



Image credit: Marguerite Tonjes

# Differentiable Simulation for Muon Cooling

## *AI4MuC mini-workshop*

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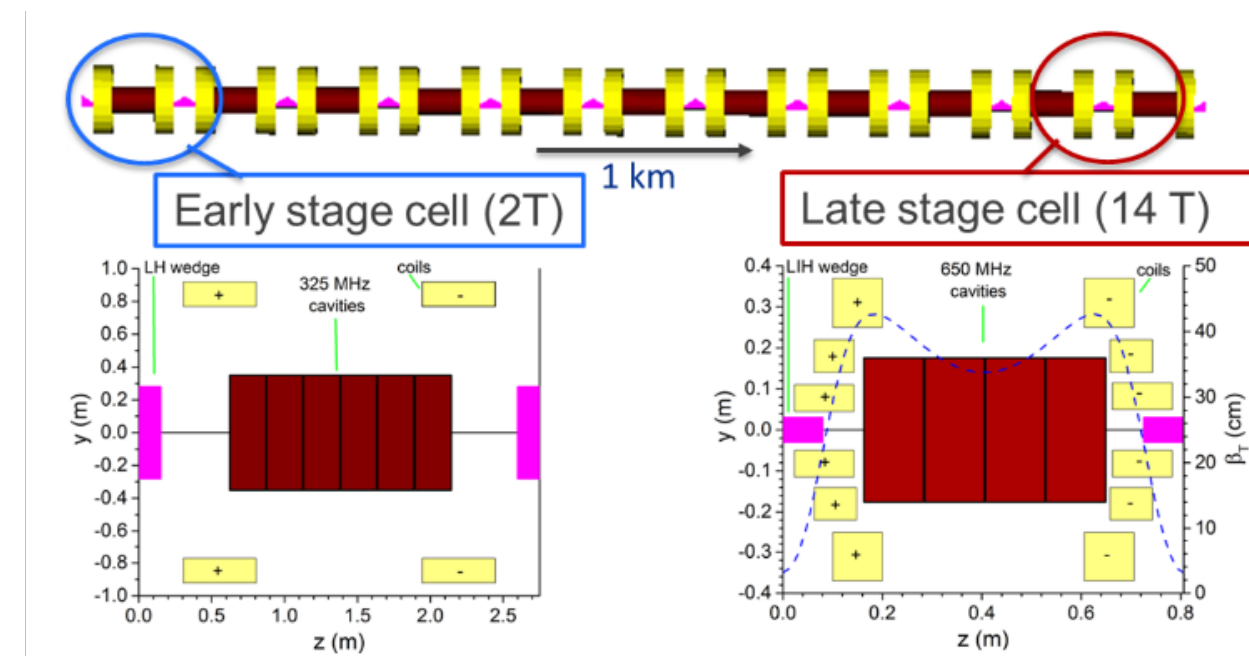
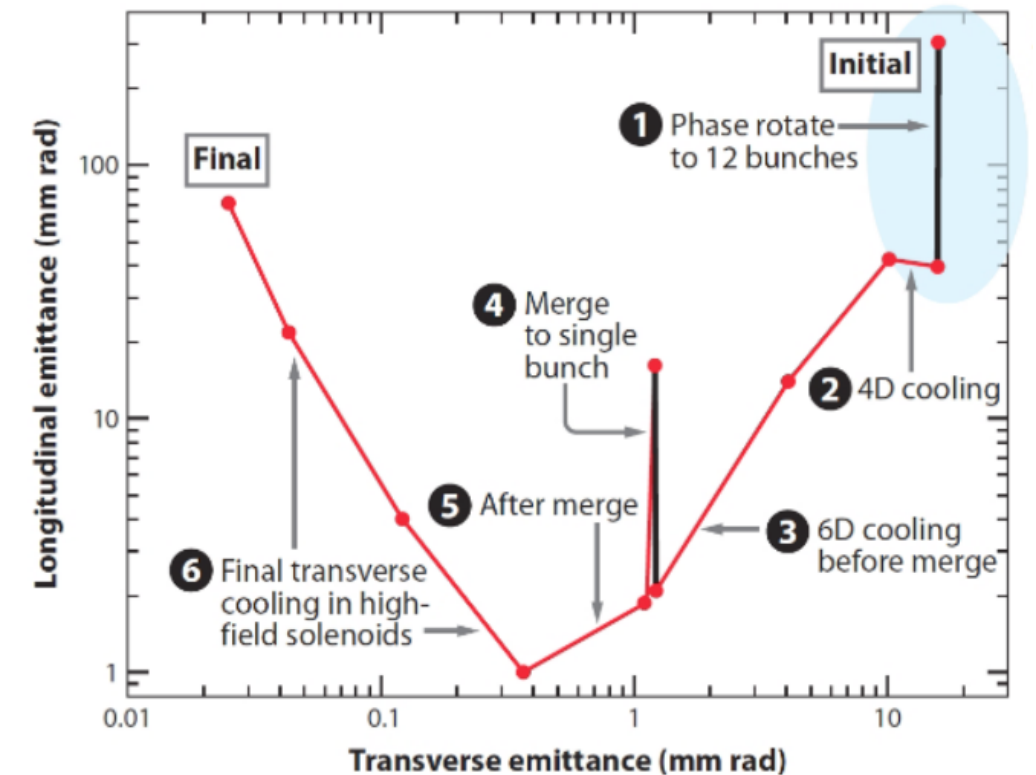
Eric Anderson, Corrinne Mills (UIC)

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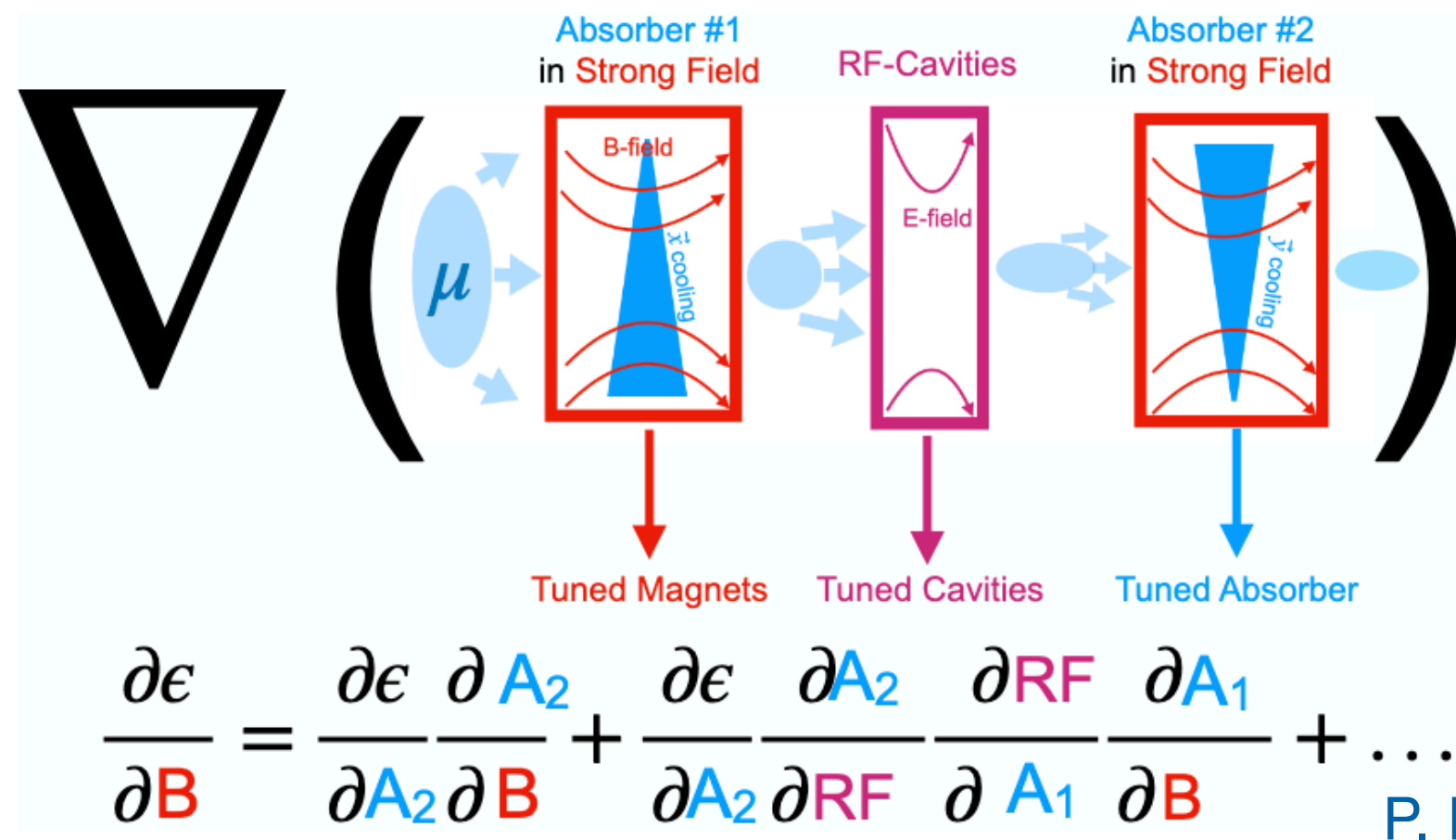
# The challenge

- A muon collider at FNAL would be cool, but the beam must be cooled
- Rectilinear design:  $10^5$  reduction in emittance through several stages in a 1km channel
  - Conservative design, hand-optimization
- Each of the 12 stages in the rectilinear ionization cooling channel has 20-100 cells, each of which has several tunable parameters
  - Optimal working point may differ in each stage, and improvements *compound*
- Simulation requires atypical mix of robust beam transport optics, material interaction, and collective beam dynamics



# The opportunity

- Large community interested in applying differentiable programming techniques to HEP problems, such as detector design and beam tuning
  - See e.g. [MODE workshops](#)
- Differentiable simulation enables efficient high-dimensional optimization
  - For our problem, a reasonable set of building blocks is understood!
    - TM010 pillbox, thick solenoid, wedge absorber, ... already a [benchmarking activity](#)



# The plan

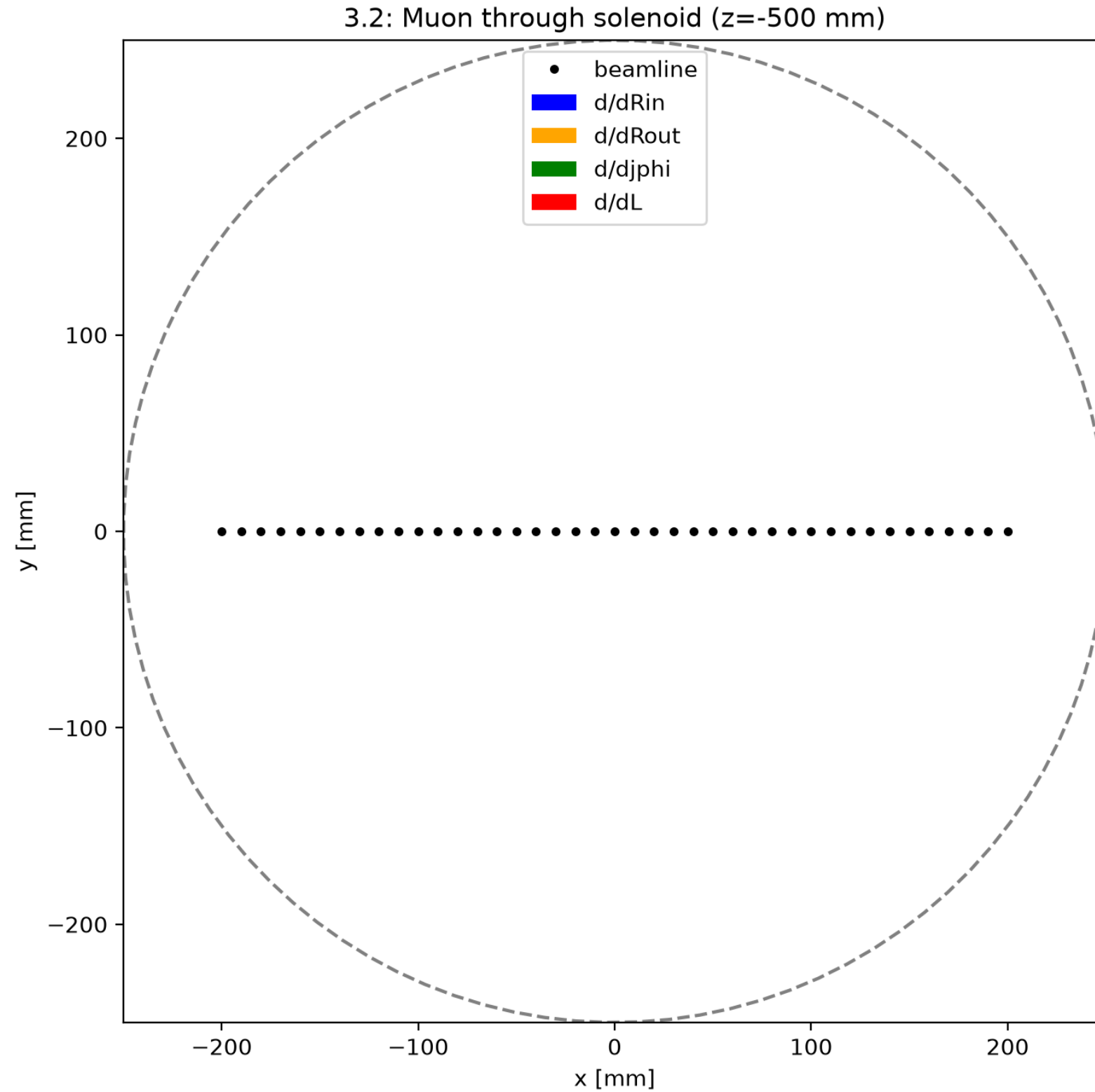
- There is plenty of prior art in both muon cooling simulators and differentiable beamline simulations to draw on
  - iCOOL, G4Beamline, BDSim, RF-Track, ...
  - Cheetah, WarpX, Synergia, Bmad-Julia, ...
- We are building a ground-up simulation in Jax
  - Lack of an existing tool at the intersection of all requirements
    - Optics + material interaction (+ collective effects) +  $\partial$
  - Desire to leverage a computational framework with first-class (automatic\*) differentiation support and GPU acceleration "for free"
  - **LLM coding tools change the calculus for refactor vs. rewrite**
    - At least for "Minimum Viable Product" stage
    - Once we understand the challenges, maybe we pivot?



\* AD is not always correct nor always the fastest!

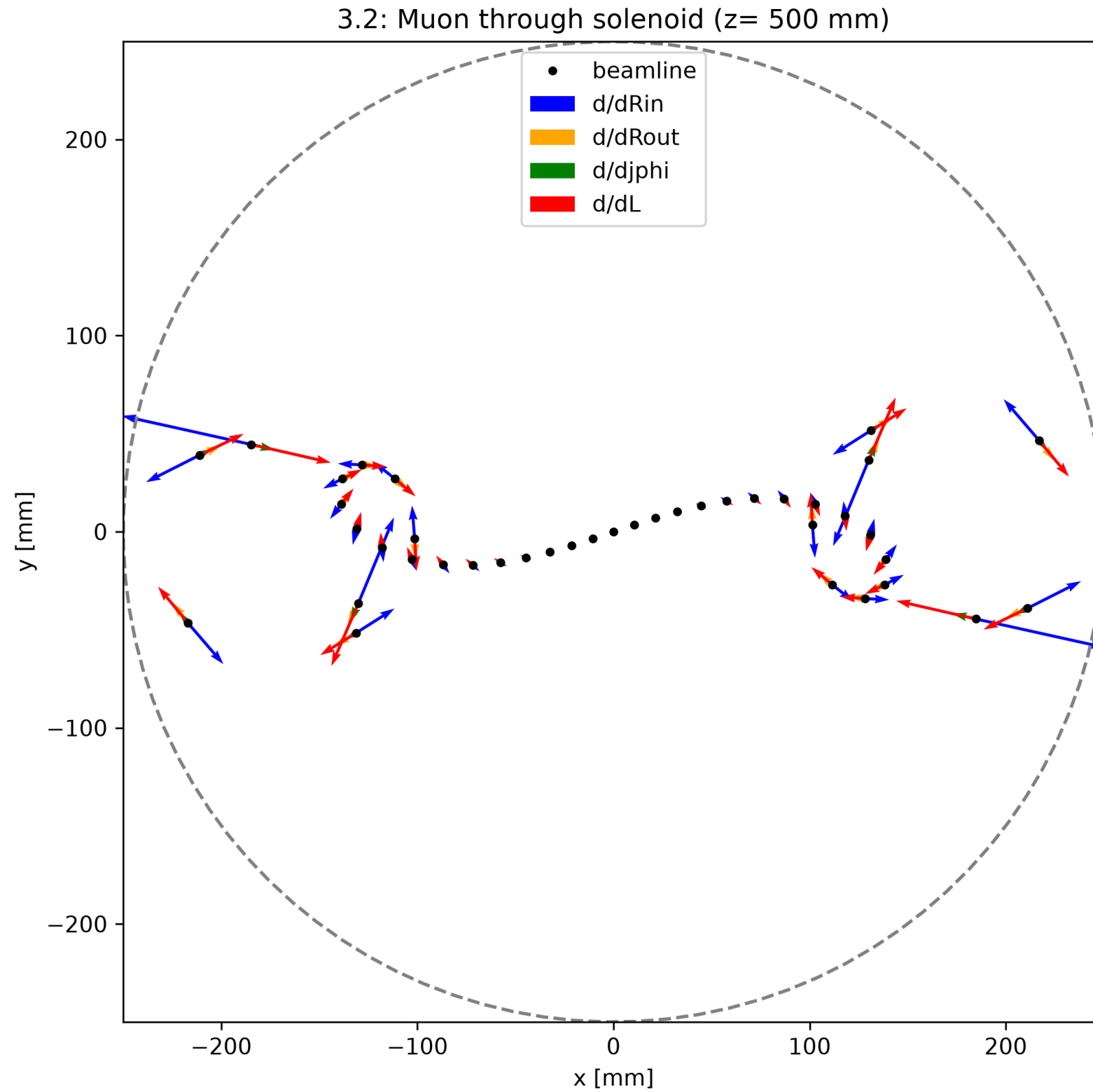


# Progress



## Benchmark 3.2: thick solenoid

# Progress

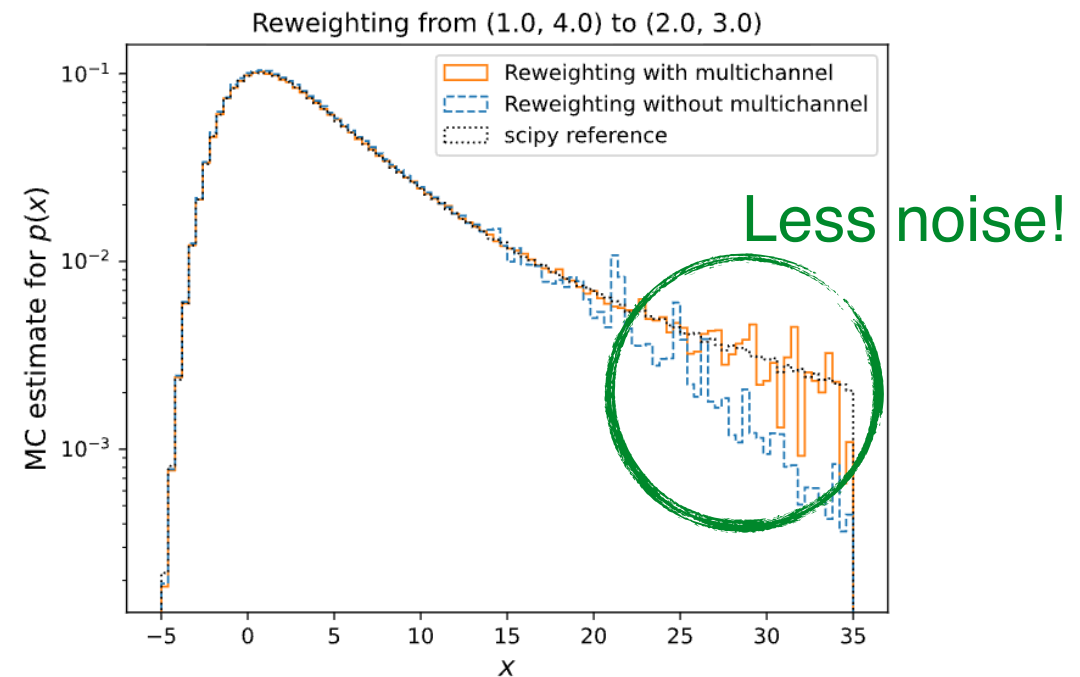
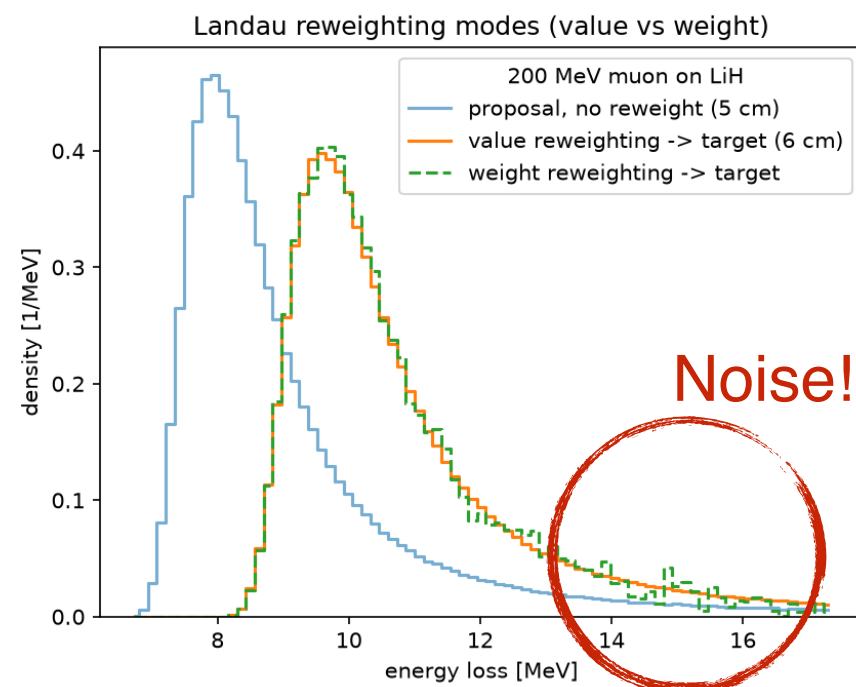


## Benchmark 3.2: thick solenoid



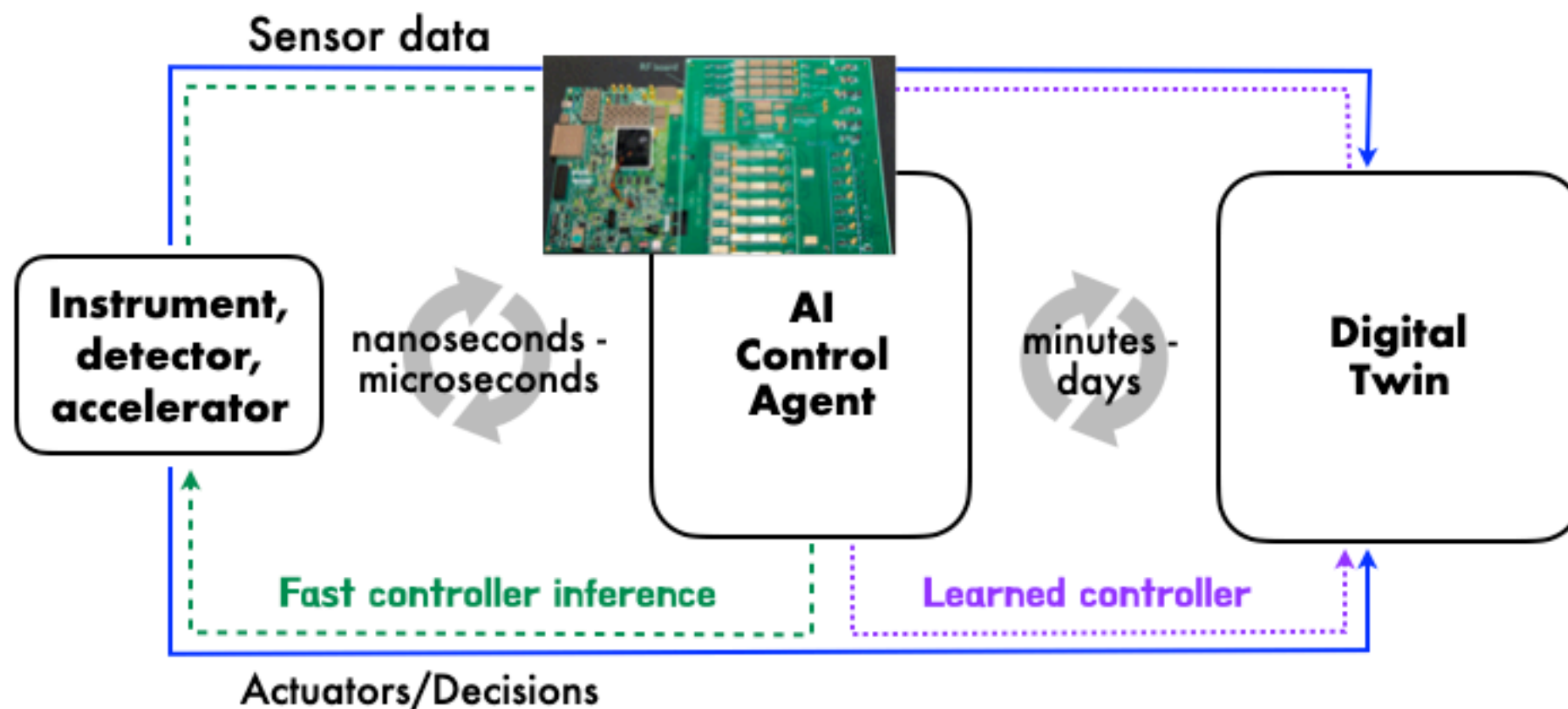
# Stochastic effects

- The hard part is taking gradients through stochastic elements
  - Absorber: energy straggling, multiple scattering
- The objective is emittance, extracted from moments of 6D distribution  $p(x, \dot{x})$ 
  - $\epsilon_\theta(t) \sim \int dx d\dot{x} f(x, \dot{x}) p_\theta(x, \dot{x}; t) = \int dx d\dot{x} f(x, \dot{x}) \int^t dt' \frac{dp_\theta(x, \dot{x})}{dt'}$
  - Sample  $(x, \dot{x}) \sim p(x, \dot{x}; t = 0)$  and push forward  $\Rightarrow$  weighted samples  $(x_\theta, \dot{x}_\theta, w_\theta)$
  - Two paths for  $\theta$  gradients: through the final positions, or through the sample weights
  - **The key for robust gradients is efficient sampling**



# Future tie-ins

- A differentiable simulator becomes a very useful component of a digital twin for online beam controls
  - Easier to train NN to capture delta between simulation and reality than a black box!
  - To do so effectively, “classical component” needs high-dimensional gradients
  - Here we can tie in with [MOAT project](#)



N. Tran

# Summary

- A differentiable simulator for muon cooling channels is being developed
- Stay tuned for future updates