

AI Experimental Control (and calibration) --GlueX Central Drift Chamber

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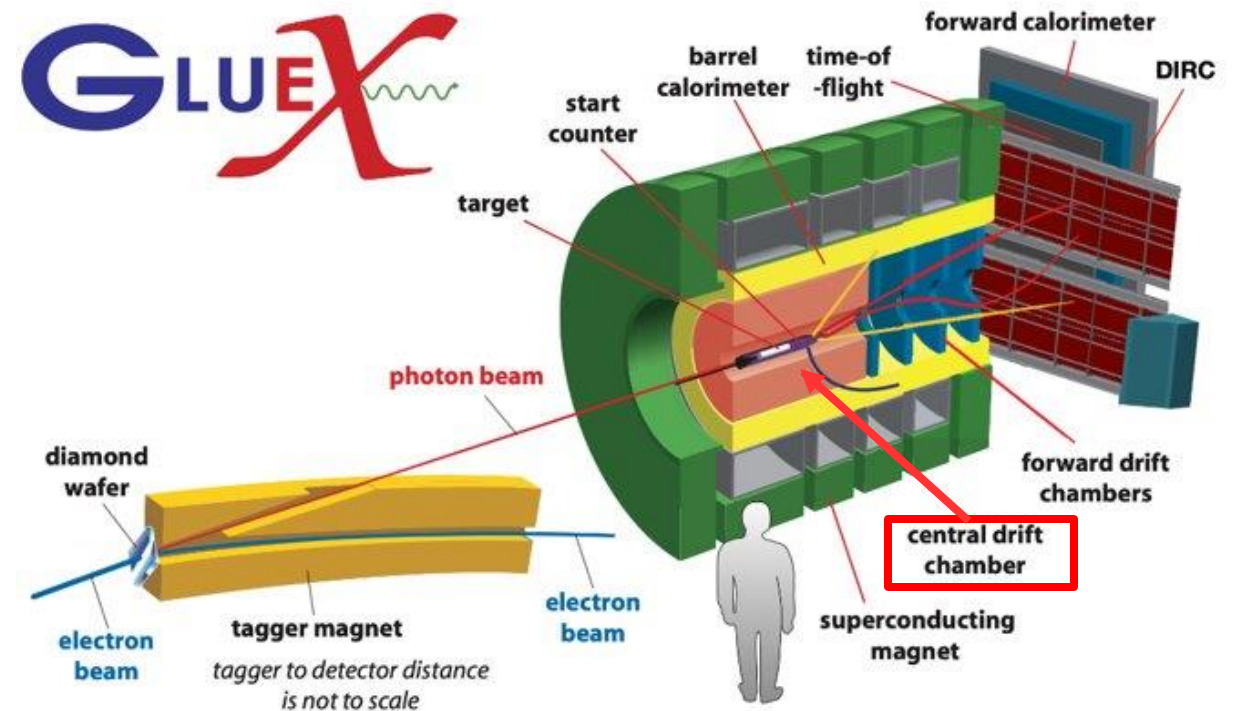
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AI4EIC 2022



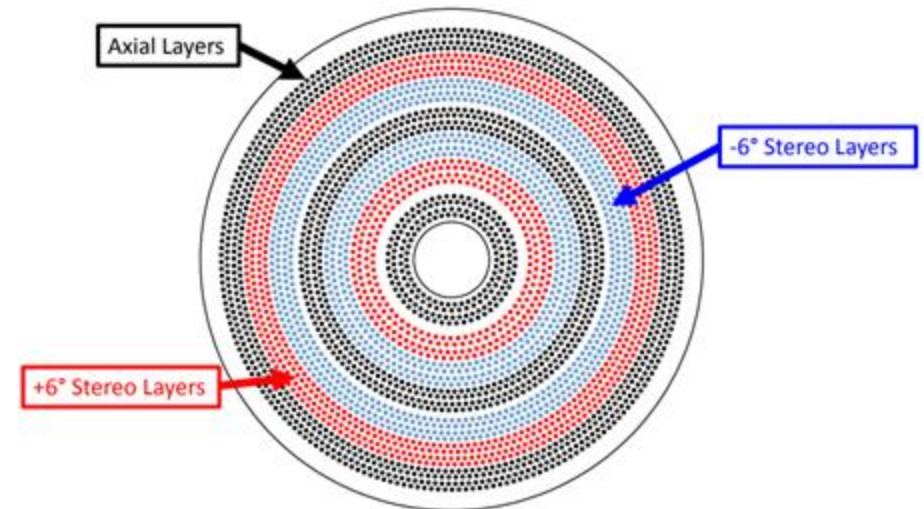
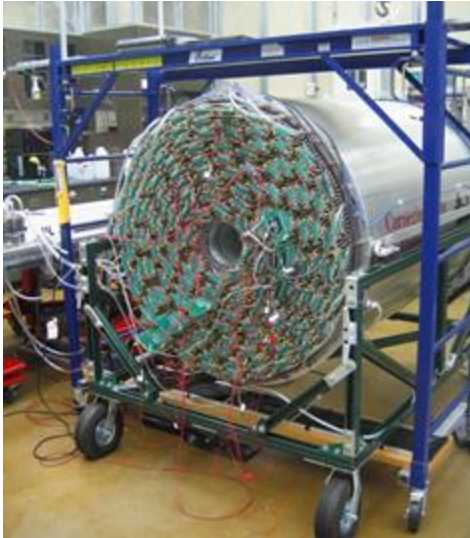
GlueX Experiment at Jefferson Lab

Designed to search for exotic hybrid mesons produced in photoproduction reactions and study the hybrid meson spectrum



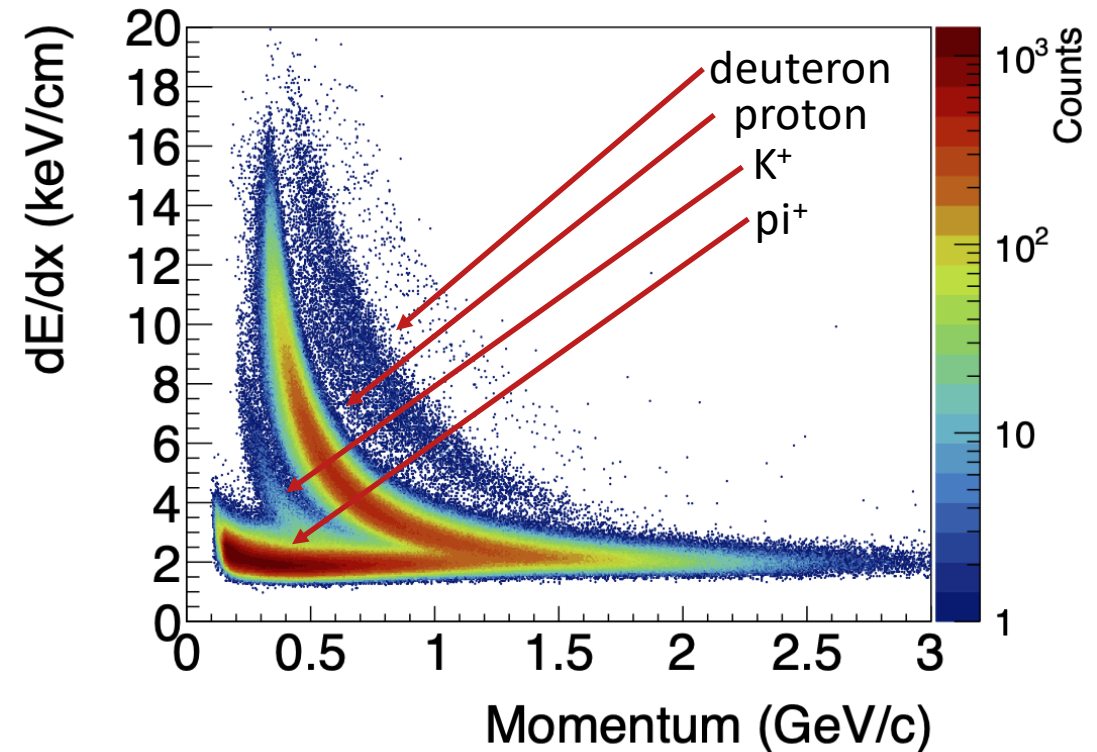
GlueX Central Drift Chamber (CDC)

- **Used to detect and track charged particles with momenta $p > 0.25$ GeV/c**
- 1.5 m long x 1.2 m diameter cylinder
- 3522 anode wires at 2125 V inside 1.6 cm diameter straws
- 50:50 Ar/CO₂ gas mix
- Requires two calibrations: **chamber gain** and drift **time-to-distance**
 - **Gain Correction Factor (GCF):**
 - GCF calibrations have most variation $\pm 15\%$
- Has **one** control: **operating voltage**
 - Nice to start with something simple/easy to vet



CDC Calibrations

- Gain affects PID selections in analysis
 - Known sensitivity to environmental conditions
 - Atmospheric pressure
 - Temperature
 - Known sensitivity to experimental conditions
 - Beam conditions change with the experiment
 - **Traditionally:**
 - GCF obtained from Landau fit to amplitudes
 - Post iterative ttod
 - Calibration constants are generated per run
 - Approximately 2 hours of beam time Partly due to atm changes



Motivation: Conventional vs. Online, ML Calibration Paradigms

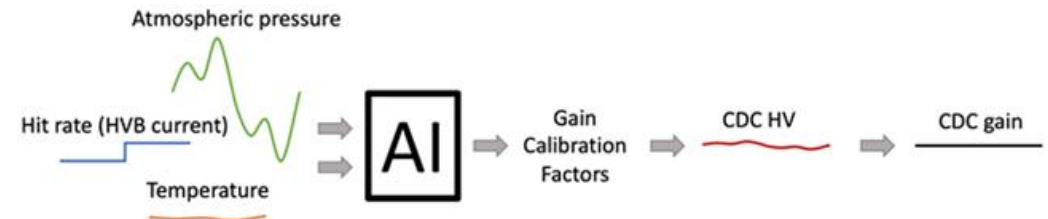
Conventional

- **Calibrate:** calibration values **iteratively**, produced after the experiment
 - ~2 hour runs
- ~~**Control:**~~ CDC operating voltage is **fixed** at 2125 V



Online and ML

- **Control:** Stabilize detector response to changing environmental/experimental conditions by **adjusting** CDC HV
- **Calibrate:** **online** calibration values produced during the experiment



Approach

- Can we predict existing GCF and calculate the recommended HV using data that is readily available before/as a run begins/while running?
 - Similar for time to distance
- Can we control the HV of the CDC to stabilize gain during an experiment?
 - Think coarse and fine adjustment
 - Potential removal of the need to calibrate
- Does the system generalize for differing conditions?
 - How well?
 - Do we trust interpolations and extrapolations?
 - Uncertainty quantification (UQ)
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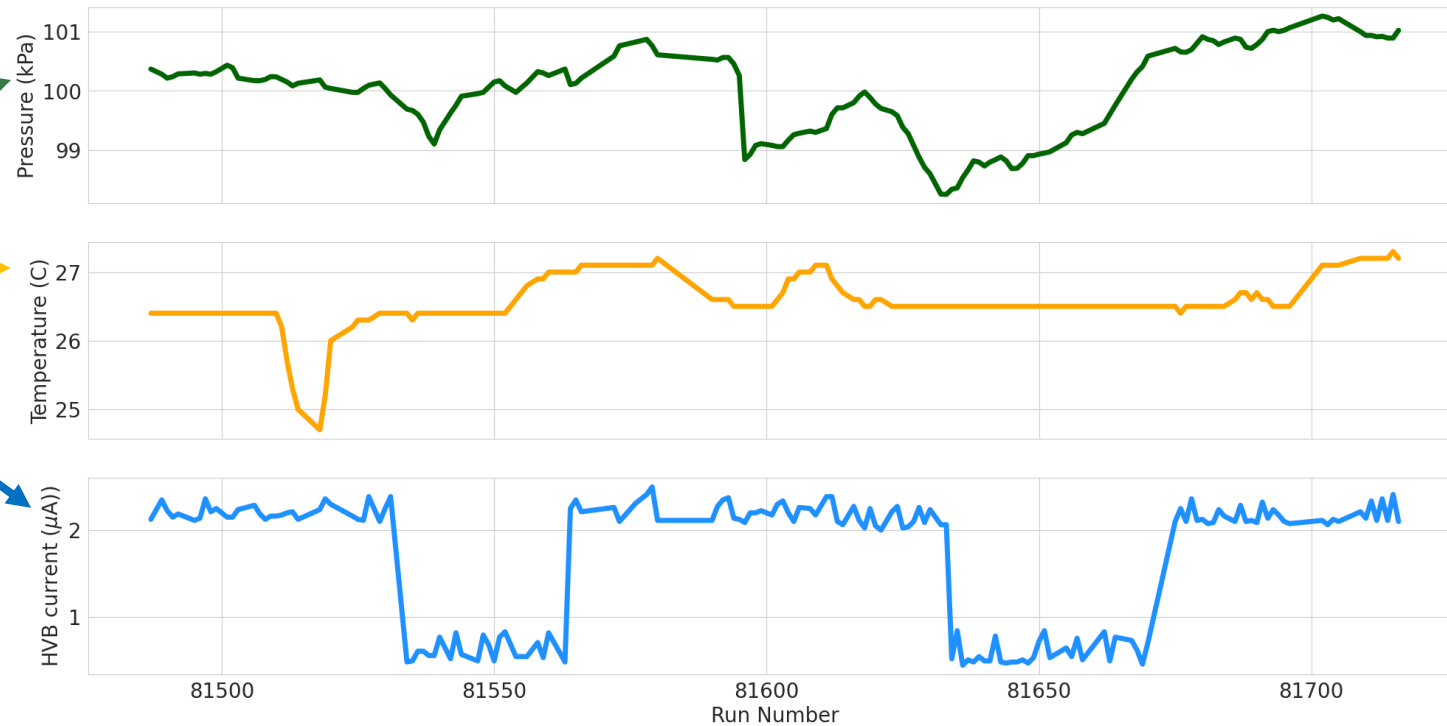
Uncertainty
by crayon.com

Can we predict GCFs: our input features

Can we predict GCFs using data that is **readily available as a run begins?**

For input features:

- Data extracted from Experimental Physics Industrial Controls System (EPICS)
 - Atmospheric pressure
 - Gas temperature
 - Current drawn from CDC HV boards (proxy for beam current)
- All readily available during the experiment
- Not dependent on other detectors
 - No reconstruction!



Can we predict GCFs: our model

ML Technique

Gaussian Process (GP)

- 3 features:
- **atmospheric pressure** within the hall
 - **temperature** within CDC
 - **CDC high voltage board current** -> a proxy for the intensity of the electron beam current
 - 1 target: the traditional Gain Correction Factor (**GCF**)
- GP calculates PDF over admissible functions that fit the data
- GP provides the standard deviation
 - we can exploit for uncertainty quantification (UQ)
- We used a basic GP kernel:
 - Radial Basis Function + White
 - Tested isotropic (1 length scale) and anisotropic (length scale per input variable) kernels

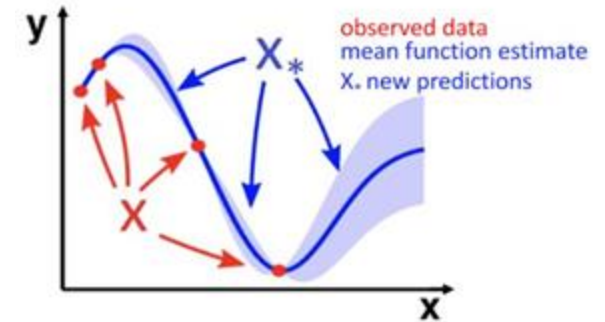
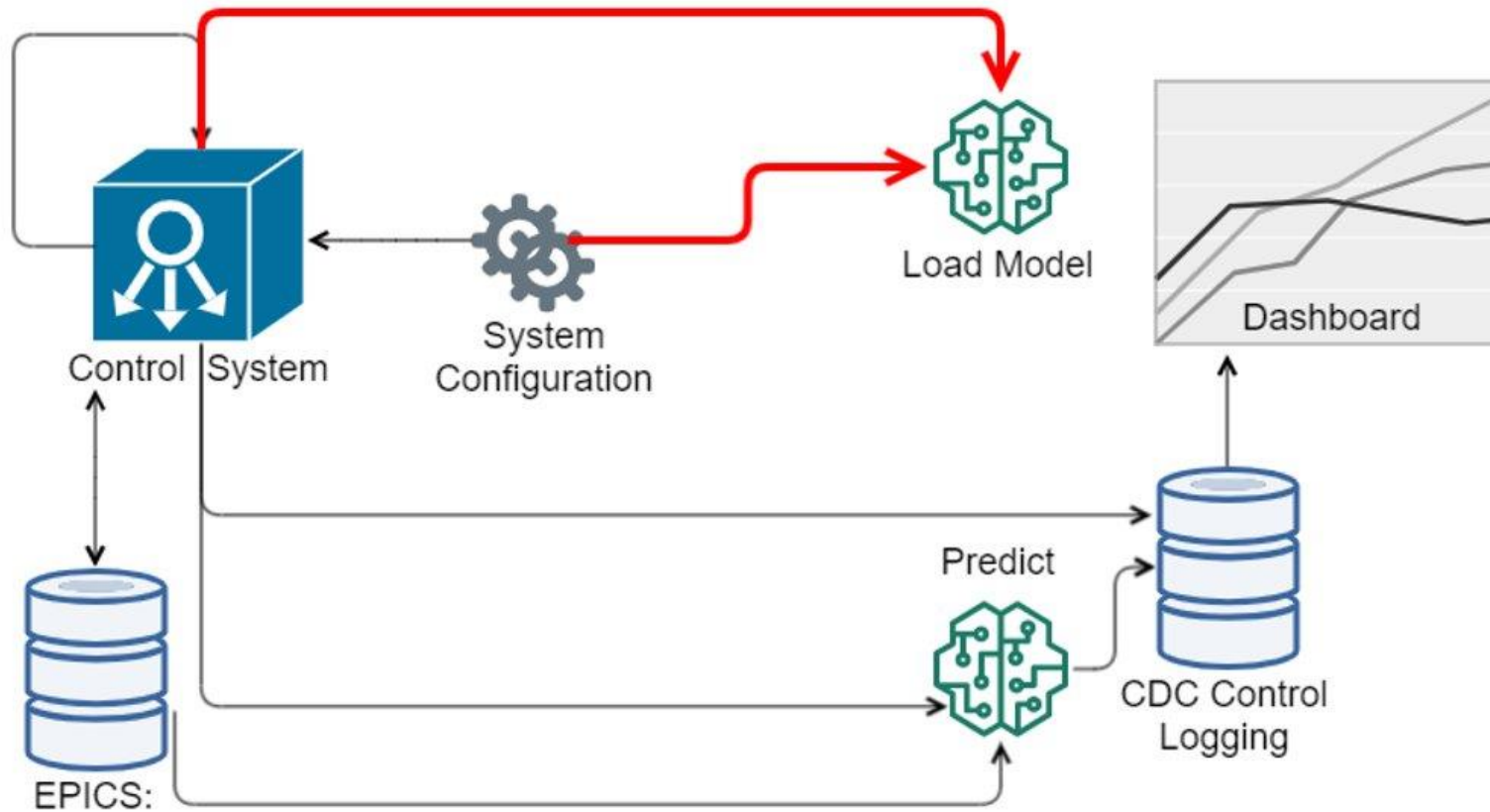


Illustration training a Gaussian process

Threshold was better than 5% error

RBF kernel (length scale(s))	R^2	RMSE	Mean % err
Isotropic (1.412)	0.97	0.002	0.8%
Anisotropic (1.4,1.17,.171)	0.97	0.002	0.8%

The system

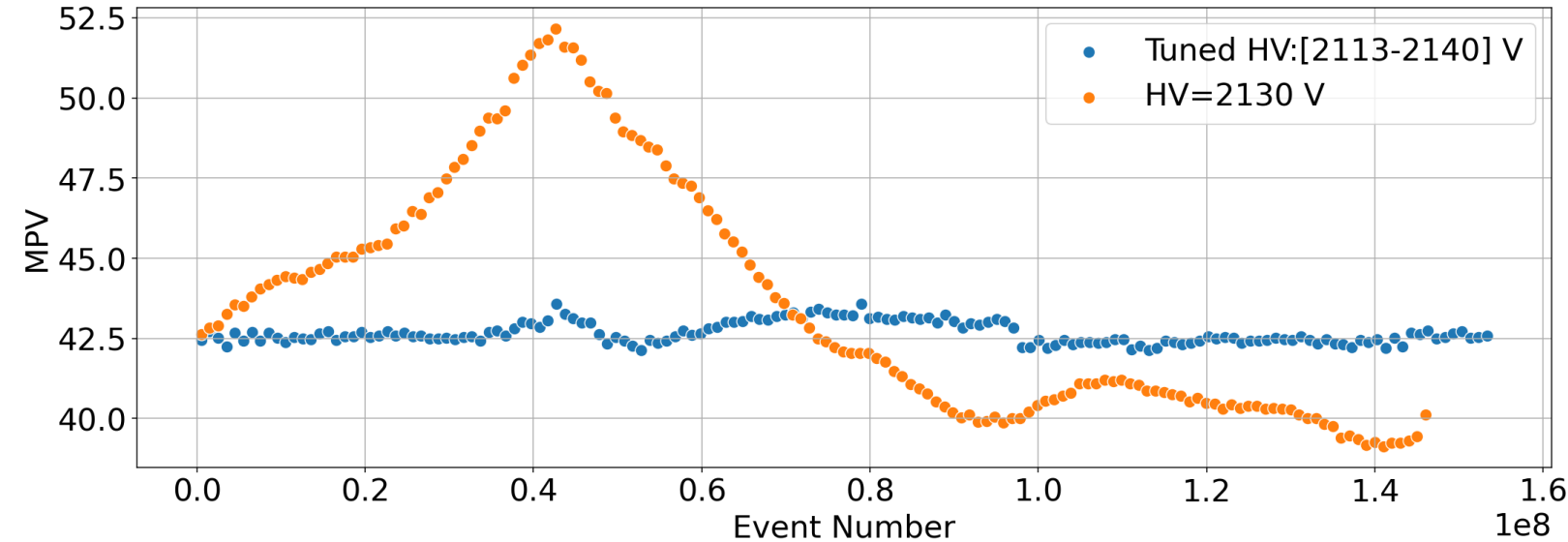


- On the fly configuration
 - Always on. Not always recommending/controlling
- **UQ threshold dictates projection**
 - Never be too far out of bounds
- Full logging for future investigations
 - Interpretability studies

Experimental Physics and Industrial Control System

Commissioning (cosmics)

*MPV to gcf is just a scaling

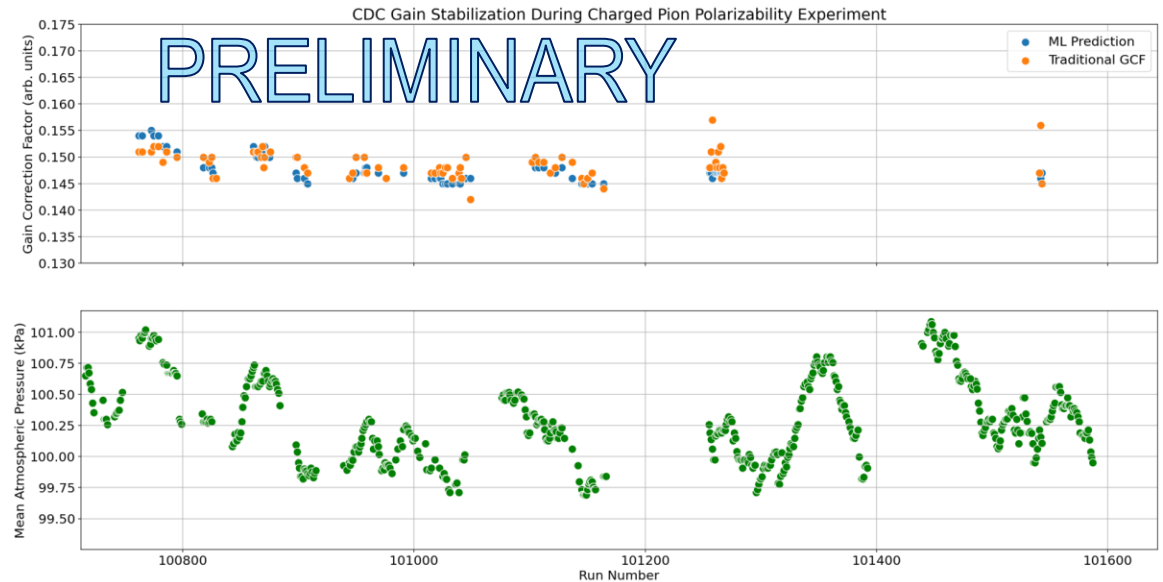


Big success!

- Half the CDC (orange) at fixed HV
- The other half (blue) had its high voltages adjusted every 5 minutes
- The drop at 1.0 was due to a temperature adjustment to bring things back in line with nominal running
 - It is a small effect

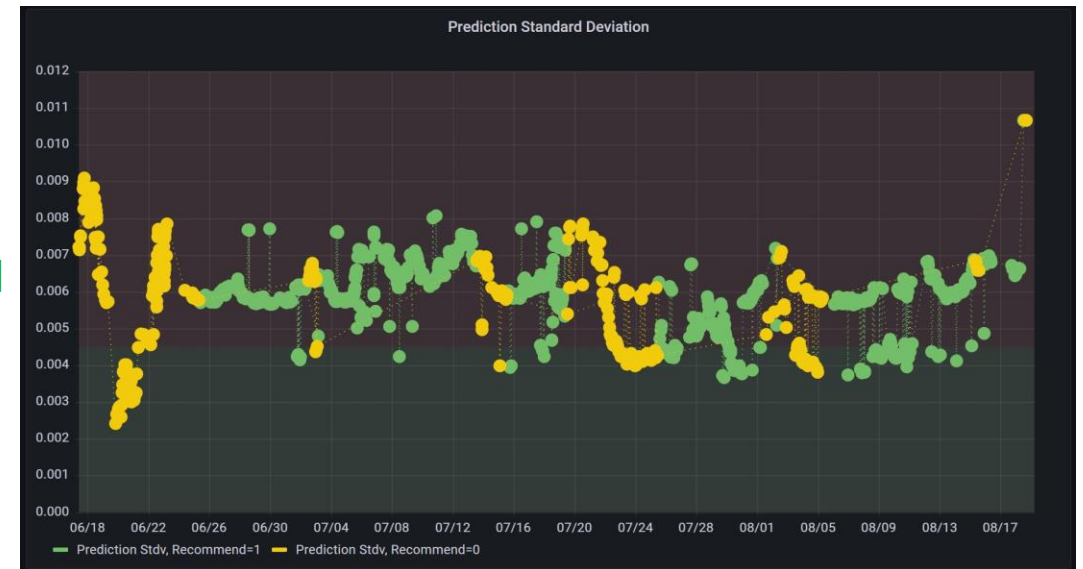
Running (CPP)

- Differences to GlueX
 - Lower hvbi
 - Quarter of the CDC electronics borrowed for a CPP specific downstream detector
 - Solid target
- Update HV at start of run
- Atmospheric pressure didn't see as wild variations
 - If only we could control the weather....
- Almost all the time was spent well above our UQ bubble
 - ~.006



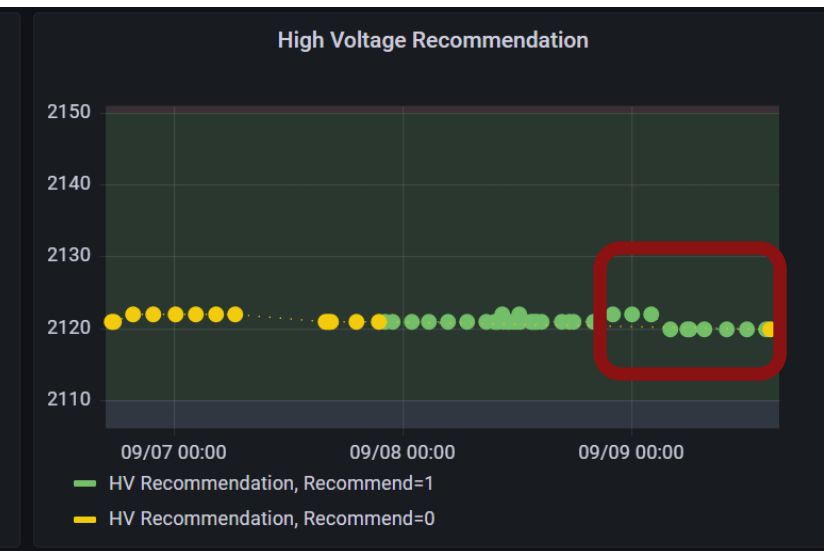
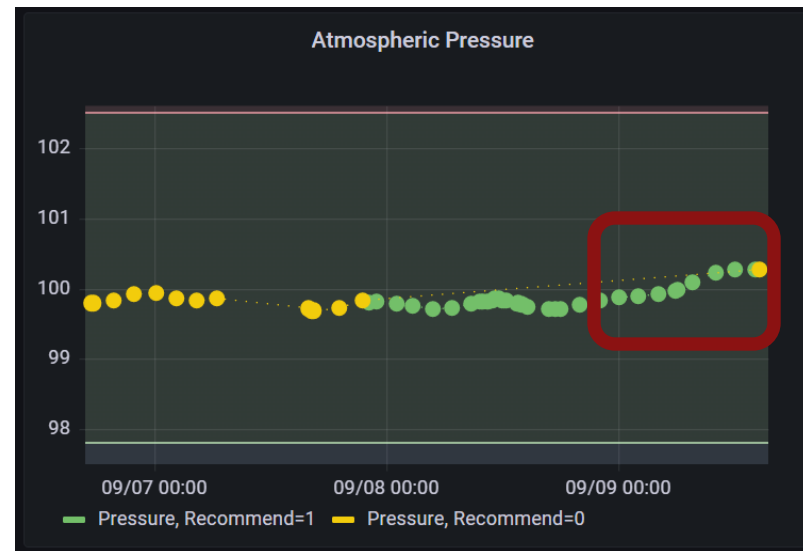
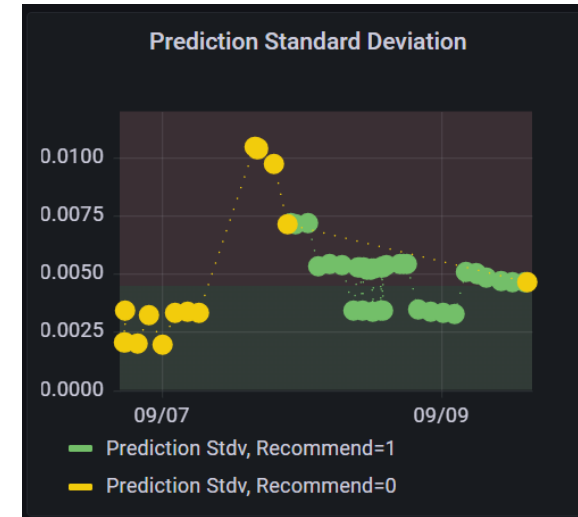
HV not
controlled

HV
controlled



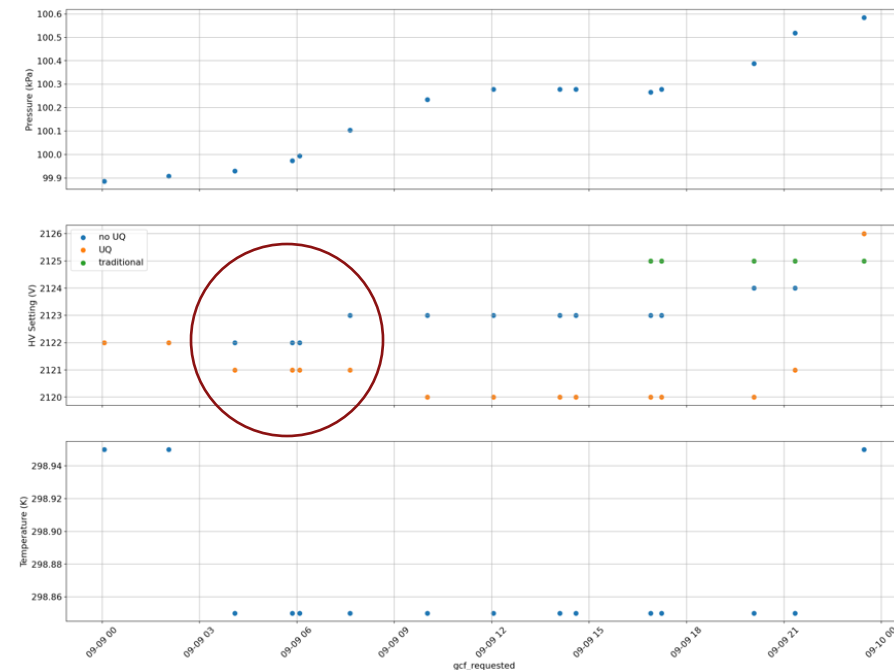
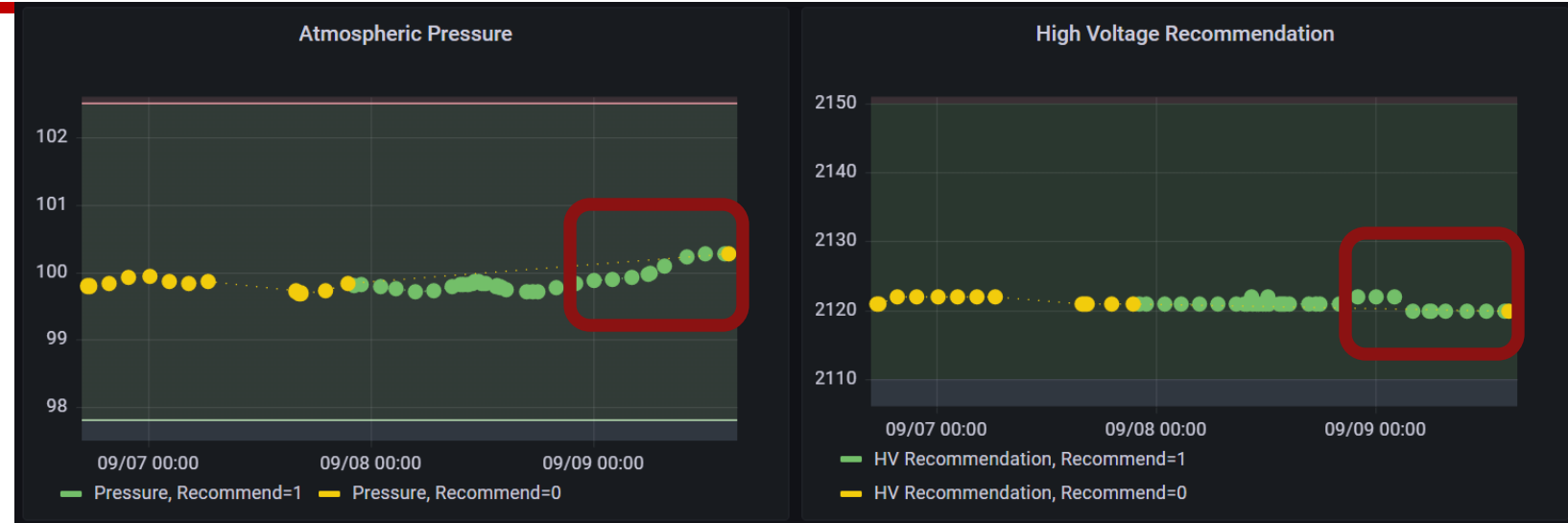
Running (Primex)

- Also dissimilar to GlueX
 - Lower current but closer to nominal running
 - Whole CDC
 - Much closer to the uncertainty threshold
- Update HV at start of run not during
- Anomalous behavior seen
 - Up atm should be up HV
 - Everything else flat
- Why?



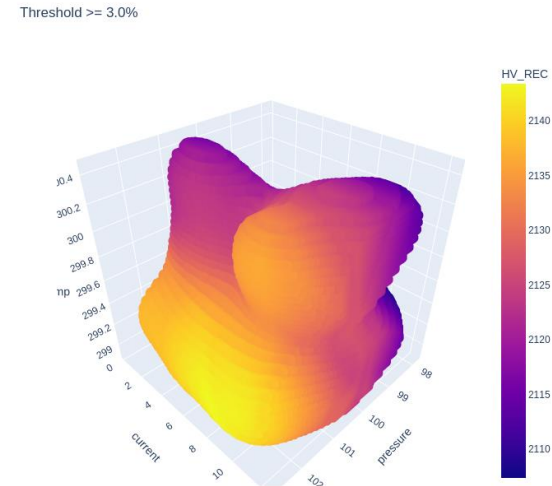
Anomalous behavior

- How bad is it?
 - 1% diff in gain \sim 1V diff
 - Maximal error 4V \Rightarrow gain error of 4%
 - Below our 5% target
- Still within tolerances
- Looking at the UQ corrected HV versus the raw HV we see a **bifurcation**
 - Anomalous behavior strictly from the UQ correction

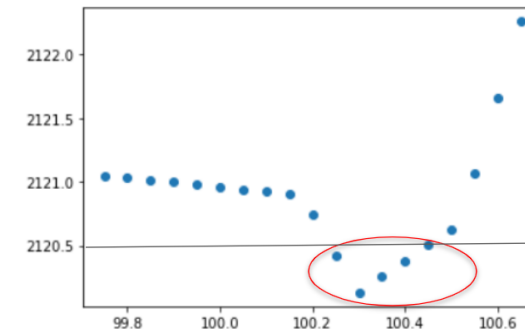


Subtleties

- Our uncertainty surface is not really a bubble
 - By that I mean purely convex
- Any dents or divots can affect the projection
 - Imagine being at the center of a bowl, tiny movements can send you from one side of the bowl to the other possibly with drastically different HV recommendations
- Maybe we didn't see this in CPP because the distance away from the surface was much greater
 - When you are far from the surface it takes much greater motion to switch points of projection



The “divot” in our uncertainty vertices



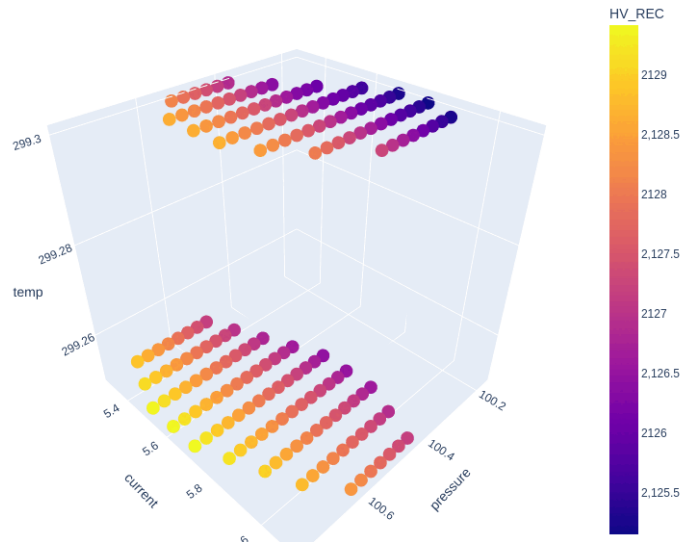
Here is one! In a slice of our surface

Raw HVBI: 9.0, Raw Temp: 298.85
Raw Pressure: 99.9298 UQ_Pressure: 99.95 UQ_hvbi: 9.0 UQ_temp: 298.9 UQ_GCF: 0.1429838685738982 HV_REC: 2120.979390944207 HV_REC rounded: 2121.0
Raw Pressure: 99.9754 UQ_Pressure: 99.95 UQ_hvbi: 9.0 UQ_temp: 298.9 UQ_GCF: 0.1429838685738982 HV_REC: 2120.979390944207 HV_REC rounded: 2121.0
Raw Pressure: 99.9952 UQ_Pressure: 100.0 UQ_hvbi: 9.0 UQ_temp: 298.9 UQ_GCF: 0.1429838685738982 HV_REC: 2120.960977904075 HV_REC rounded: 2121.0
Raw Pressure: 100.104 UQ_Pressure: 100.1 UQ_hvbi: 9.0 UQ_temp: 298.9 UQ_GCF: 0.1429838685738982 HV_REC: 2120.9214342124624 HV_REC rounded: 2121.0
Raw Pressure: 100.235 UQ_Pressure: 100.25 UQ_hvbi: 9.0 UQ_temp: 298.9 UQ_GCF: 0.14213505840759677 HV_REC: 2120.4169039125636 HV_REC rounded: 2120.0
Raw Pressure: 100.279 UQ_Pressure: 100.3 UQ_hvbi: 9.0 UQ_temp: 298.9 UQ_GCF: 0.14170531958731739 HV_REC: 2120.1308776985165 HV_REC rounded: 2120.0

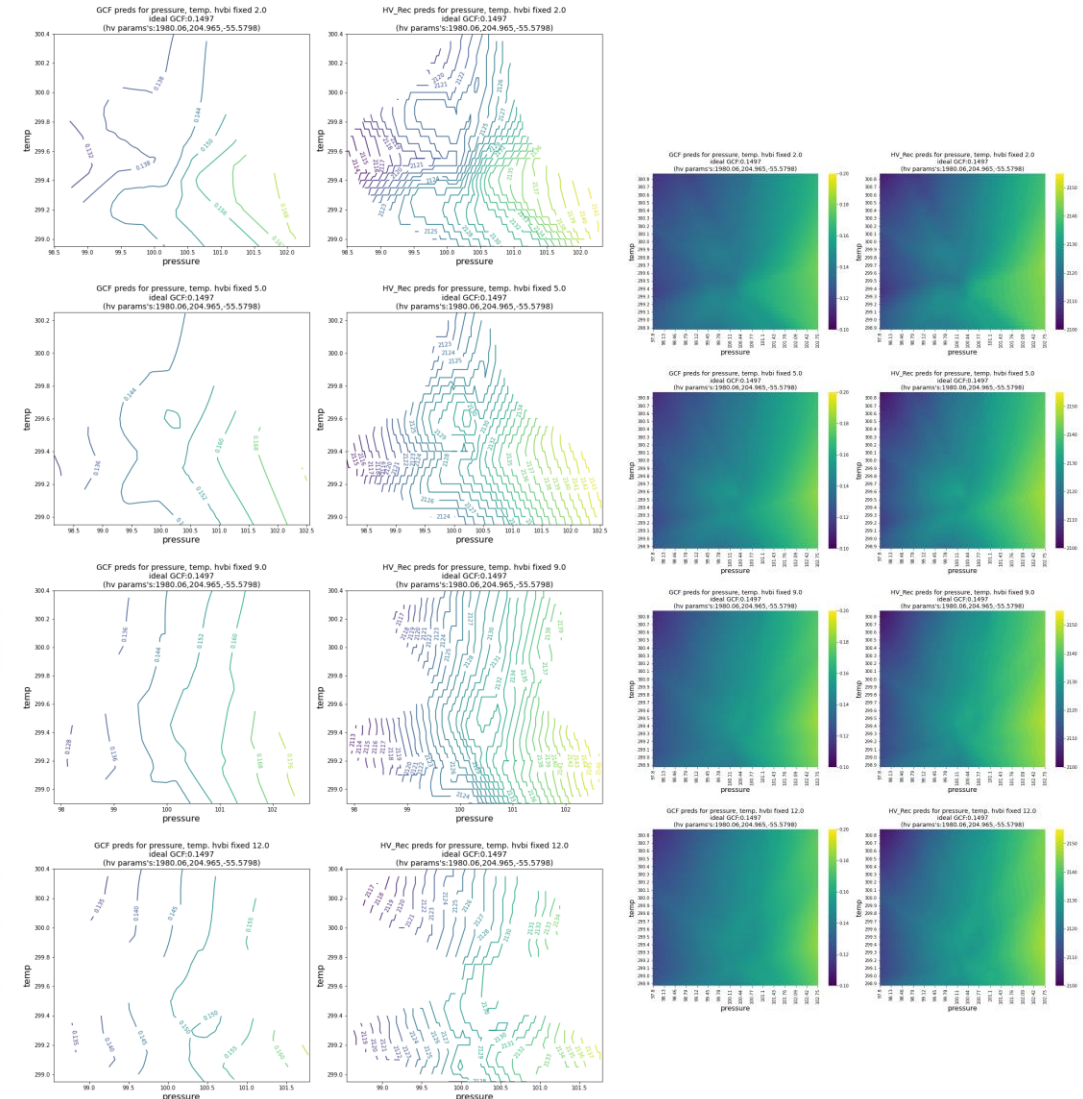
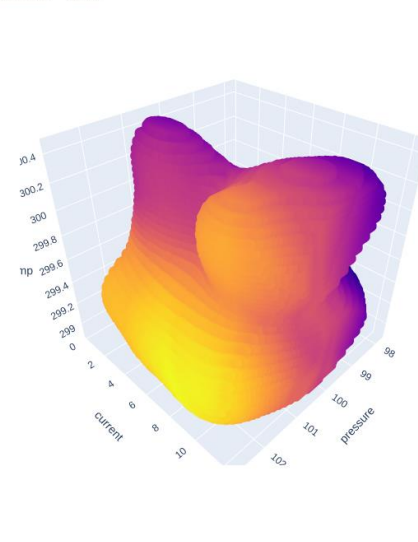
Analysis/Debugging the Black Box

- Plot the surface and the point(s) of measurements
 - Ends up being a bug in projection but indicates a closer look at the UQ
- Looks less like pancakes/a bubble and more like swiss cheese
 - Pockets of space that are "inside" but above UQ

Threshold $\geq .7\%$

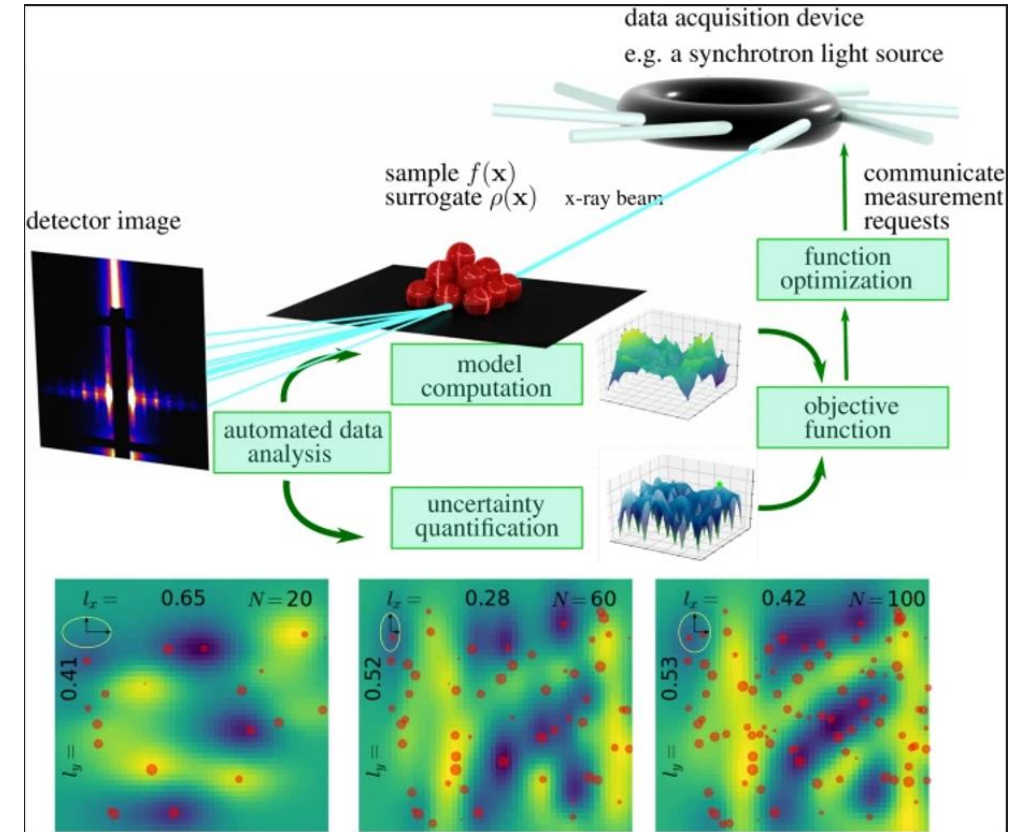


Threshold $\geq 3.0\%$



Towards Self-Learning

- One application of ML is in experimental design (parameter selection)
 - Better/more efficient than exhaustive grid search
 - Based on the idea that all data is not created equal
- Turning things on its head, it may be possible to use a similar system to decide whether it is better for us to take new data or use our projection
 - Complex error surface
- There are lots of avenues for research into UQ and its actionable utilization

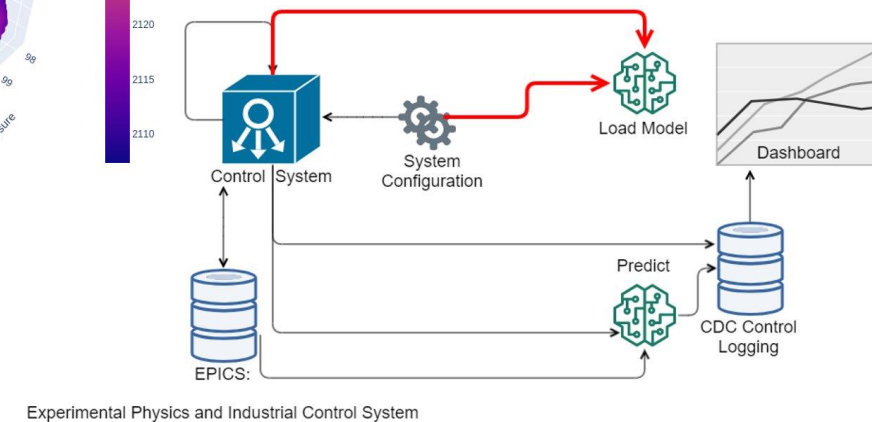
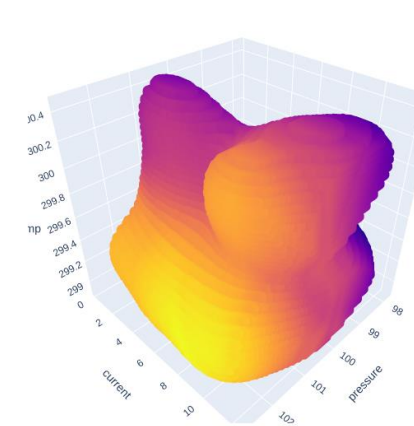


<https://www.nature.com/articles/s41598-020-74394-1>

Conclusions/Future

- Successfully deployed an AI system to control and calibrate the GlueX CDC
 - ~ 1% error in predicting gcf (<< promised 5%)
 - Able to stabilize gcf via HV control
 - The system is reconfigurable on the fly
 - Uses estimated UQ in an actionable way
 - Lots of subtleties and avenues for research
 - Complex 3D surface
 - Limited data in some dimensions
 - Islands of data
- Looking to finishing TtoD
 - HV dependence so a bit of bootstrapping probably required
- Probing other detectors and systems to apply these techniques
 - Potential self-learning to enable eminently deployable system which will only control when "confident" and efficiently learn when not

Threshold $\geq 3.0\%$



Backup slides

Calibrations with AI: Gain

- AI generated calibration constants agree with conventional gain calibration results
- GCF are more stable compared to GlueX 2020 run period

