

ML+LLRF: From Demonstrated Laser Control to SRF Resonance Control

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Dedicated AI/ML Accelerator Test Facility

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SLAC

Acknowledgements to all contributors / collaborators

- **Teams & Collaborators:**

LBNL: Larry Doolittle, Russell Wilcox, Qiang Du, Shreeharshini Dharanesh Murthy, Alessio Amodio, Stijn Wielandt, Curtis Berger, Samuel Barber, Jeroen van Tilborg, Remi Lehe, Tong Zhou, Fanting Kong, Gang Huang, Yilun Xu, Qing Ji, Hai-En Tsai, Alex Picksley, Zak Eisentraut, Neel Rajeshbhai Vora, Mahek Logantha, Anthony Gonsalves, Cameron Geddes, Thorsten Hellert, Jean-Luc Vay

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Outline

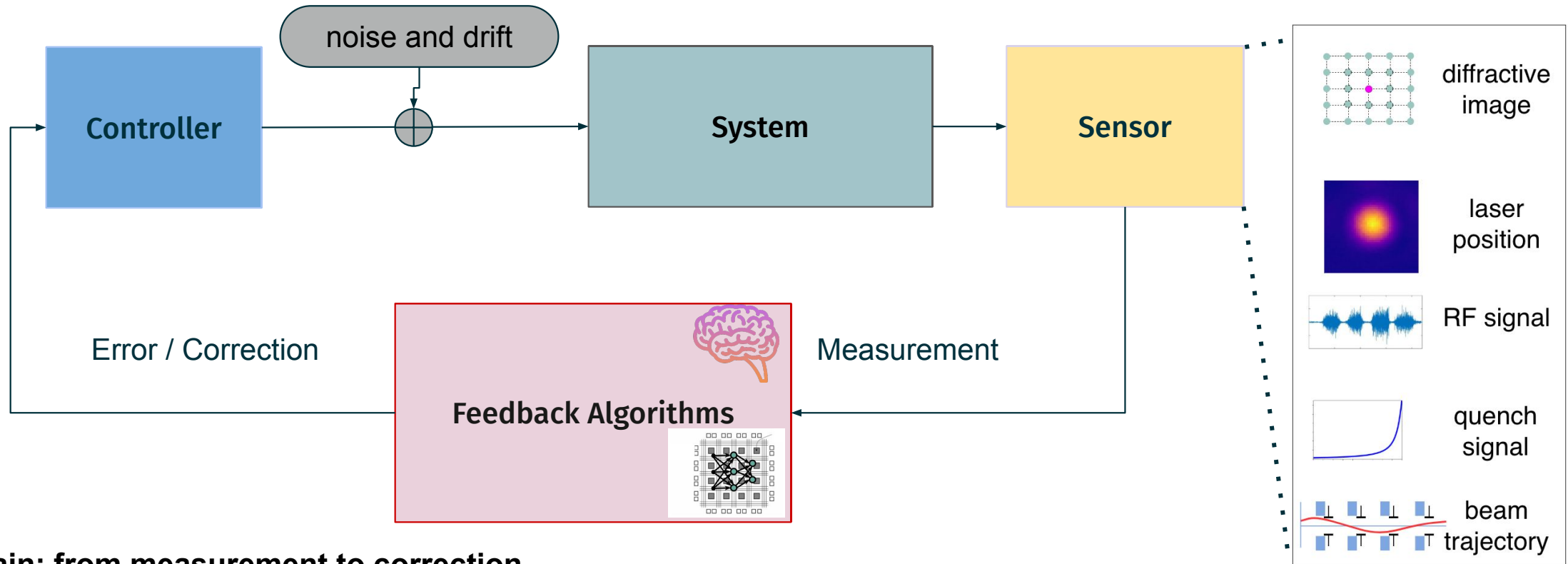
- **ML enables simple, robust and effective controls in general**
- **LBNL-LLRF (Low Level RF) platforms enable science over decades**
- **ML+LLRF:**

Data from LLRF is ML-ready (time-stamped, >kHz, synchronized), this open-source substrate can serve as a tool for general real-time control:

- (1) demonstrated on lasers
 - (2) in progress on SRF
 - (3) open testbed & opportunities
- **Discussion and Outlook**

ML offers a simple, robust path to deterministic control (in general)

Feedback controls are widely used in Lasers & Accelerators:



The brain: from measurement to correction

Hardware-aware AI — data-driven, deterministic (not stochastic exploration) + timing- and latency-aware

- [1] A. Edelen & X. Huang, Annu. Rev. Nucl. Part. Sci. 74, 557 (2024)
- [2] A. Scheinker, A. Edelen, et al., Phys. Rev. Lett. 121, 044801 (2018) — LCLS
- [3] C. Emma, A. Edelen, et al., Phys. Rev. Accel. Beams 21, 112802 (2018)
- [4] S. Leemann, et al., Phys. Rev. Lett. 123, 194801 (2019) — ALS
- [5] A. Scheinker, F. Cropp, D. Filippetto, Sci. Rep. 11, 19187 (2021)

LBNL's control and timing technology serves DOE's accelerator projects opens a new opportunity for lasers, qubits and complex system controls

FPGA Soft. defined radio **LCLS** 10s fs fiber optics timing 1E-4 SRF LLRF Qubit / laser controls RFSoc AI-LLRF

2001



2006



2009



2014



2017



2022

2026



<https://github.com/BerkeleyLab/marble>



LCLS-II chassis

L. Doolittle et al., SLAC-PUB-17108

**LLRF for LCLS-II and LCLS-II HE
PIP-II Accelerator complex controls
LLRF Upgrade for AWA, ANL
LLRF for ALS-U AR, LBNL**

...

Scalable digital control in fiber laser combining
(Russell Wilcox, Qiang Du, Tong Zhou, et. al.,)

Digital controls for qubits
(Gang Huang et. al.)

AI-enhanced LLRF (in progress)

L. Doolittle, et al.; <https://github.com/BerkeleyLab/Marble>

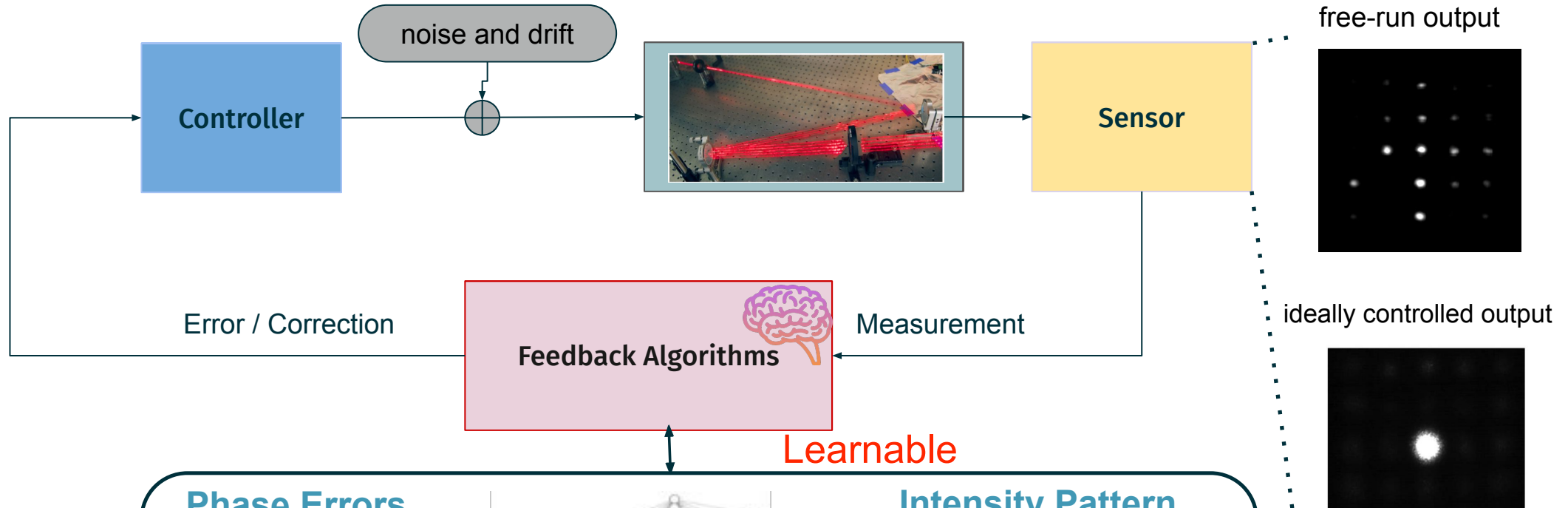
Q. Du, et al., Digital Low-Level Optical Control for Coherent Laser Combining, 2023/10/23 LLRF Workshop 2023

G. Huang, et al., How a novel radio frequency control system enhances quantum computers, LBNL Lab News, 2021

Over decades of innovation in precision control & timing technology and application under support of DOE SC.

Laser combining: ML offers solution with complex pattern recognition

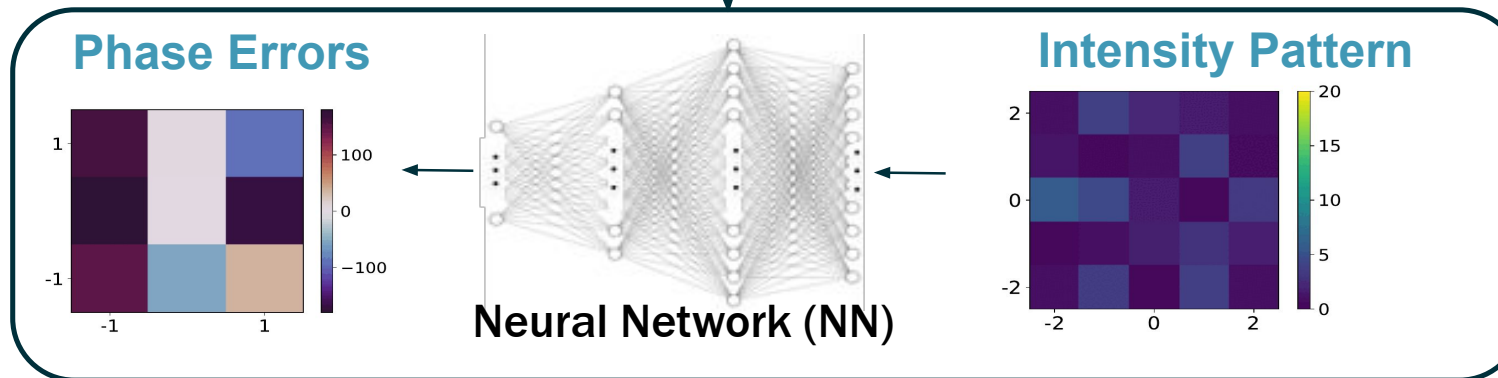
Feedback controls in laser combining:



Example of one control chassis in the laser system



Qiang Du, [et.al](#),



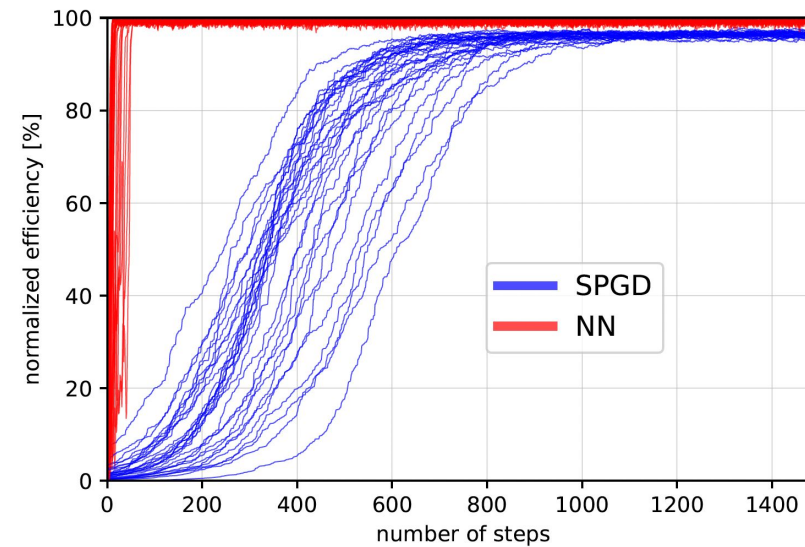
Synchronized kHz data
+ Physics-guided ML
enables learning in the
real world.

Recognition Neural Networks (NN): from pattern to error

Physics-guided ML delivers closed-loop simulation and real-world demonstration

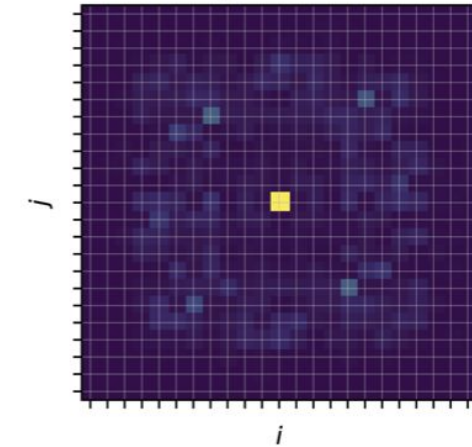
Simulation closed the loop:

- Fast convergence
- Better performance

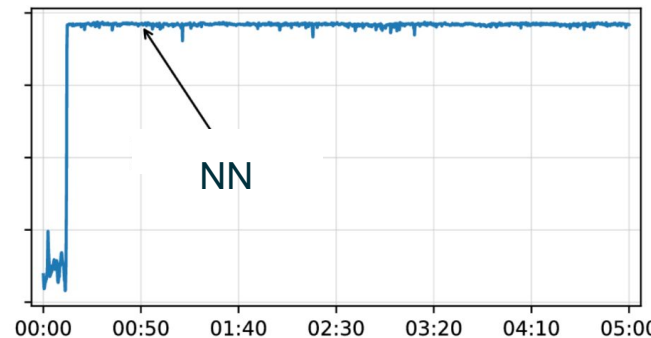
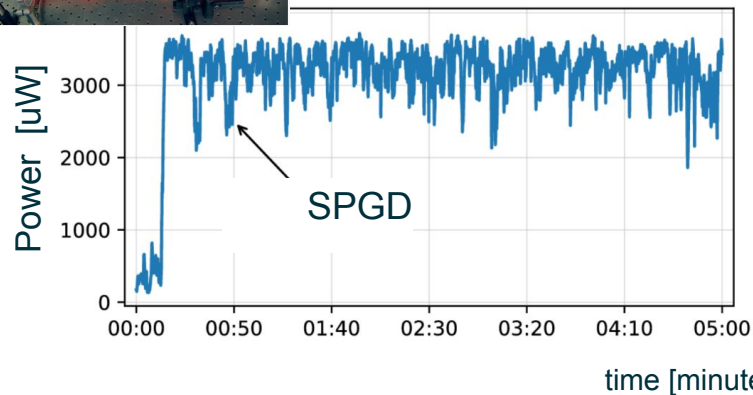
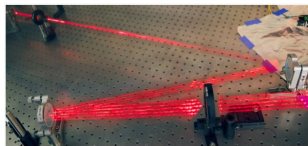


D Wang, Q Du et.al., (2021). Optics express, 29(4), 5694-5709

Simulated, center saturated



Q Du, D Wang et.al., (2021). Optics express, 29, 5407-5418

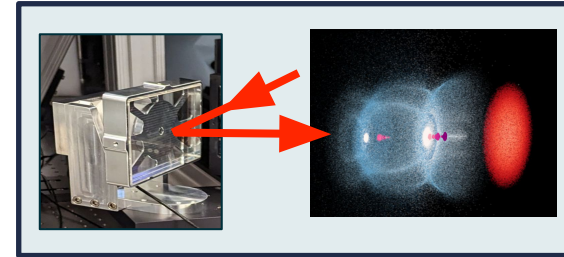
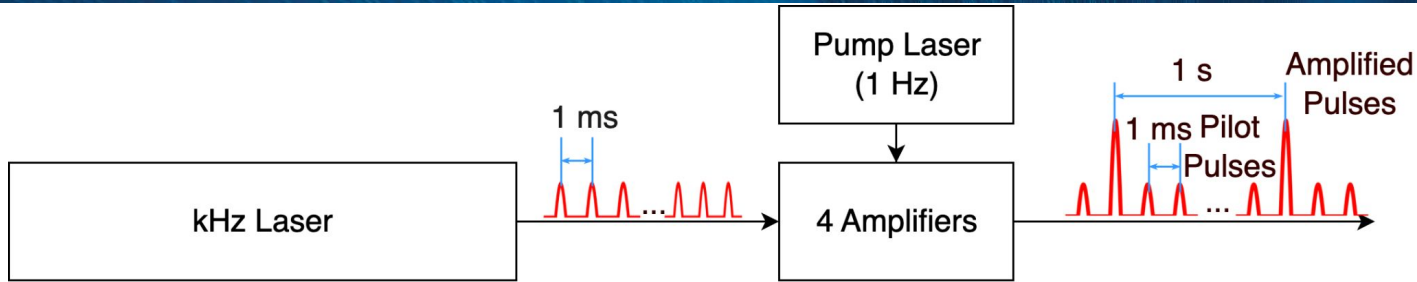


With NN control in **experiment**:

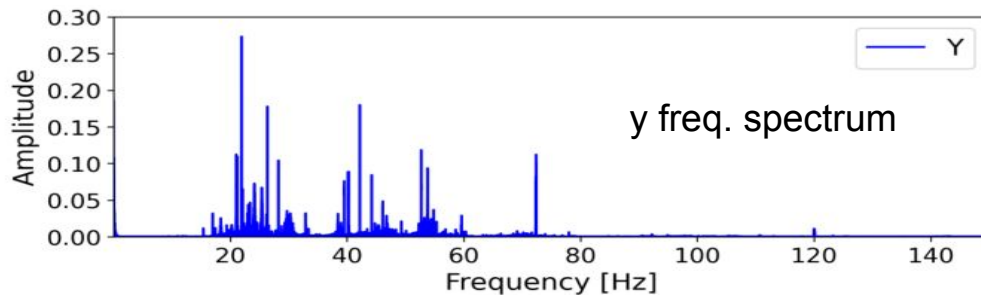
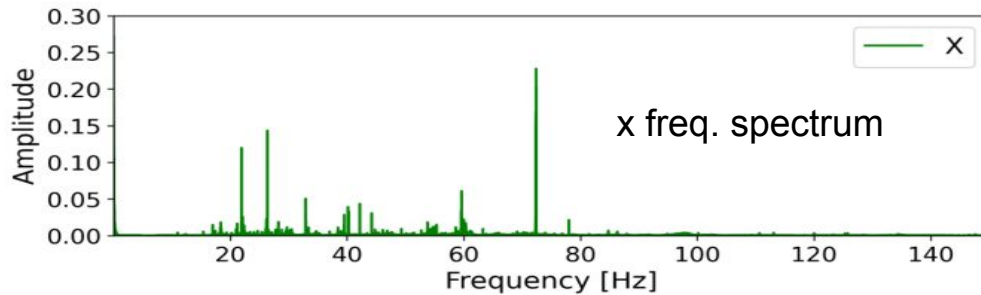
- Unprecedented stability

Q. Du, D. Wang, et al., Opt. Express 30(8): 12639 (2022)

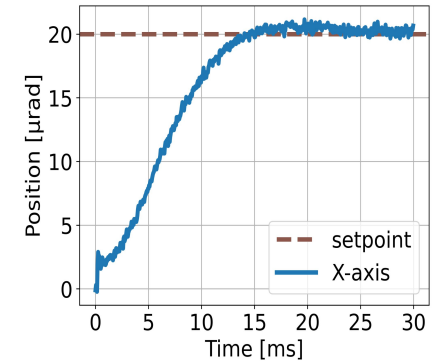
Laser pointing: traditional feedback is insufficient — low control bandwidth



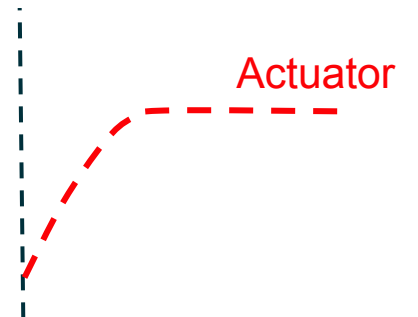
Noise Spectrum -> high frequency up to 100 Hz



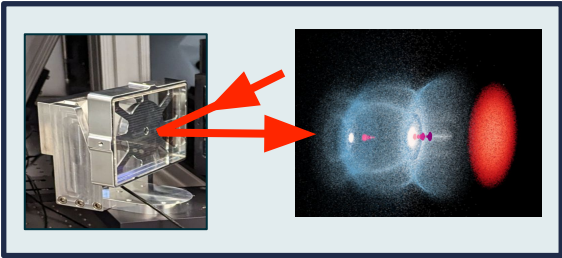
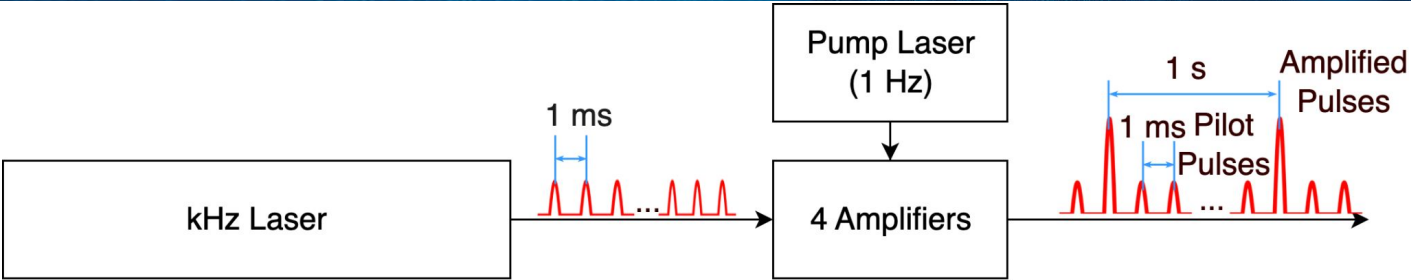
Slow mirror -> response time up to 20 ms



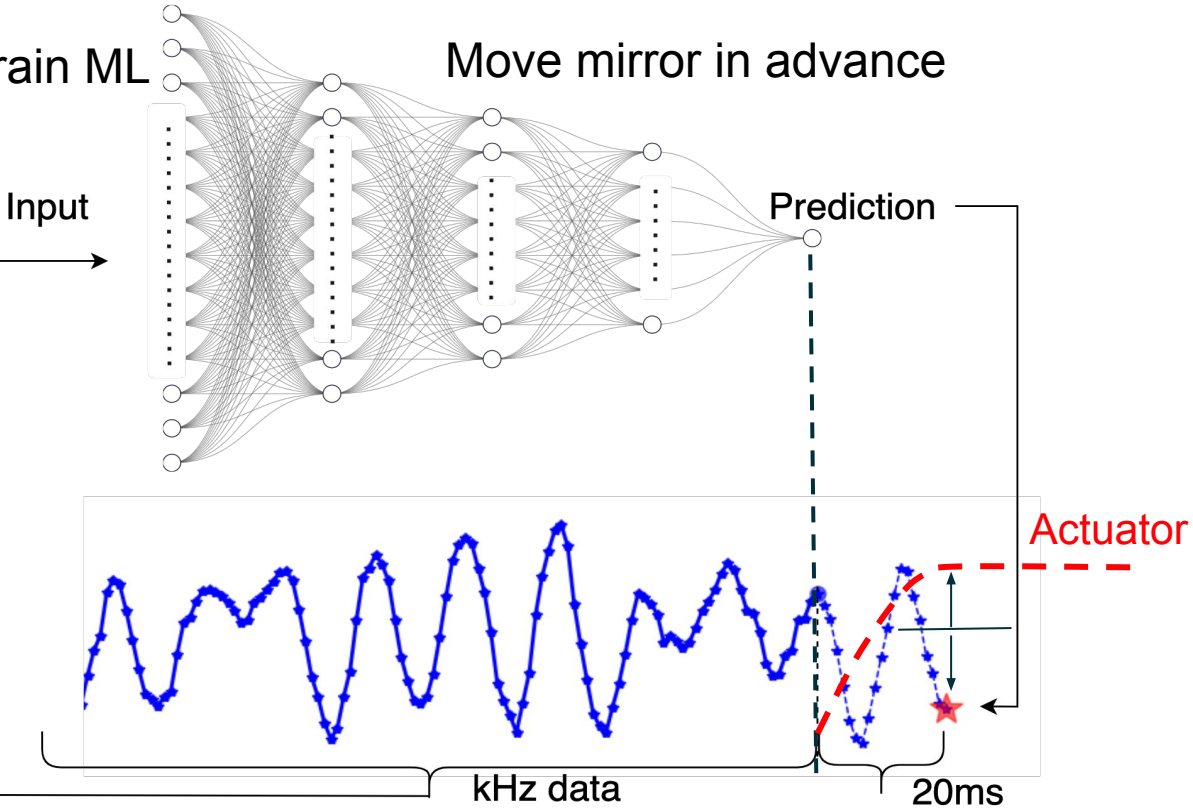
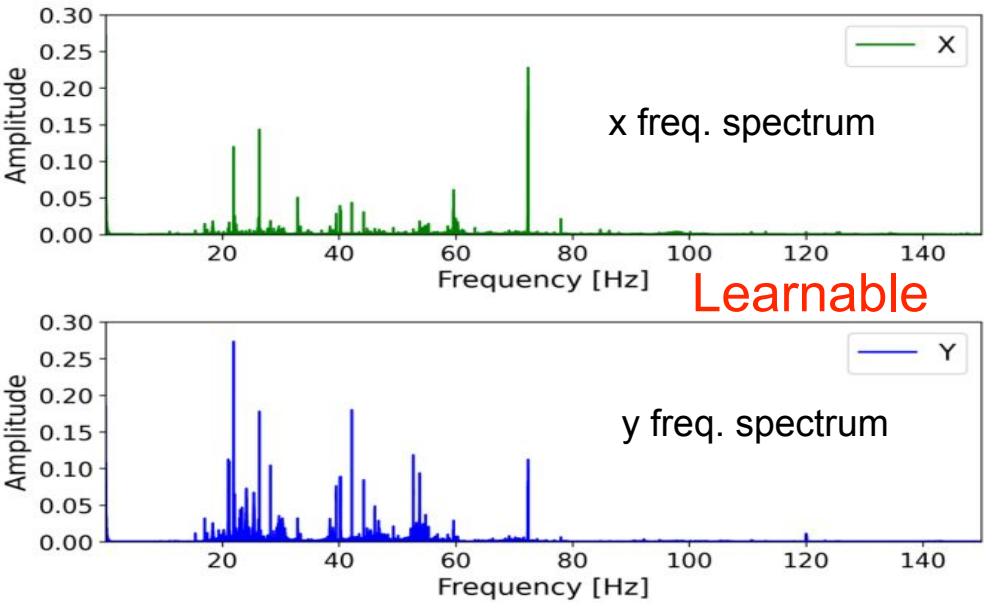
Challenges



Feed-forward prediction from a kHz pilot beam enables predictive control



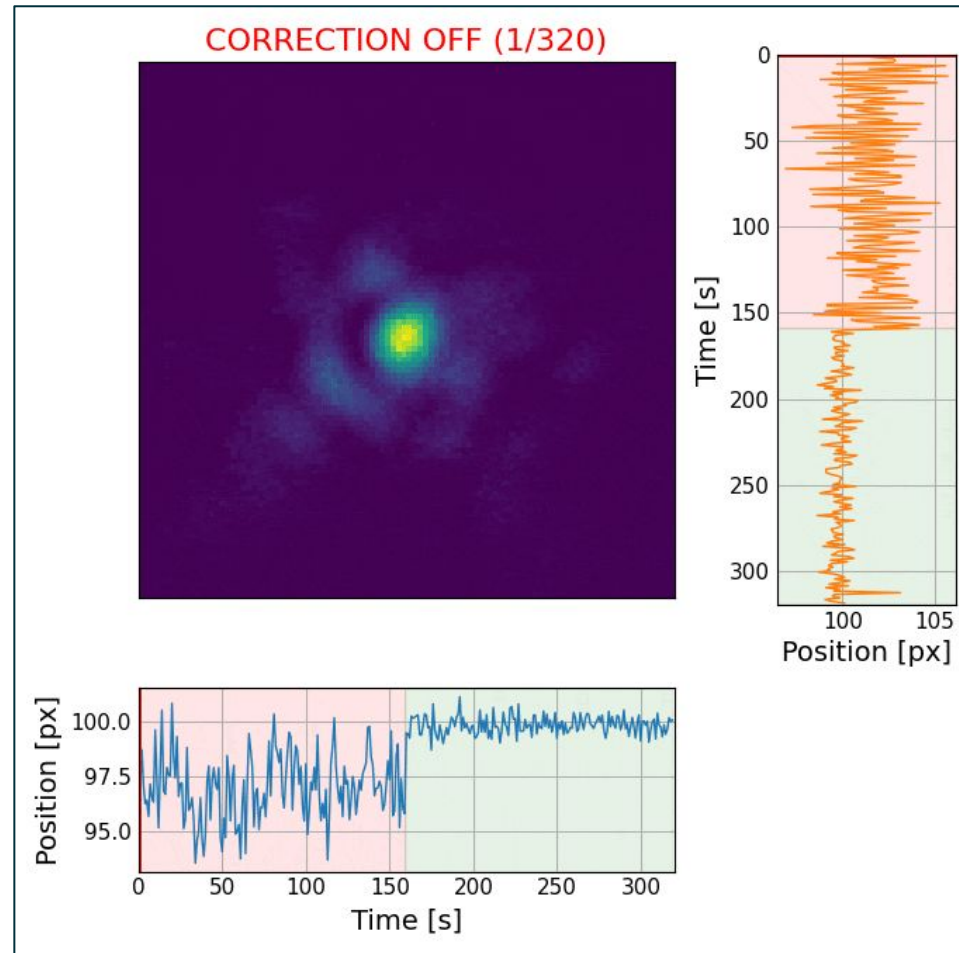
kHz pilot beam (complete noise information) used to train ML



-F. Isono et al. High Power Laser Science & Engineering, 9, e10 (2021)
-C. Berger et al. NIMA 1078 (2025) 170561

Experimental Results for a 30 mJ (TW) Operation, Demonstrating Correction

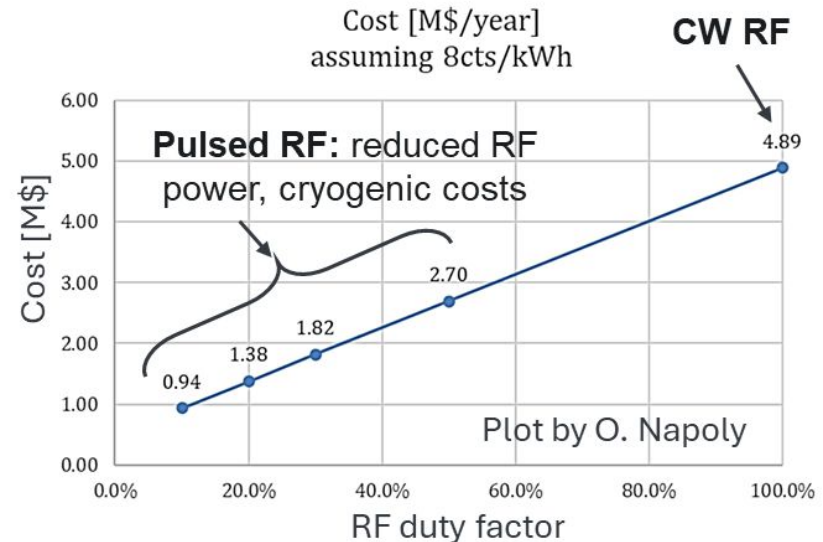
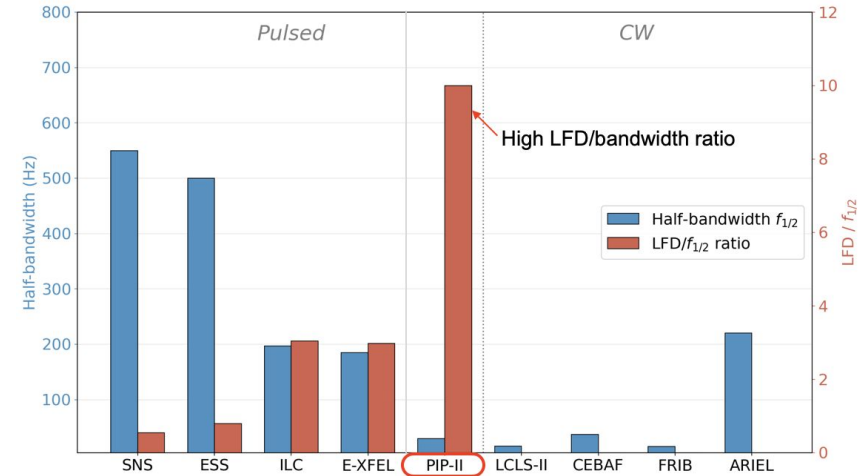
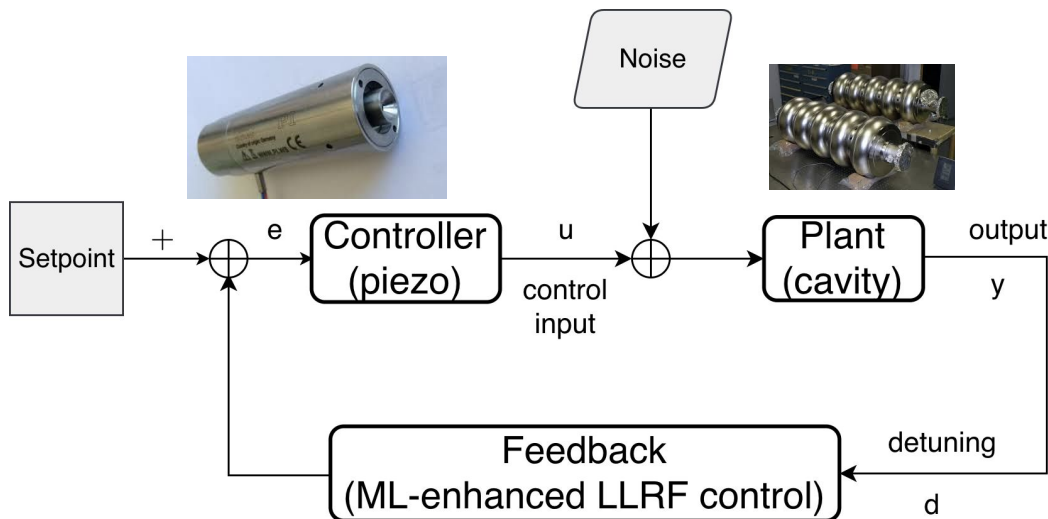
- Achieved jitter reduction of ~65% in X and ~47% in Y!



- A Amodio, D Wang, C Berger, et al. *High Power Laser Science and Engineering*. 2025;13:e35 (2025)
- <https://newscenter.lbl.gov/2025/06/10/machine-learning-helps-ease-the-jitters-of-high-power-lasers/>

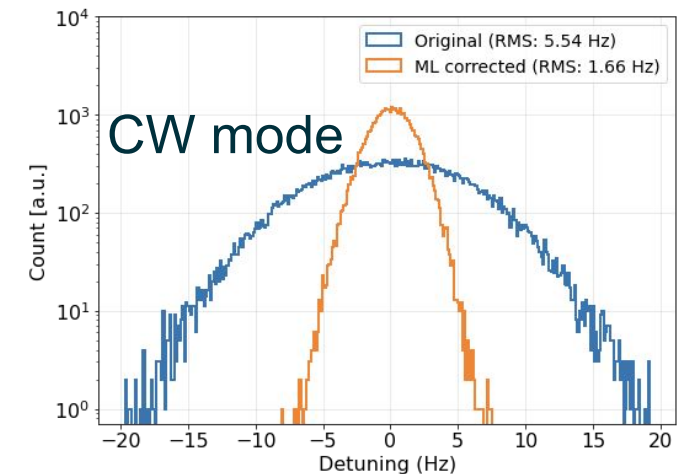
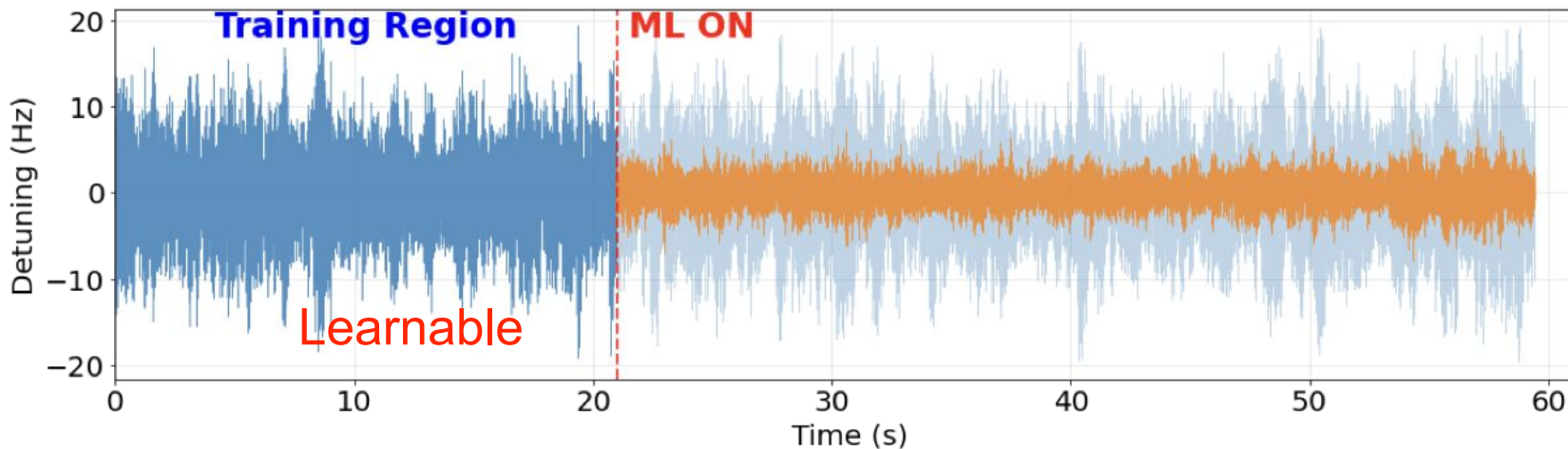
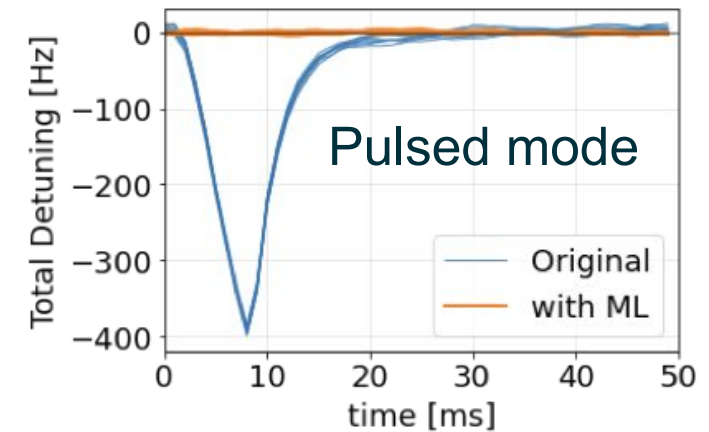
AI+LLRF for SRF resonance control, enabling PIP-II pulsed operation

- SRF cavities: thin walls + narrow bandwidth → deformations from Lorentz-force detuning and microphonics detuning → off resonance (waste power + unstable operation) → active resonance control essential
- PIP-II LLRF baseline is CW operation — same as the LCLS-II LLRF system; the challenge comes from narrow-band cavities at high gradient, giving large + dynamic lorentz force detuning (LFD)
- Ongoing R&D on hardware-aware AI aims to enable PIP-II pulsed mode — building on today's LCLS-II LLRF systems — meeting the designed pulsed-beam specification at a potential \$M/year lower operating cost.
- The complexity — and the opportunity for AI — comes from nonlinearity, disturbances spanning different amplitudes and time scales, and cavity-to-cavity and facility-to-facility variation.
- Again, years of LBNL-LLRF development provide a strong starting point.



Recent updates: ML learns detuning patterns in a decoupled setup (pulsed & CW)

- Physics enables decoupling across gradient scales; Marble supports multiple functions. LFD can be cancelled out (feedforward / adaptive / ML), and the remaining microphonics — dominated by narrow-band frequency components — is therefore learnable [1-2].
- CW-mode results here are based on data from LCLS-II CM19-Cav2 [3-4].
- Data are available for all cavities (SLAC — Andy/Jorge's group), enabling broad studies and educational use.
- Preliminary but promising results — experimental demonstration is the goal for the coming years;



[1] Schappert, Warren. No. FERMILAB-CONF-16-742-TD. 2017.

[2] M. Liepe, et al., 2001 PAC, Chicago, IL, USA, 2001.

[3] A. Neumann et al., Phys. Rev. ST Accel. Beams 13, 082001 (2010).

[4] Bellandi, et al., arXiv preprint arXiv:2209.13896. (2022).

Summary: open source ML+LLRF substrate

- The foundation**

LLRF data is ML-ready (time-stamped, >kHz, synchronized); open Marble + DSP, portable across plant classes (lasers, SRF, qubits)

- Demonstrated — lasers @ BELLA**

Physics-guided ML: laser combining deterministically with fast convergence speed + pointing feed-forward → ~65% / 47% jitter reduction at TW

- In progress — SRF resonance control (PIP-II)**

Decouple Lorentz-force detuning from learnable microphonics; early VTS tests promising; cross-lab (FNAL, SLAC, ANL, MSU, JLab)

- Next**

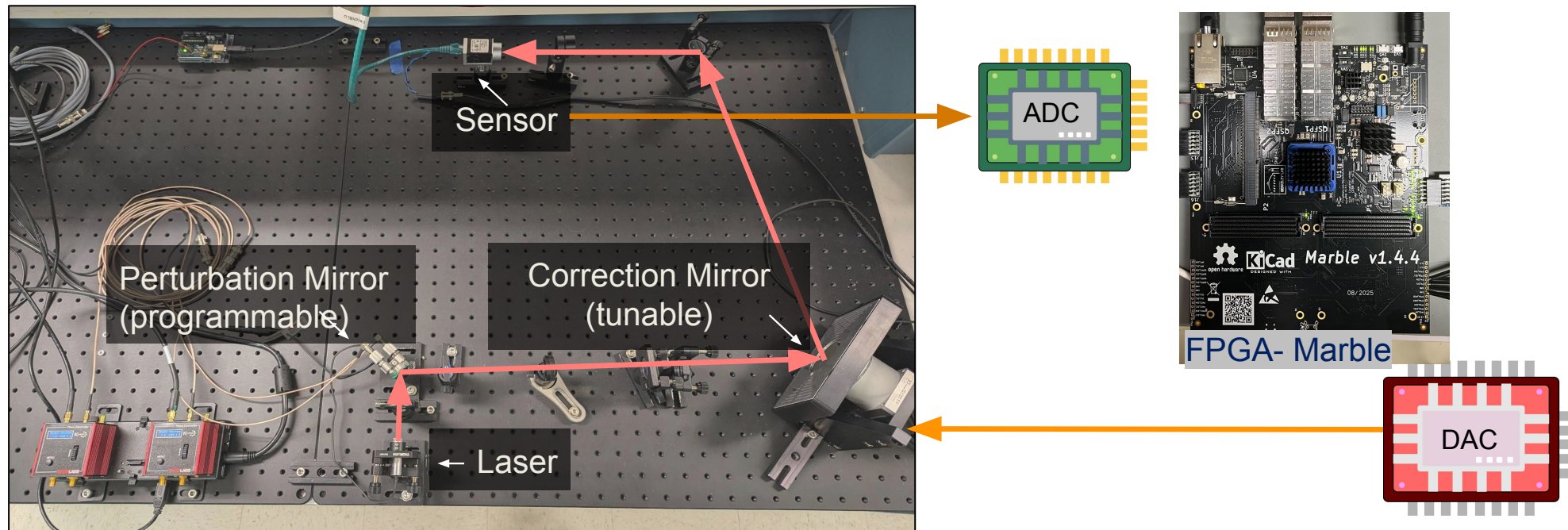
One open ML + LLRF substrate, many control problems:

Cross-lab collaboration can push it further into more useful, higher-impact tools for the community

Thanks!

hardware-AI: next step toward a more general scope

- Timing from FPGA is more accurate / stable → expect x2 further reduction for laser pointing
- Based on years of LBNL open-source Marble/common DSP development (used in LLRF, laser, quantum [1-3])
- A general-purpose testbed for control development, usable across many BES facilities: replay system noise, emulate different actuator dynamics, and validate algorithms before deployment.



[1] L. Doolittle, [et al.](https://github.com/BerkeleyLab/Marble); <https://github.com/BerkeleyLab/Marble>

[2] Q. Du, [et al.](#), Digital Low-Level Optical Control for Coherent Laser Combining, 2023/10/23 LLRF Workshop 2023

[3] G. Huang, [et al.](#), How a novel radio frequency control system enhances quantum computers, LBNL Lab News, 2021