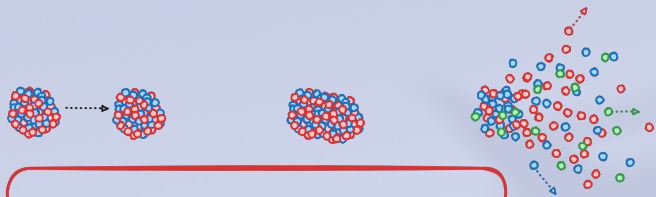


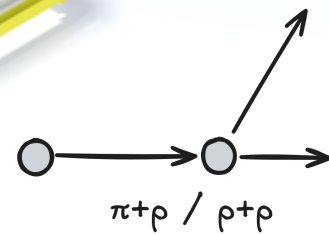
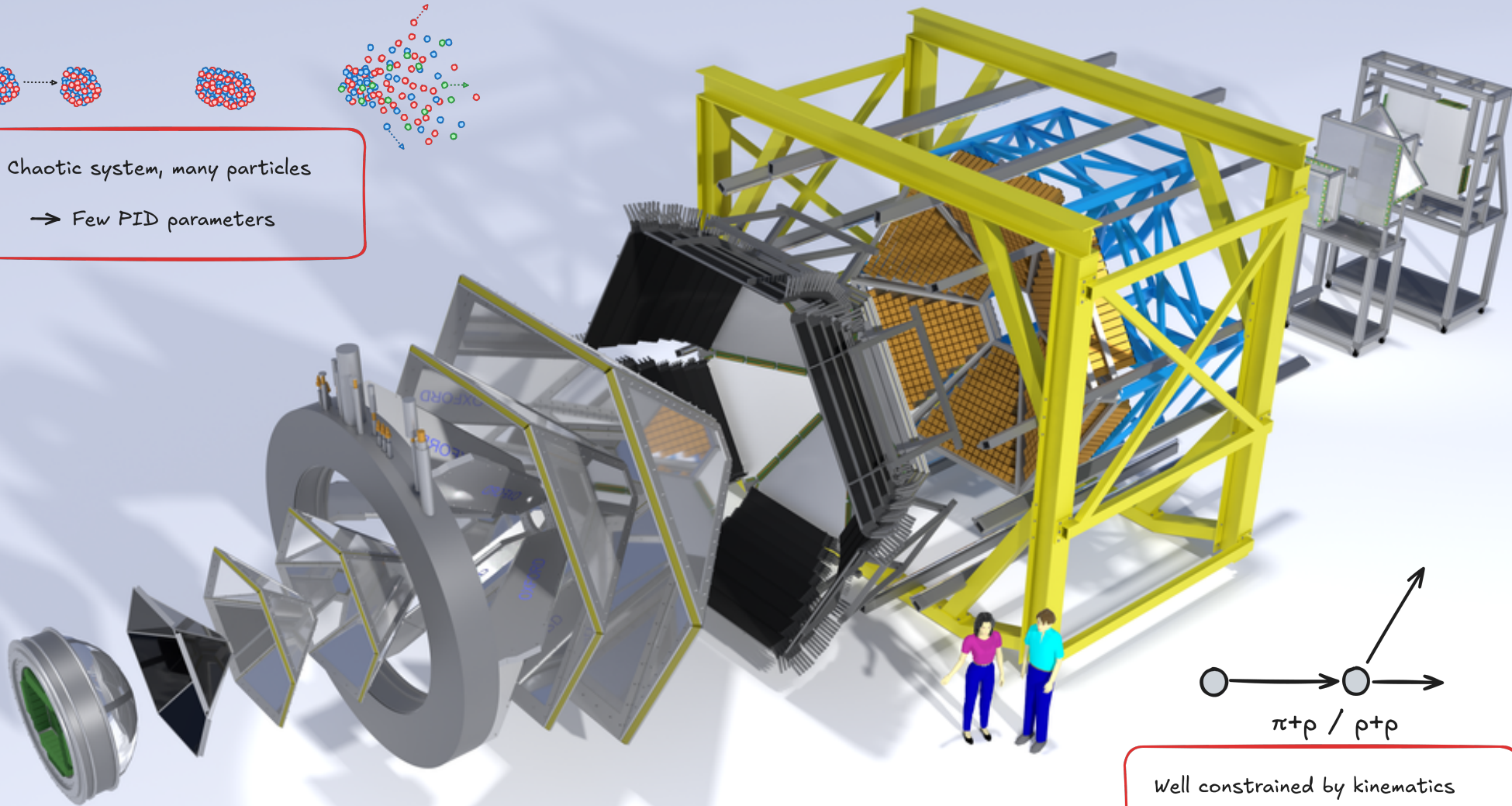
Well constrained by kinematics

→ Many PID parameters



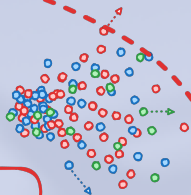
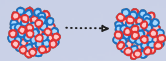
Chaotic system, many particles

→ Few PID parameters



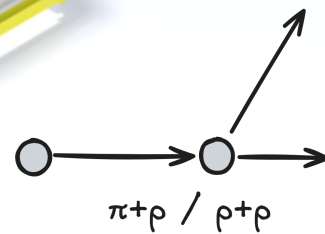
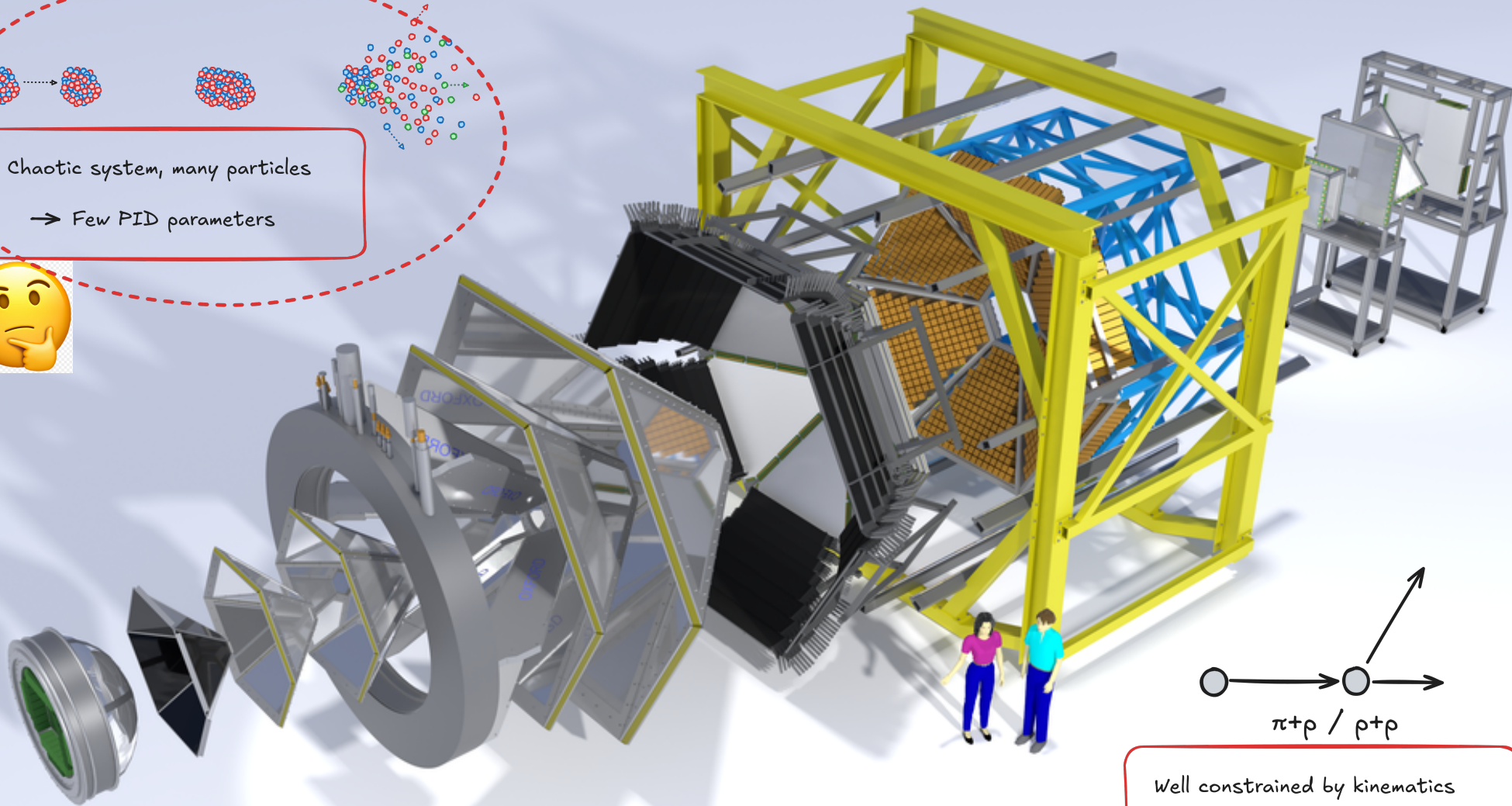
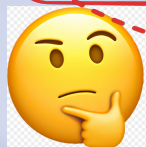
Well constrained by kinematics

→ Many PID parameters



Chaotic system, many particles

→ Few PID parameters

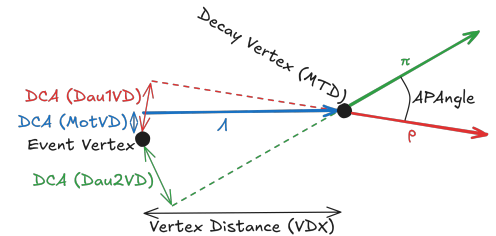
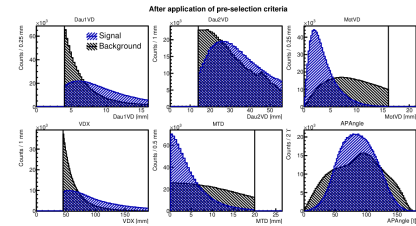


Well constrained by kinematics

→ Many PID parameters

Existing PID Methods

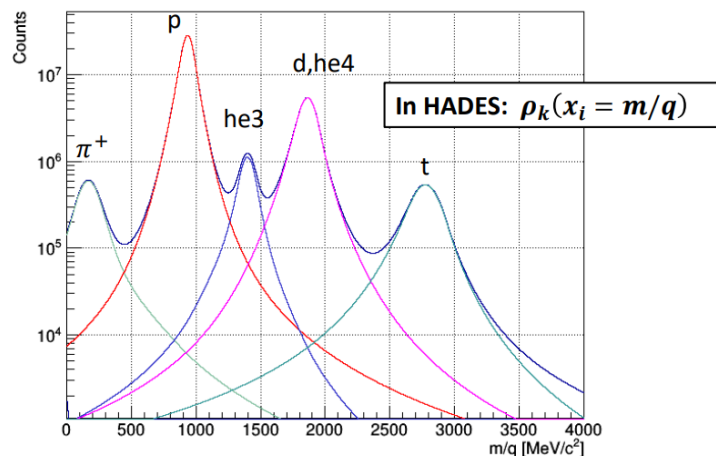
Decaying Particles



Multi-Layer Perceptron

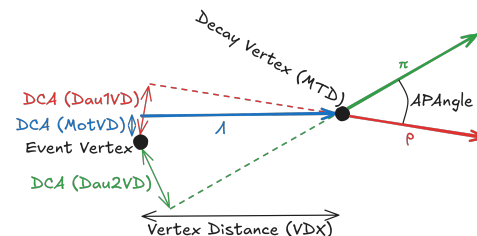
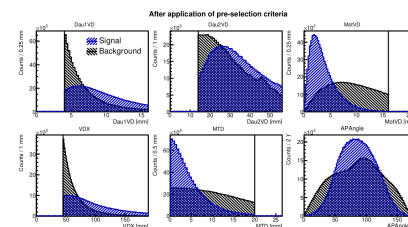
Existing PID Methods

Abundant Particles



Bayesian Methods

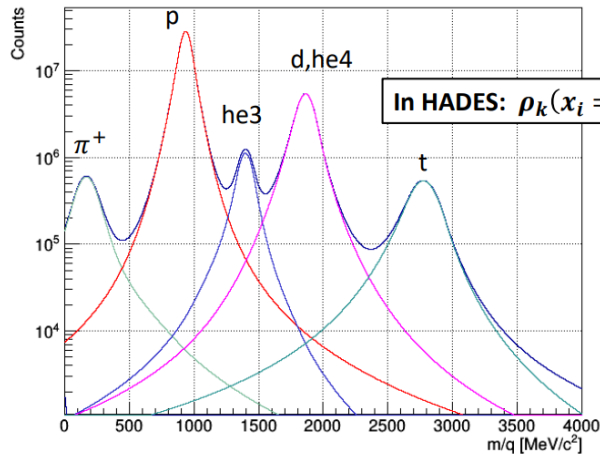
Decaying Particles



Multi-Layer Perceptron

Existing PID Methods

Abundant Particles



Bayesian Methods

Rare Particles????

Physics Informed Loss Function

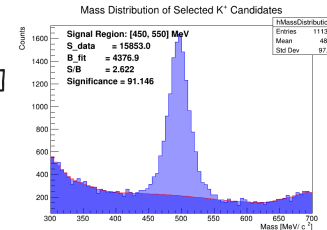
[Input Features]

- dE/dx (energy loss)
- momentum (p)
- χ^2 (track quality)

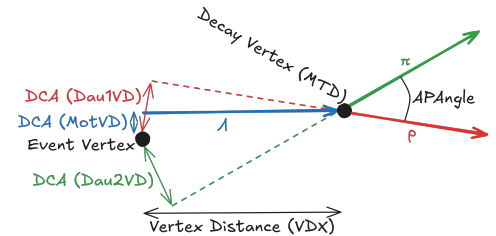
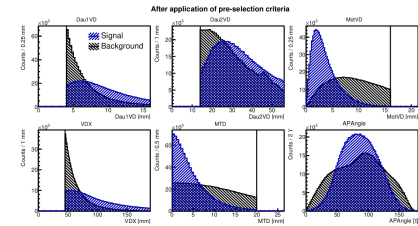
Neural Network $\rightarrow K^+ / \text{not } K^+$

[Physics-Informed Loss]
(Bethe-Bloch)

Compare predicted class
vs expected dE/dx(p)



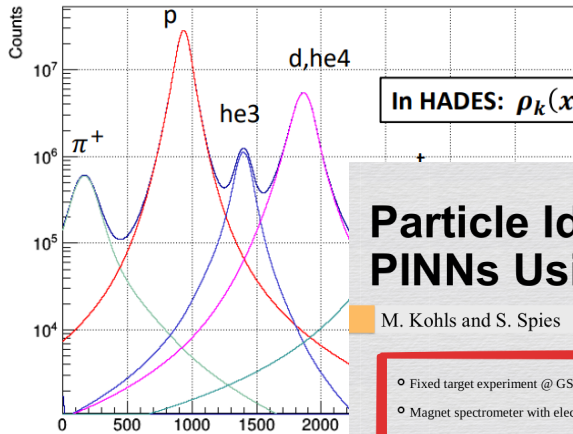
Decaying Particles



Multi-Layer Perceptron

Existing PID Methods

Abundant Particles



In HADES: $\rho_k(x_i = m/q)$

Rare Particles????

Physics Informed Loss Function

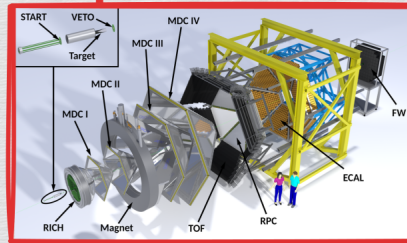
[Input Features]
— dE/dx (energy loss)

Particle Identification with MLPs and PINNs Using HADES Data

M. Kohls and S. Spies

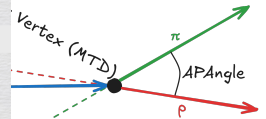
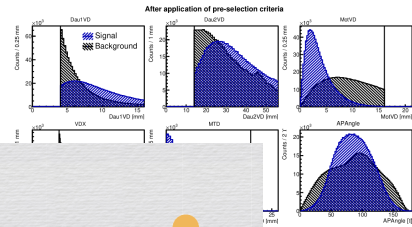
- Fixed target experiment @ GSI, Darmstadt, Germany
- Magnet spectrometer with electromagnetic calorimeter
- Heavy ion experiments (Ag+Ag, Au+Au)
 - Average particle multiplicities per event ~ 100
 - Complex systems with kinematically overlapping channels
 - Identification limited to measured final state properties and their resolution
- Elementary reactions ($p+p$, $\pi+p$)
 - Average particle multiplicities ~ 3
 - Kinematically well-defined channels
 - Identification via measured final state properties and kinematic constraints

High Acceptance Di-Electron Spectrometer



Current Methods

- Decaying (off-vertex) particles [1]:
 - MLP (FFNN) training on geometric variables and kinematic constraints
 - Simulation of decaying particles as signal
 - "Mixed Event" technique for data-driven combinatorial background emulation
- "Stable" charged particles:
 - Currently user-defined criteria (heavy ion) combined with kinematic constraints (elementary)
 - Studying the performance of Physics Informed Neural Networks (PINNs) for this task is hoped to improve the PID



Distance (VDX)

Perceptron

Bayesian Metho

PIDANN Setup

MLPs for Off-Vertex Particles