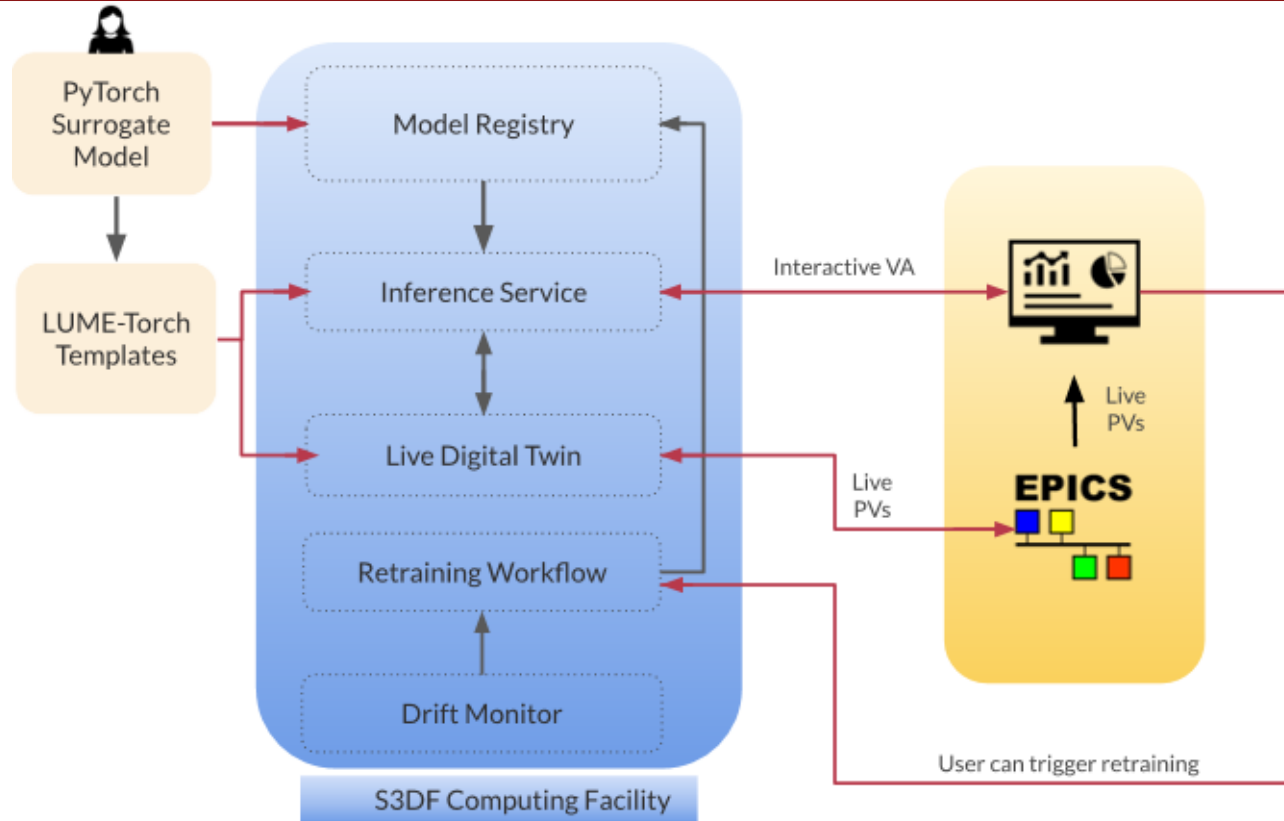


Interactive VA Mode from Live State

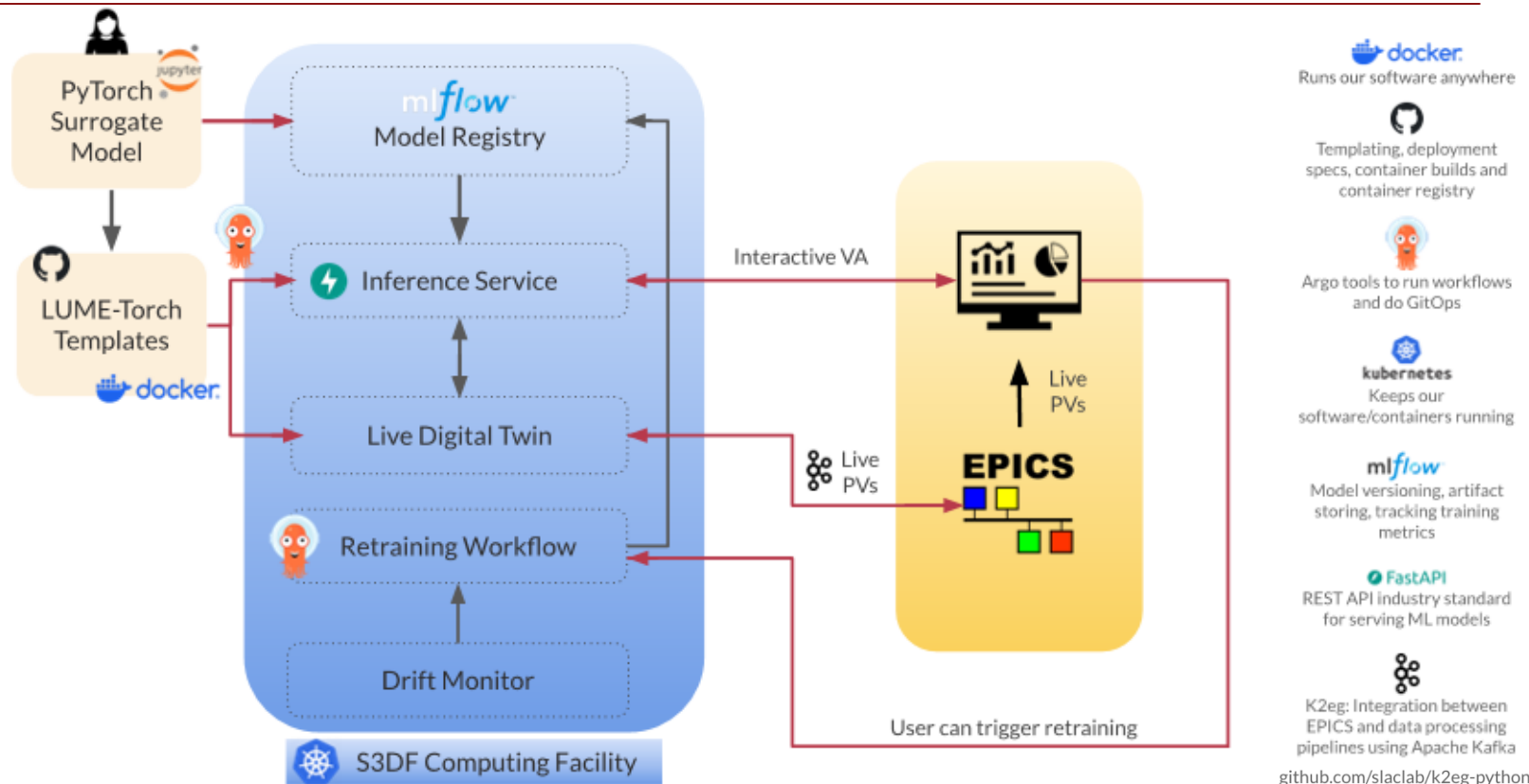


Surrogate Model Deployment and Lifecycle

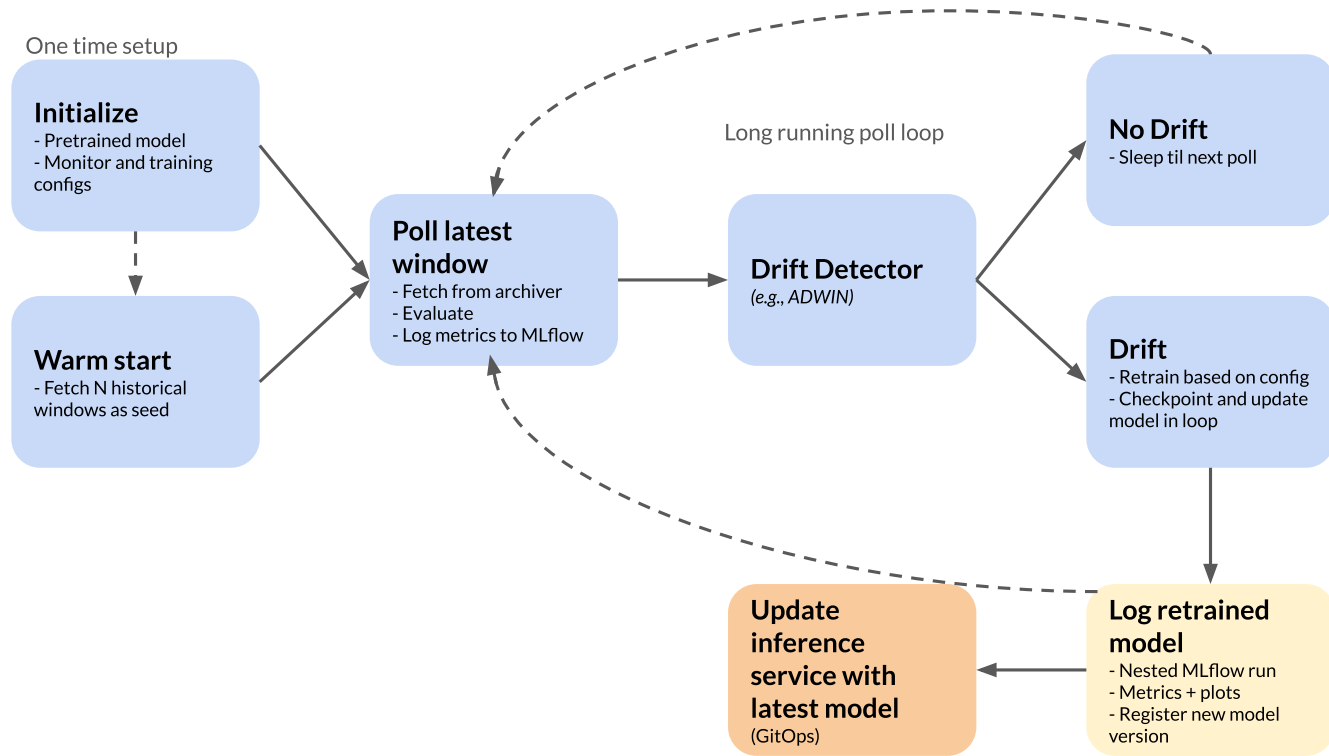
Digital Twin Deployment of Surrogate Models



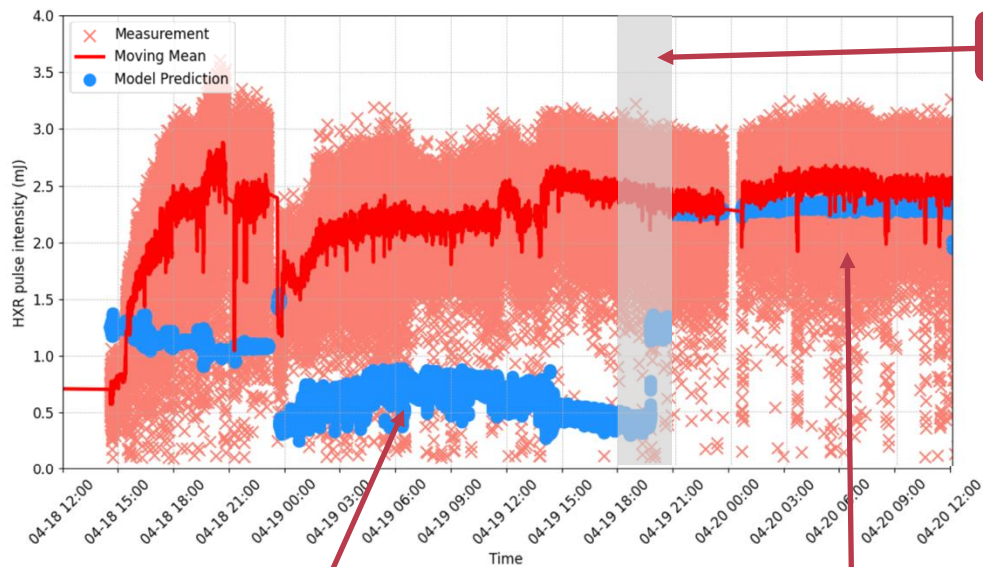
Digital Twin Deployment of Surrogate Models



Continual Learning: Drift Monitor Implementation



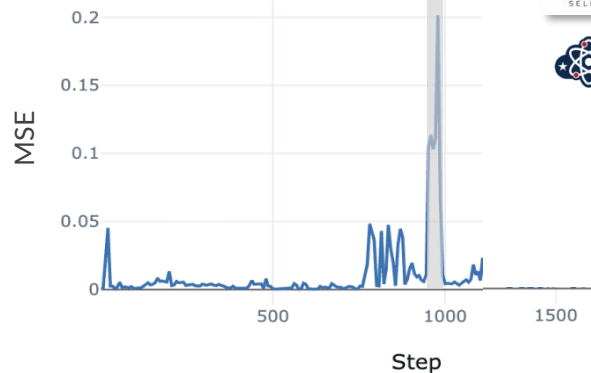
Continual Learning Demo for FEL Surrogate on Historical Data



Making progress on improving
FEL model accuracy alongside
CL workflow development

APEIRON keeps the model
synced with data during regime
changes

Drift
Detected

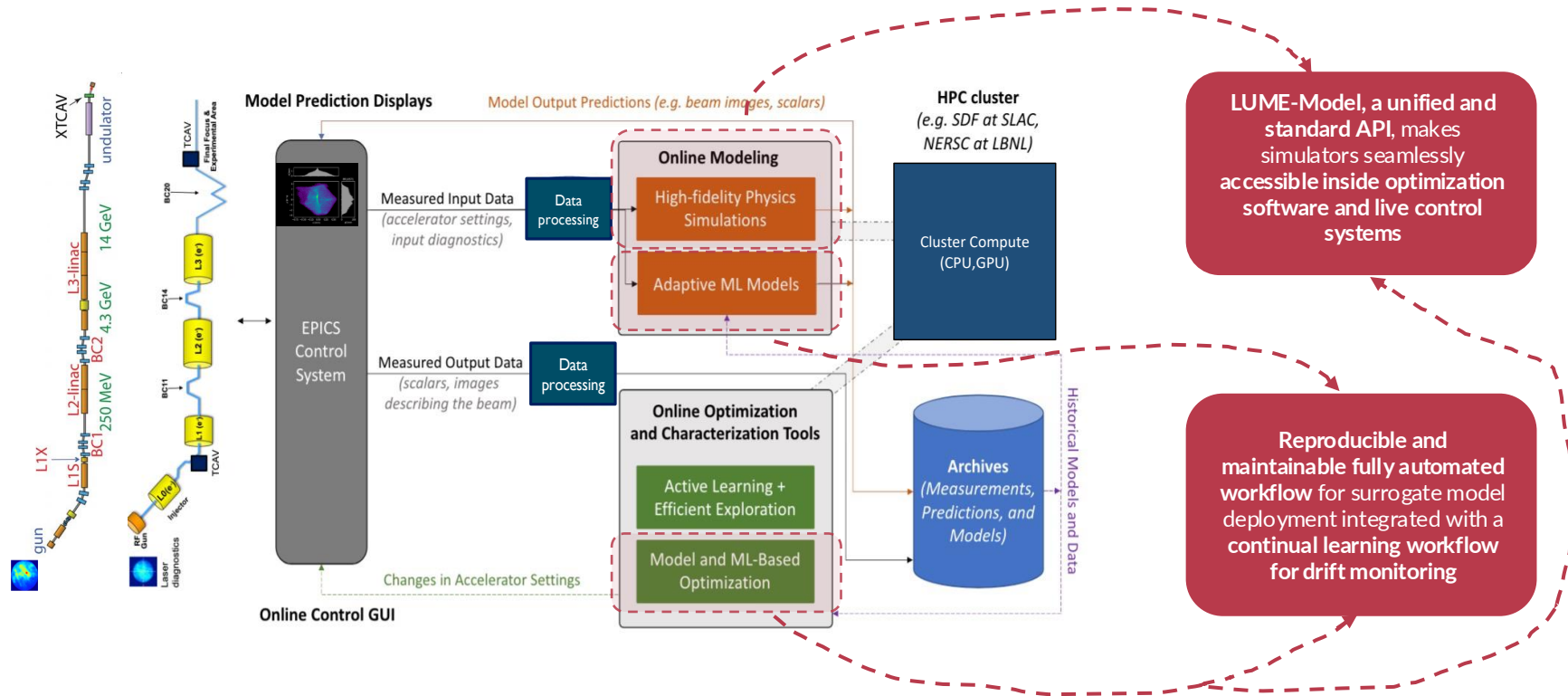


All drift monitoring and training metrics
are logged to AmSC MLflow



Summary and Next Steps

Overcoming Challenges to Enable DT & VA Integrations at Particle Accelerators



Next Steps

Short Term

- Continuing to improve LCLS virtual accelerators and deploying on-demand VA service at SLAC
- Expanding LCLS staged model to include FEL surrogate model after Bmad linac stage
- Expand VA capabilities in terms of the underlying physics sims (variables/physical processes)
- General improvements to accuracy of surrogate models
- Improvements on drift monitoring and training workflows

Long Term

- Calibration/training support in LUME-Torch
- Infrastructure for continual calibration of system models/sim2real workflow
- Operator training/adoption of tools
- Linking system models to Osprey Agentic workflow (github.com/als-apg/osprey)
- Adoption of MLDP (ML Data Platform): enables metadata tagging for easier integration with CL/training workflows
 - github.com/osprey-dcs/data-platform
 - [doi: 10.18429/JACoW-ICALEPCS2025-WEPD083](https://doi.org/10.18429/JACoW-ICALEPCS2025-WEPD083)

Thank you and resources



Thanks to our amazing colleagues

Gopika Bhardwaj, Ryan Roussel, Ernest Williams, Claudio Bisegni, Jeremy Lorelli, Jingchen Zhou, Auralee Edelen, Chris Garnier, Micha Okun, Hans Thorsten Schwander

And to our collaborators for many useful discussions:

Remi Lehe (LBNL), Revathi Jambunathan (LBNL), and Mat Leputa (ISIS)

Resources

LUME Resources

- Tech paper <https://arxiv.org/html/2606.07250v1>
- <https://github.com/lume-science>
- <https://github.com/slaclab/lume-base>

SLAC VA Implementations

- <https://github.com/slaclab/virtual-accelerator>
- ### Surrogate Templates and Inference Repos
- <https://github.com/slaclab/lume-model-deployment-template>
 - <https://github.com/slaclab/inference-service>

Continual Learning in collaboration with Modcon

- https://github.com/Al-ModCon/BaseSIM_APEIRON/tree/slac-fel