

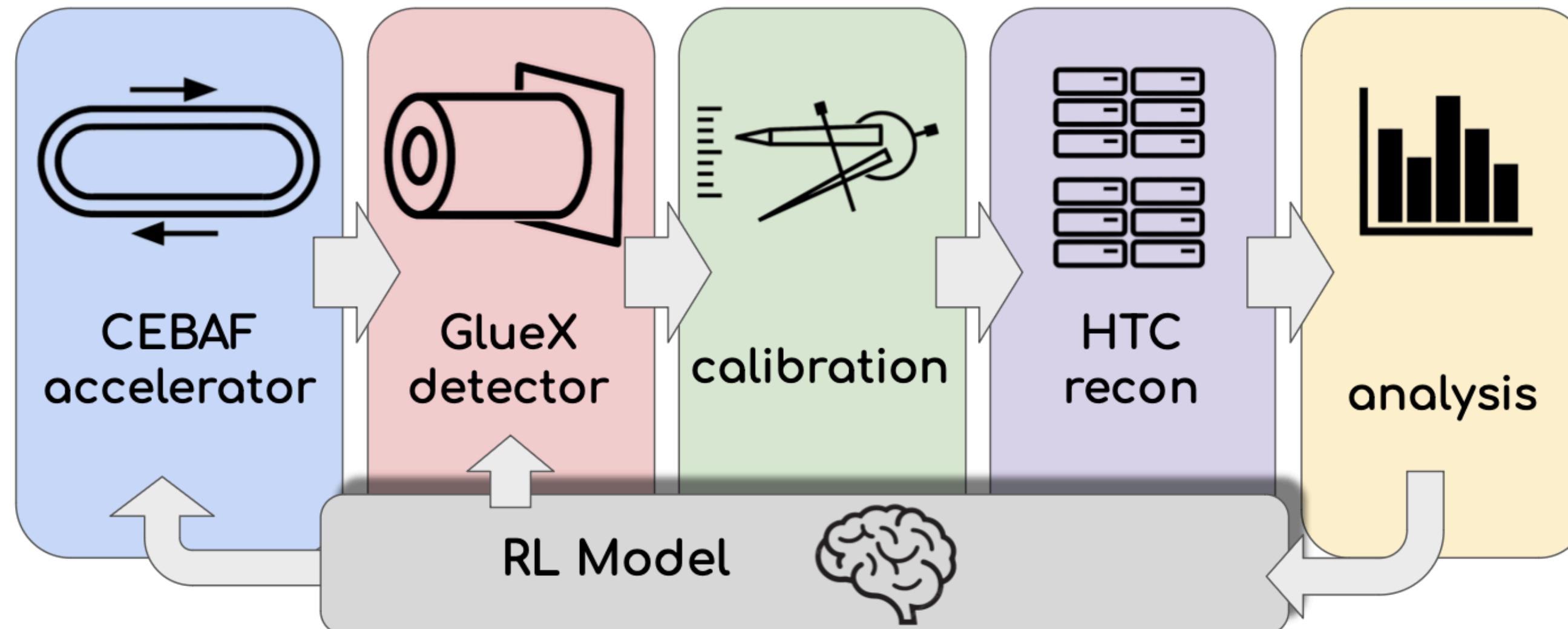
Real-Time Physics Analysis using AI Track Reconstruction Online

LDRD Proposal

G.Gavalian (Jefferson Lab)



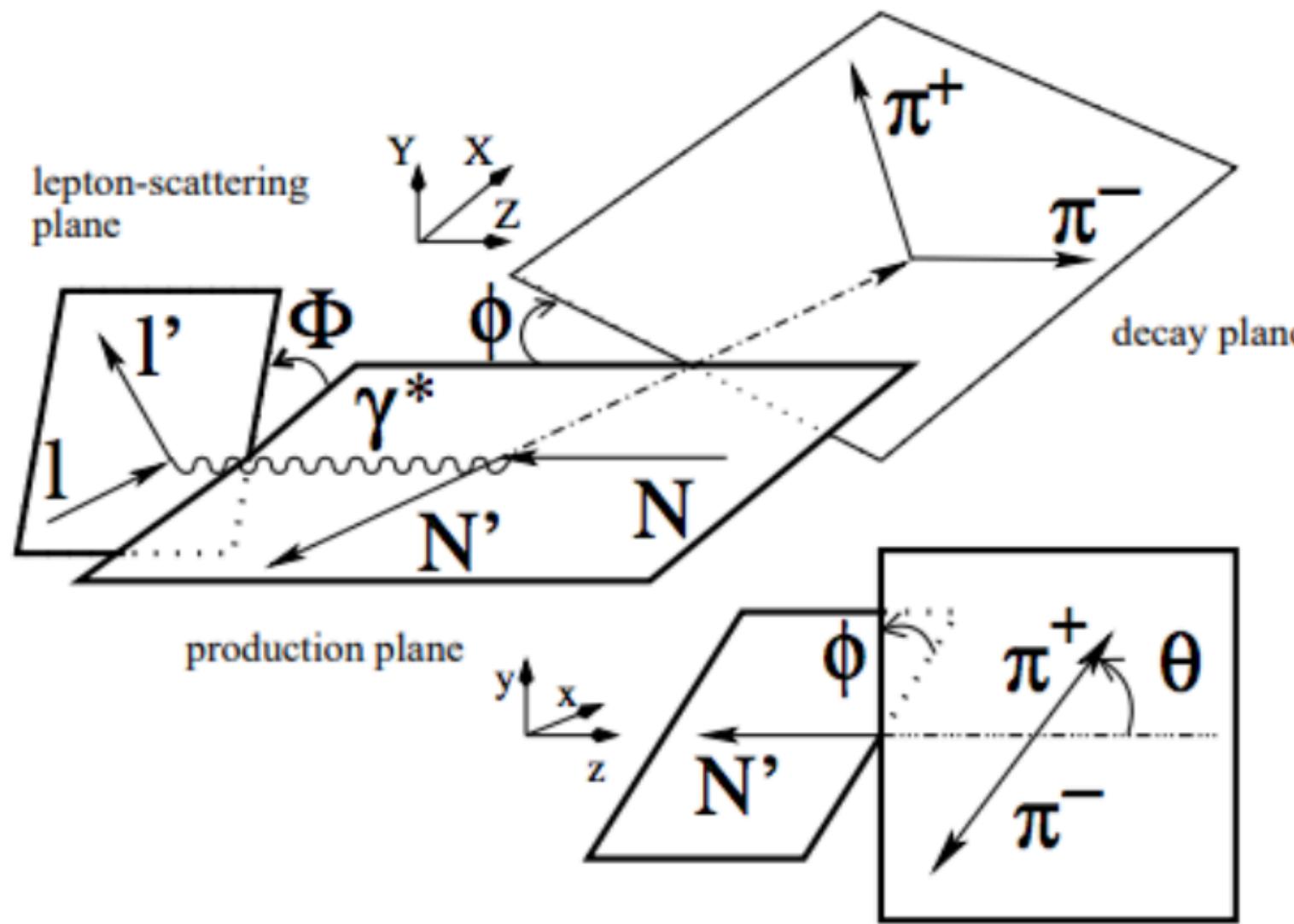
Idea



- Automated Experiments
 - Do Online Physics Analysis
 - Measure relevant Observables
 - Provide feedback to the AI to suggest changes in beam parameters (current, energy, etc...)
- The feedback is provided every hour

- The proposal includes all many experiments:
 - Hall-B CLAS12
 - GlueX (Hall-D)
 - Hall-A

Idea



$$\mathcal{W}^U(\Phi, \phi, \cos\Theta)$$

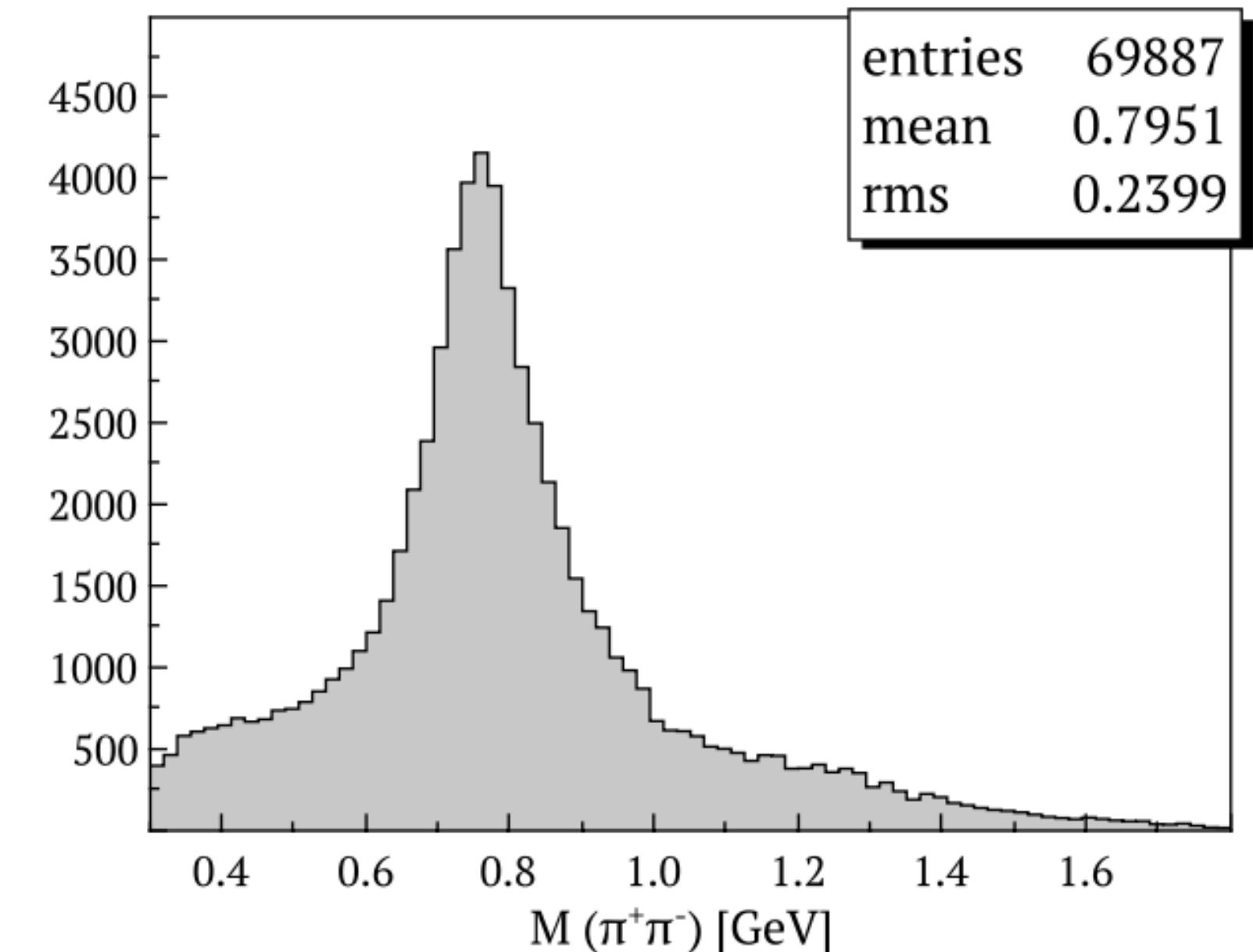
$$\begin{aligned}
 &= \frac{3}{8\pi^2} \left[\frac{1}{2}(1 - r_{00}^{04}) + \frac{1}{2}(3r_{00}^{04} - 1)\cos^2\Theta \right. \\
 &\quad - \sqrt{2}\text{Re}\{r_{10}^{04}\}\sin 2\Theta \cos \phi - r_{1-1}^{04}\sin^2\Theta \cos 2\phi \\
 &- \epsilon \cos 2\Phi \left(r_{11}^1 \sin^2\Theta + r_{00}^1 \cos^2\Theta \right. \\
 &\quad - \sqrt{2}\text{Re}\{r_{10}^1\}\sin 2\Theta \cos \phi - r_{1-1}^1\sin^2\Theta \cos 2\phi \\
 &- \epsilon \sin 2\Phi \left(\sqrt{2}\text{Im}\{r_{10}^2\}\sin 2\Theta \sin \phi \right. \\
 &\quad \left. \left. + \text{Im}\{r_{1-1}^2\}\sin^2\Theta \sin 2\phi \right) \right] \\
 &+ \sqrt{2\epsilon(1+\epsilon)} \cos \Phi \left(r_{11}^5 \sin^2\Theta + r_{00}^5 \cos^2\Theta \right. \\
 &\quad - \sqrt{2}\text{Re}\{r_{10}^5\}\sin 2\Theta \cos \phi - r_{1-1}^5\sin^2\Theta \cos 2\phi \\
 &+ \sqrt{2\epsilon(1+\epsilon)} \sin \Phi \left(\sqrt{2}\text{Im}\{r_{10}^6\}\sin 2\Theta \sin \phi \right. \\
 &\quad \left. \left. + \text{Im}\{r_{1-1}^6\}\sin^2\Theta \sin 2\phi \right) \right].
 \end{aligned}$$

- Challenge:
 - How do you analyze experimental data at data-taking within one hour?
 - DAQ rate is 16 kHz; the reconstruction rate is 1 Hz on a single CPU

- Hall-B

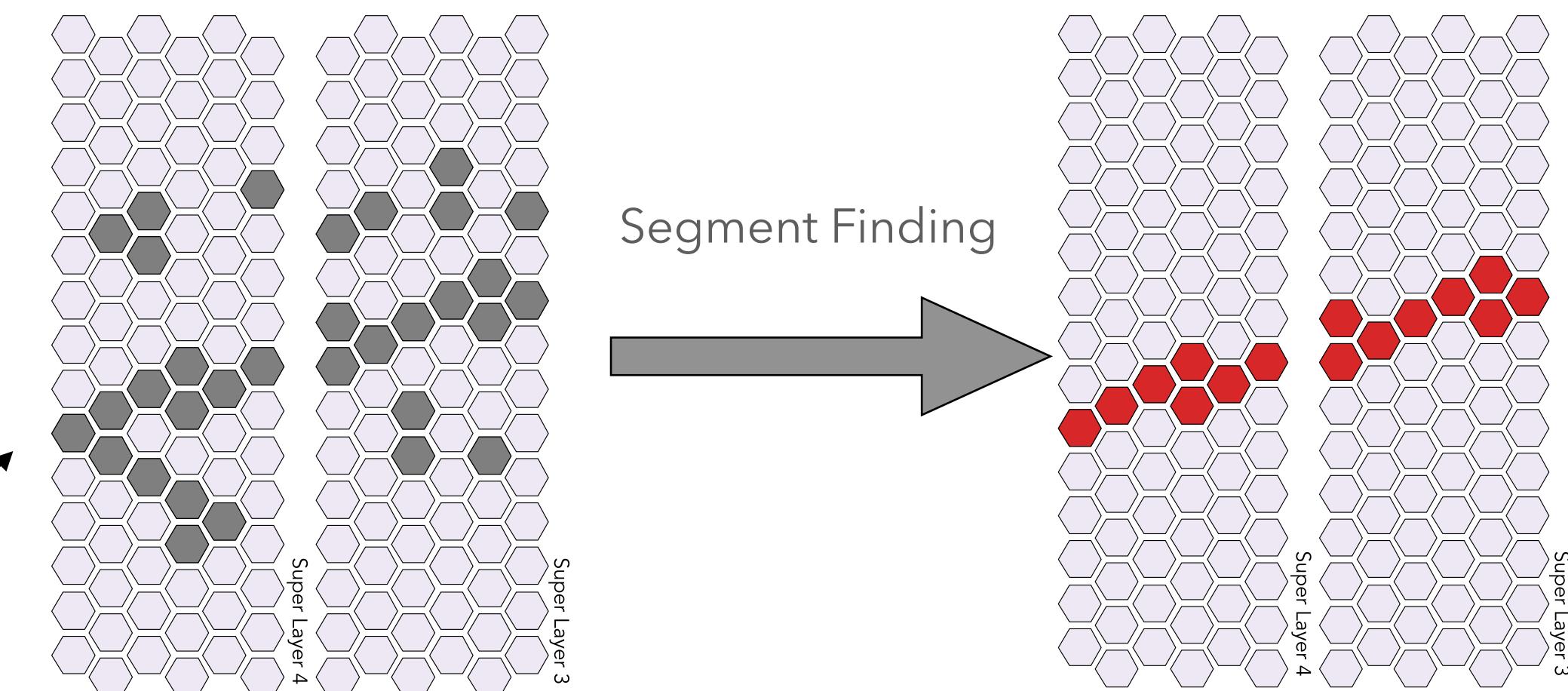
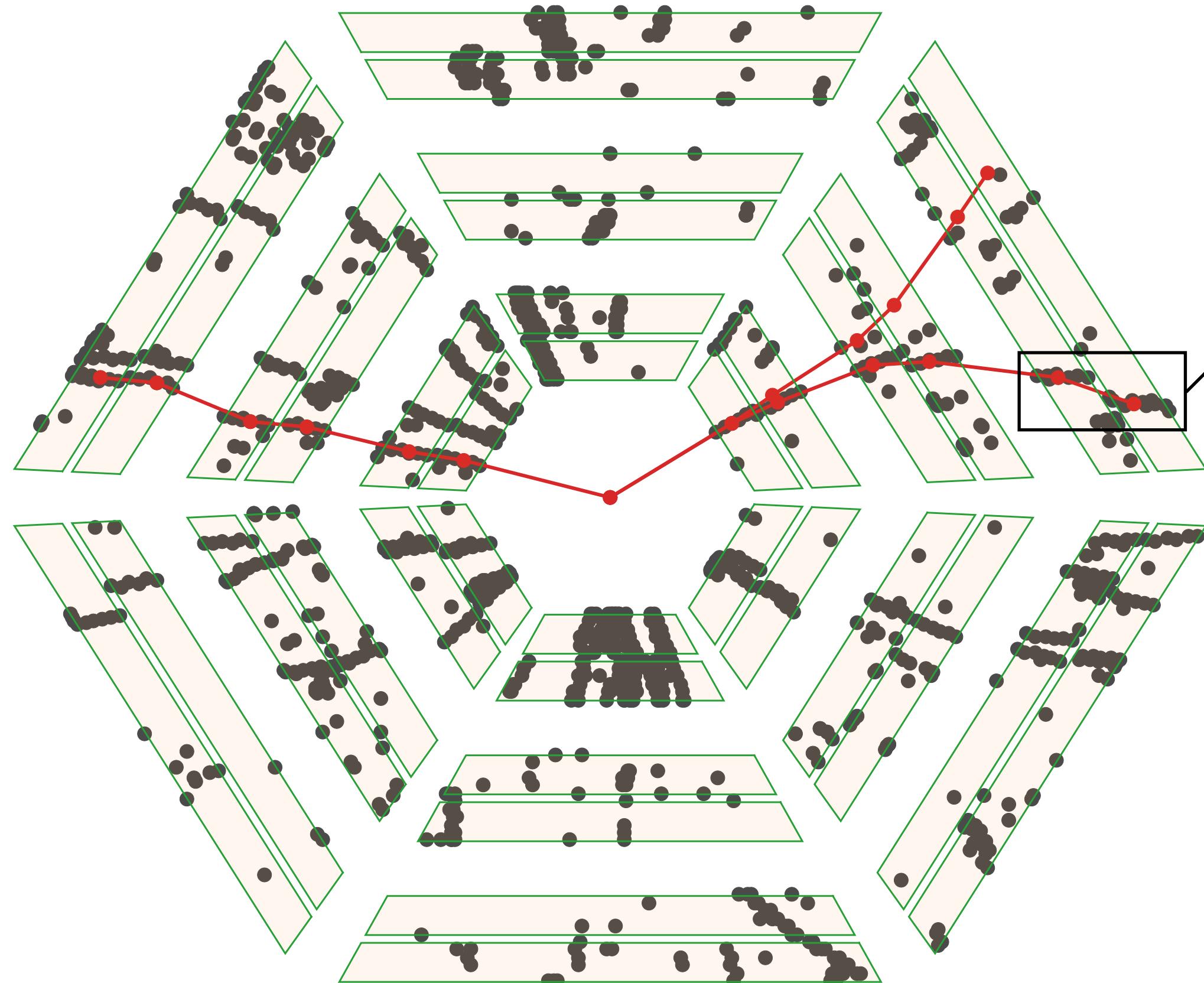
- Study rho meson
- Identify relevant observables that are impacted by beam conditions
- Provide feedback to the accelerator to change the beam conditions.

Statistics for one hour



Segment Finding (to be developed)

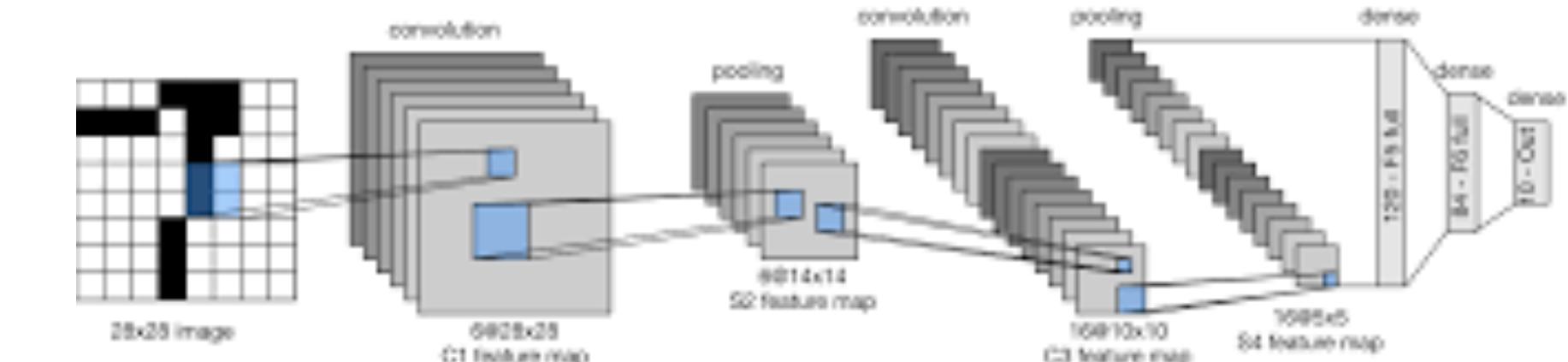
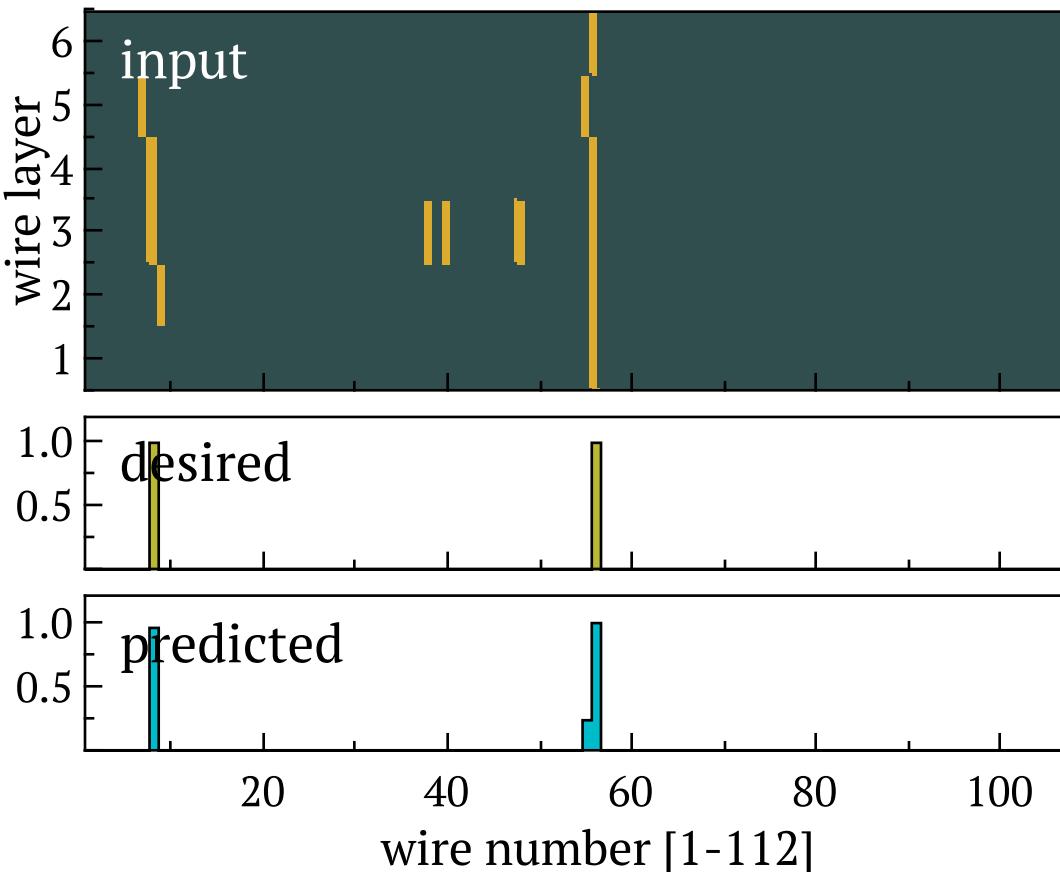
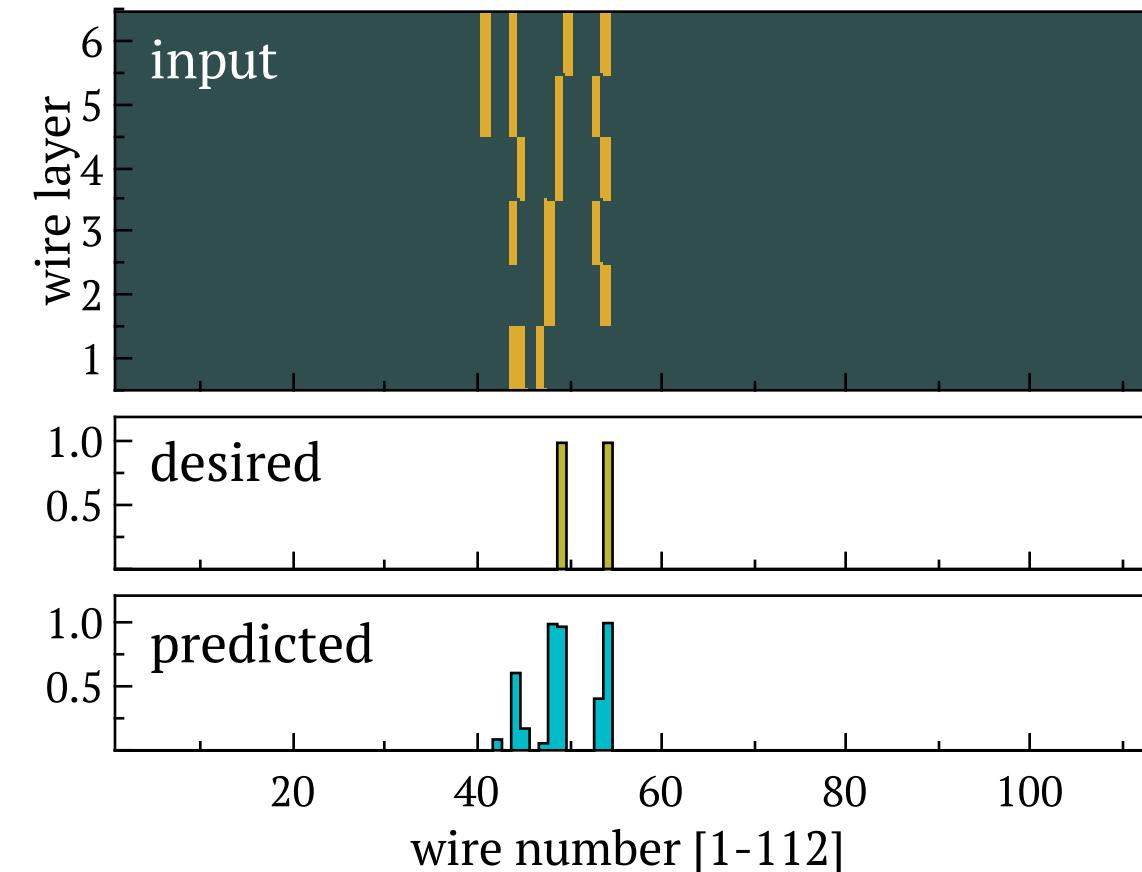
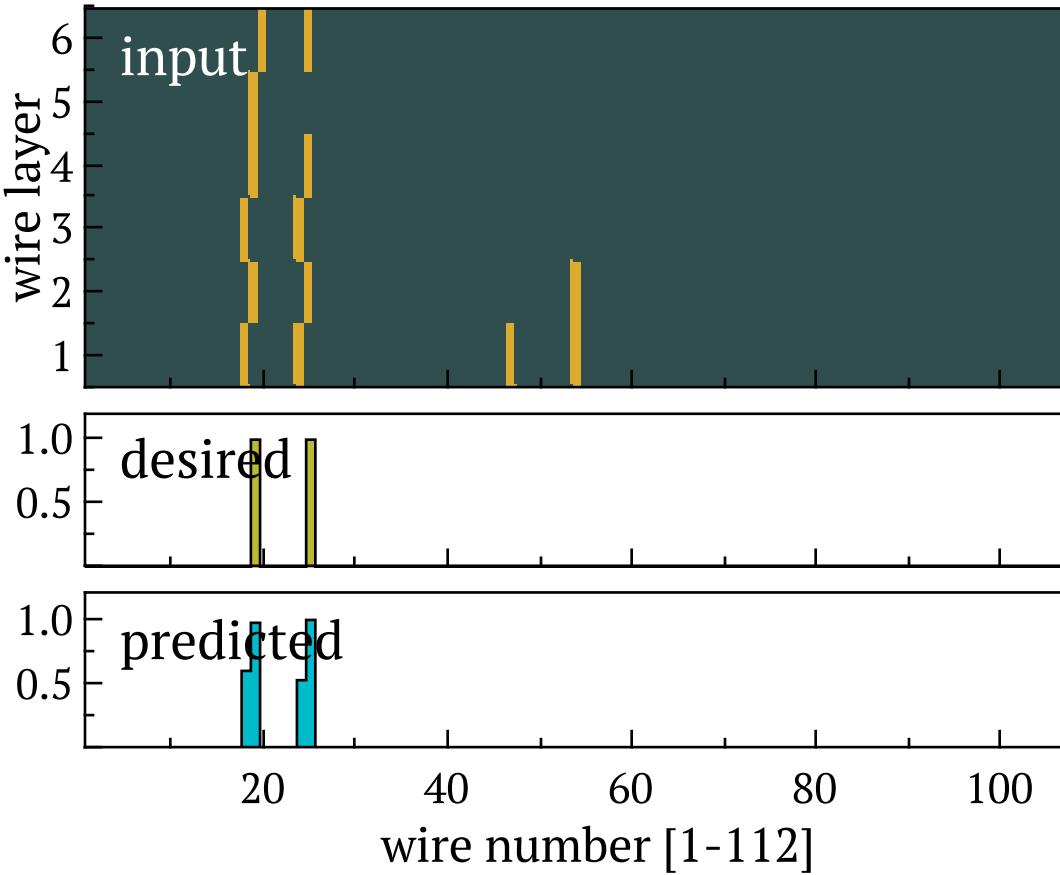
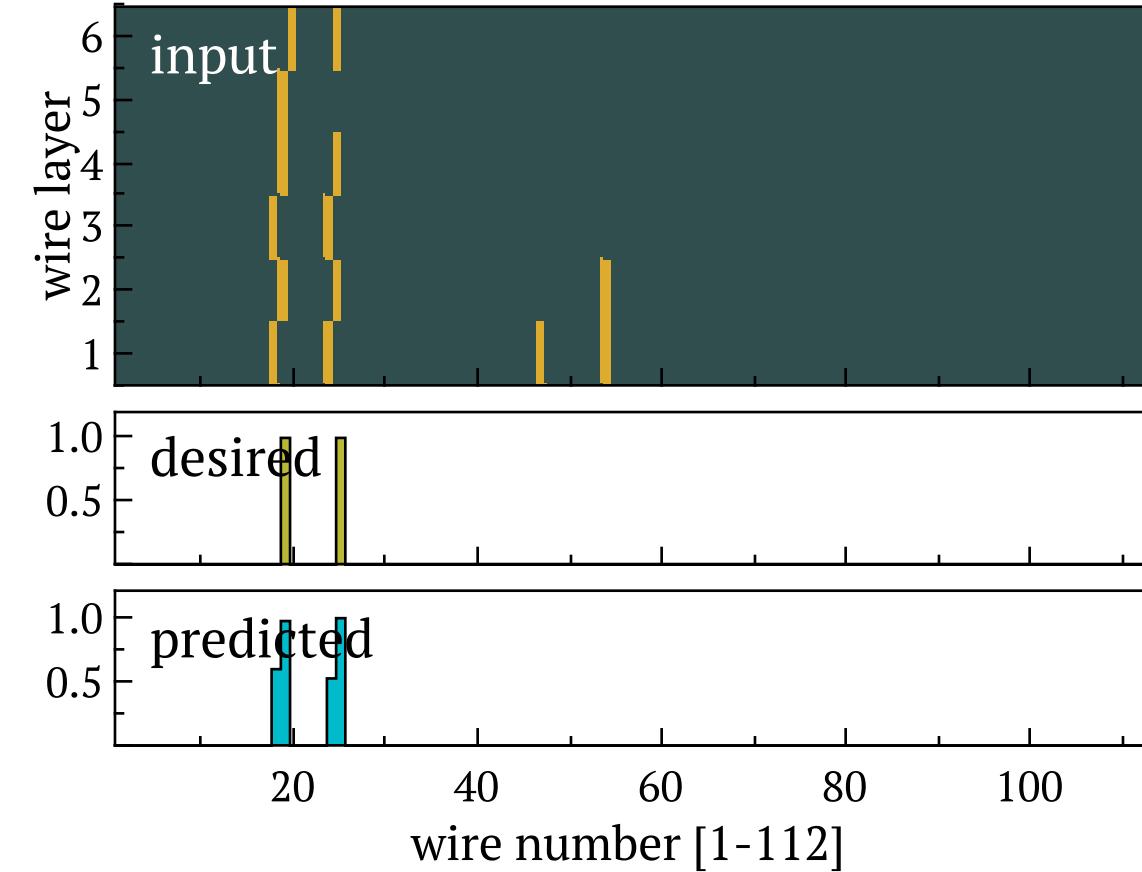
- 6 sectors with 6 chambers in each sector (called super-layers)
- 6 wire planes in each super layer with 6-degree tilt relative to each other, (112 wires in each plane)



Need to develop fast AI based segment finding algorithm
 (The reconstruction segment finder is 6 Hz)
 Not suitable for online

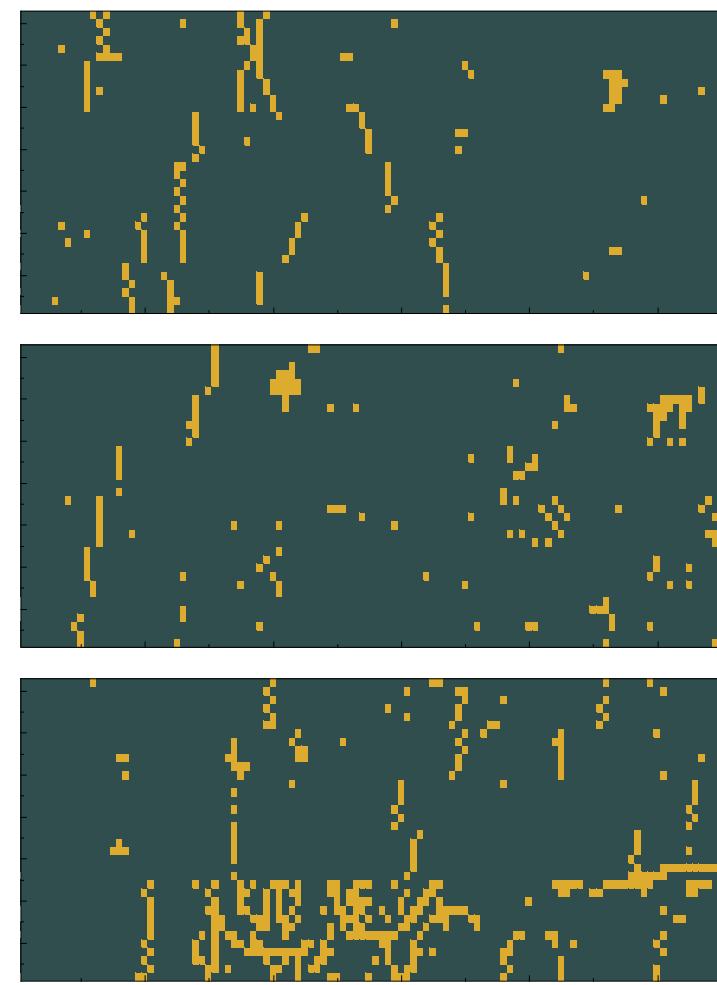
- Find segments in each super layer (remove noise)
- Combine 6 segments (one from each super layer) to make a list of possible tracks
- Identify correct combinations of segments that represent a track

Feasibility tests

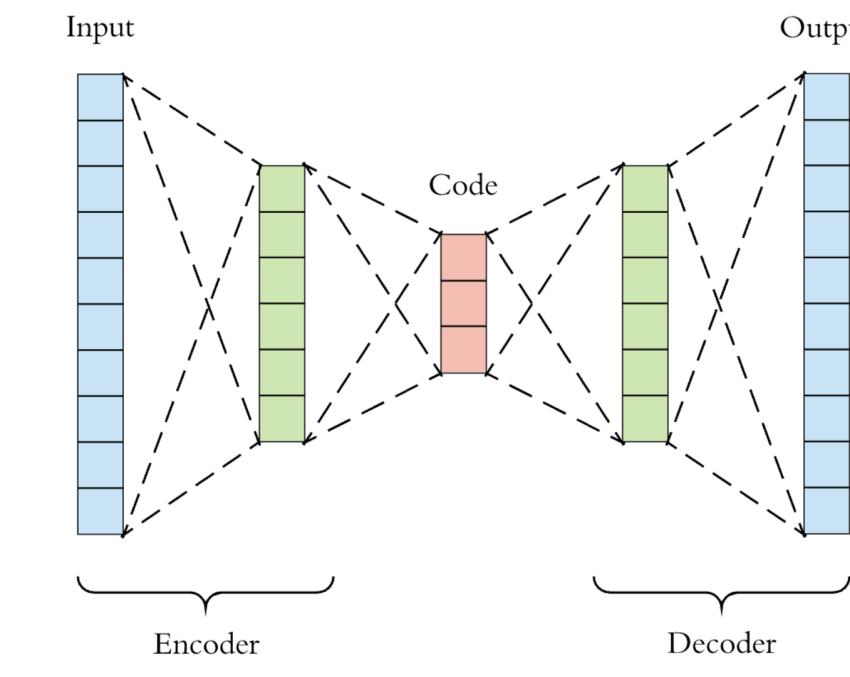


- Clustering Algorithm:
 - Convolutional Neural network with logistic regression to identify possible cluster positions.
 - Currently at 80% efficiency
- Proposed work:
 - Investigate algorithms to identify clusters fast and with higher efficiency

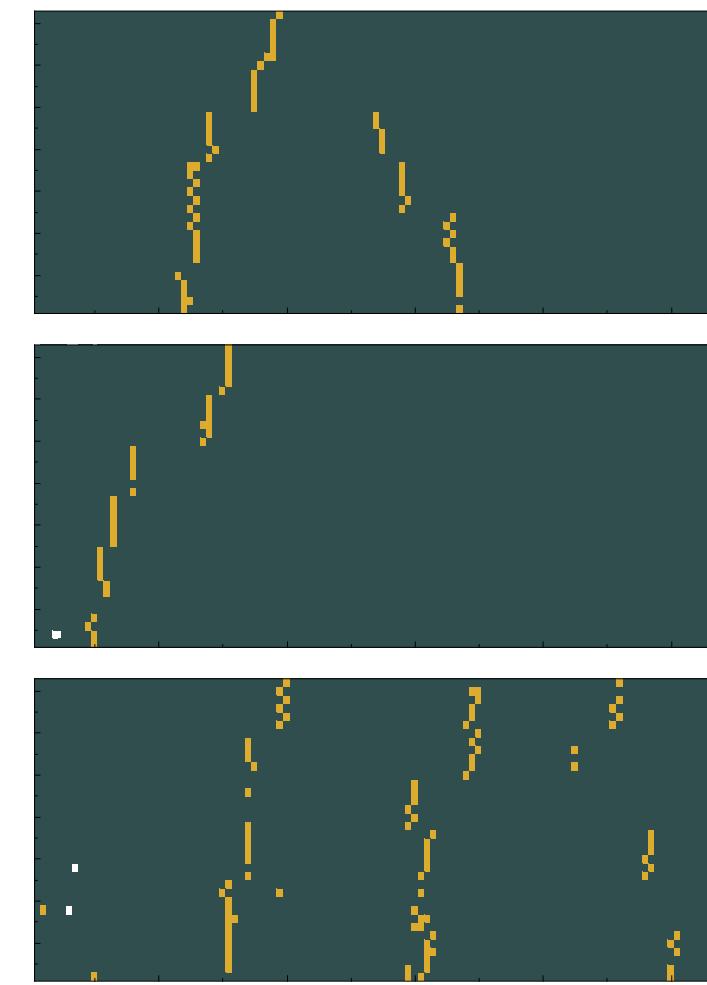
What do we have now



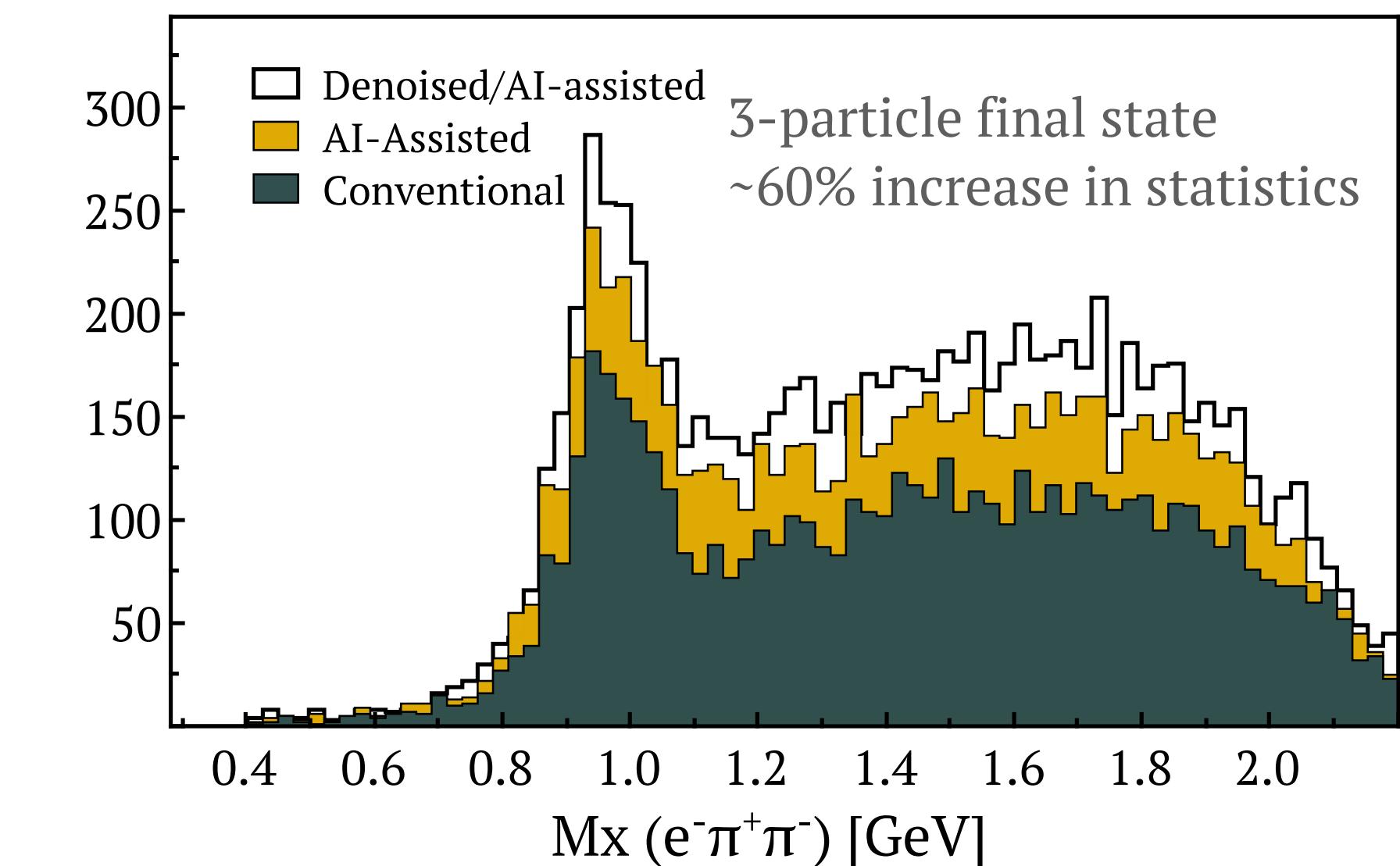
De-Noising CNN Auto-Encoder



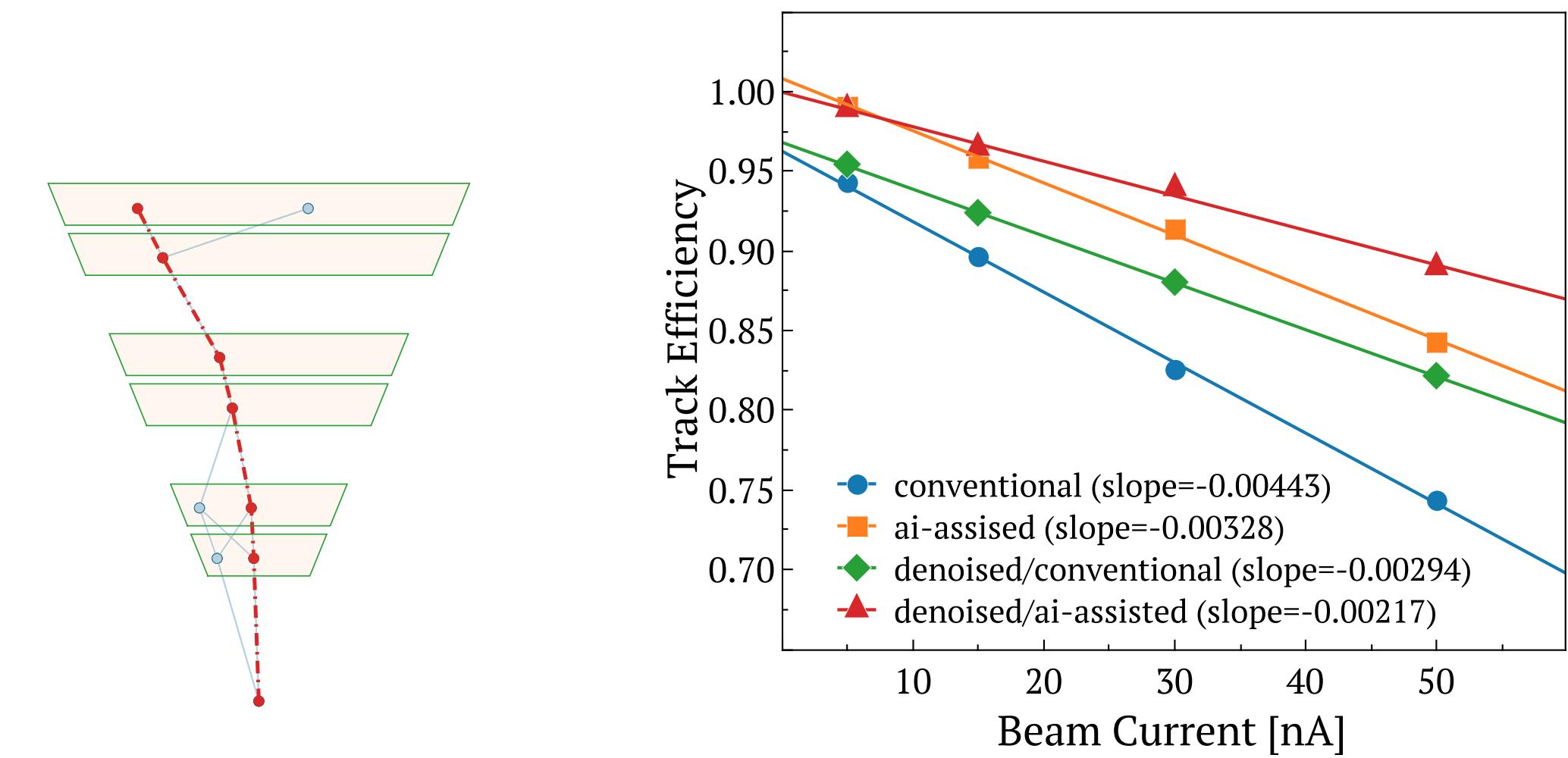
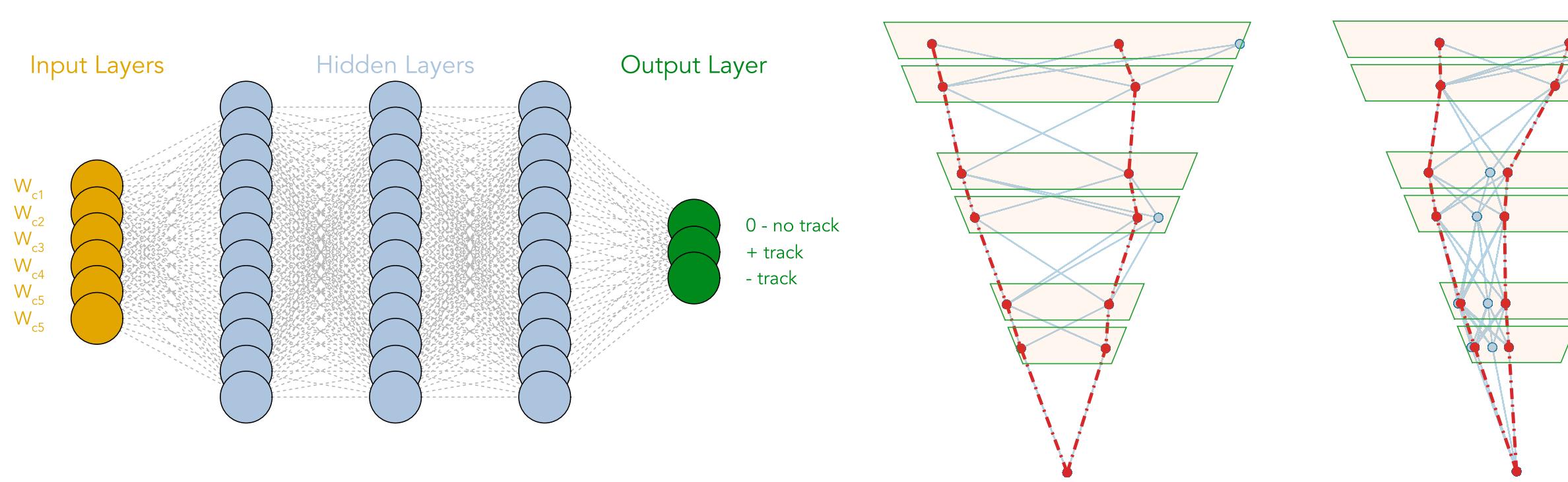
Retains >95% of hits belonging to tracks, and removes >90% of background hits



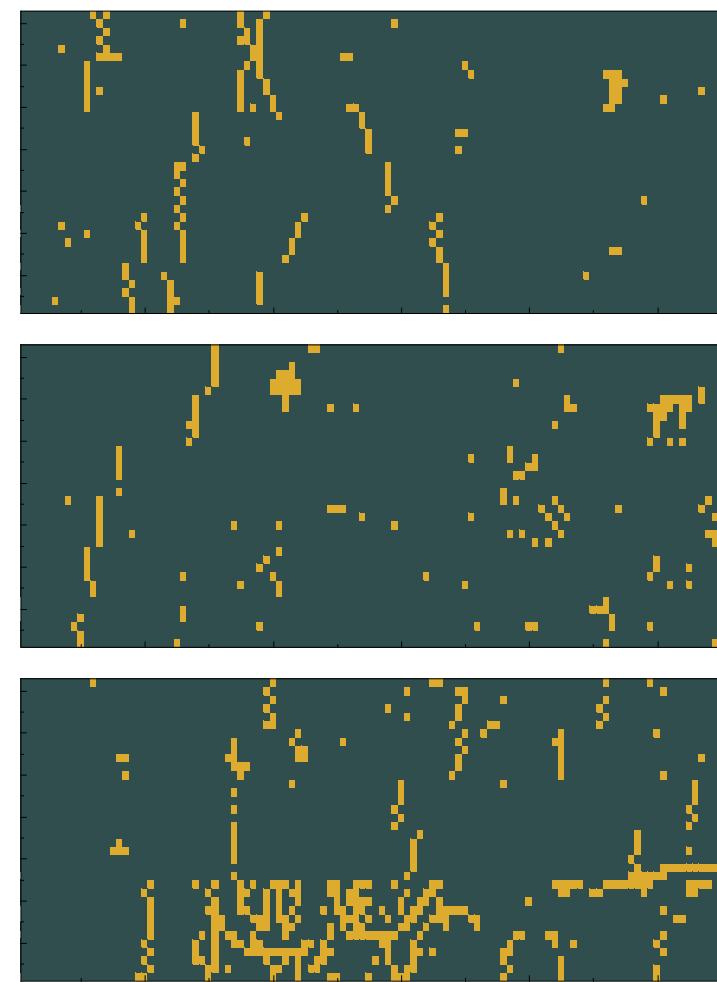
CLAS12 Track Reconstruction with Artificial Intelligence
Gagik Gavalian (Jefferson Lab), Pet all e-Print: 2205.02616 [physics.ins-det]



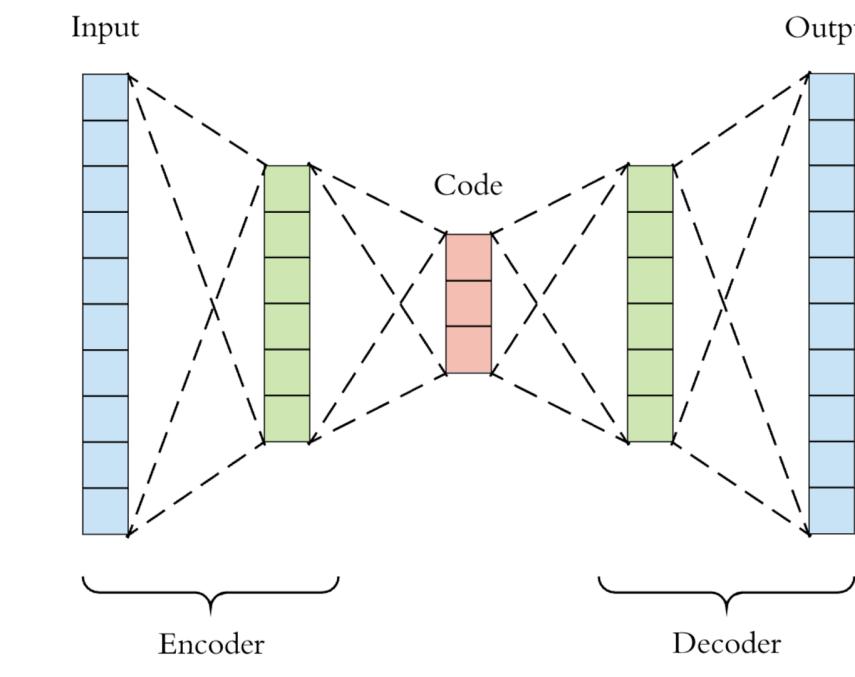
The Classifier network identifies tracks from segment combinations and identifies track charges. The AI-assisted track identification increased tracking efficiency by **15%-21%** (depending on luminosity)
Improvement of the efficiency slope as a function of luminosity.



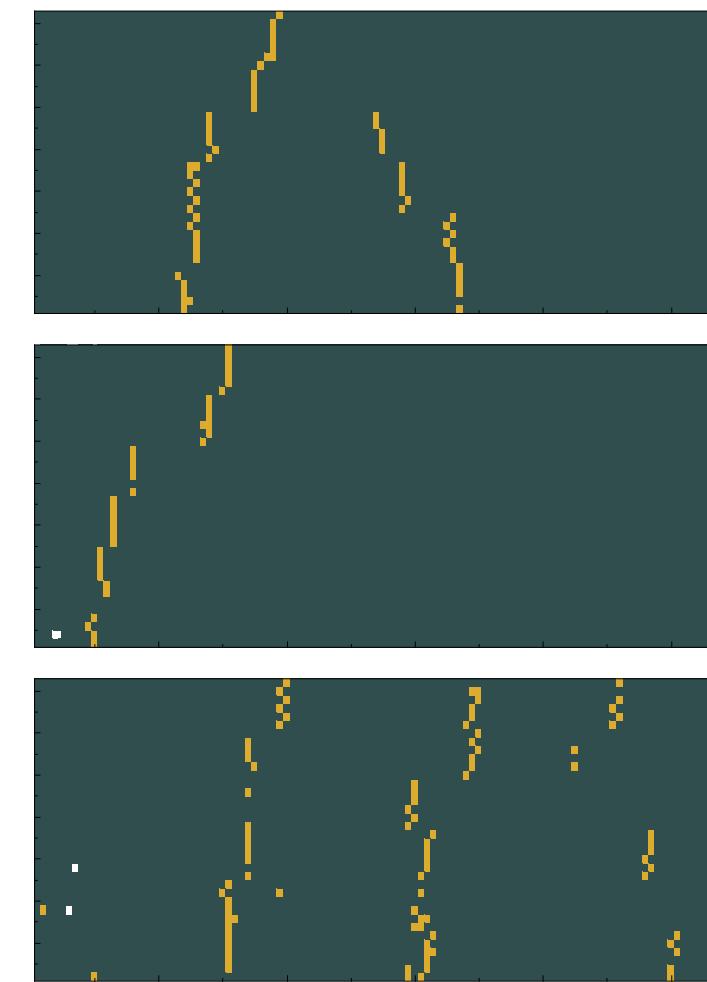
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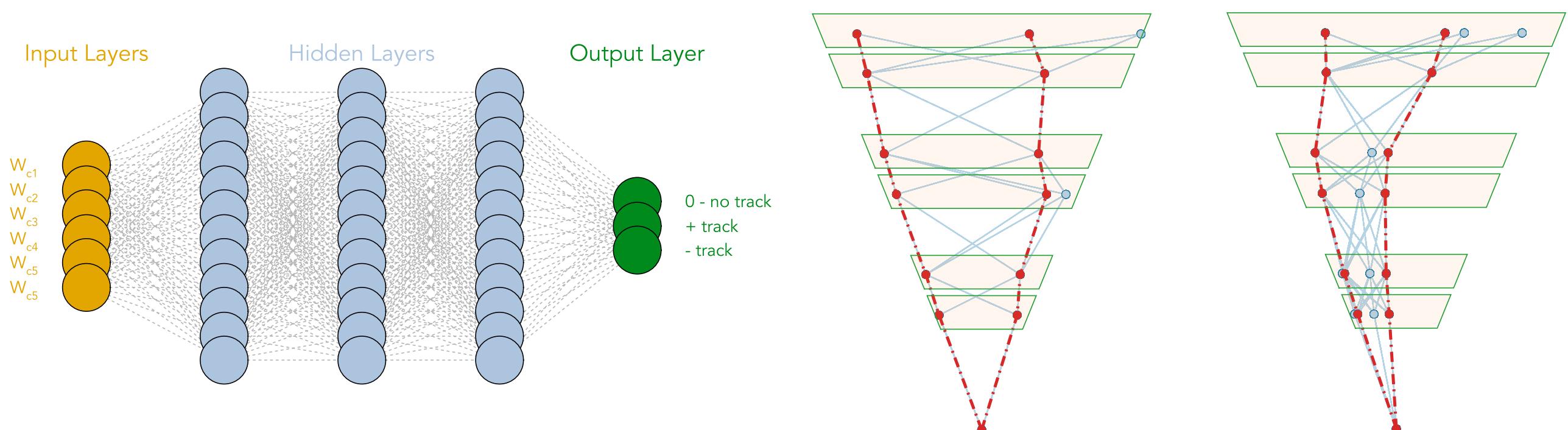
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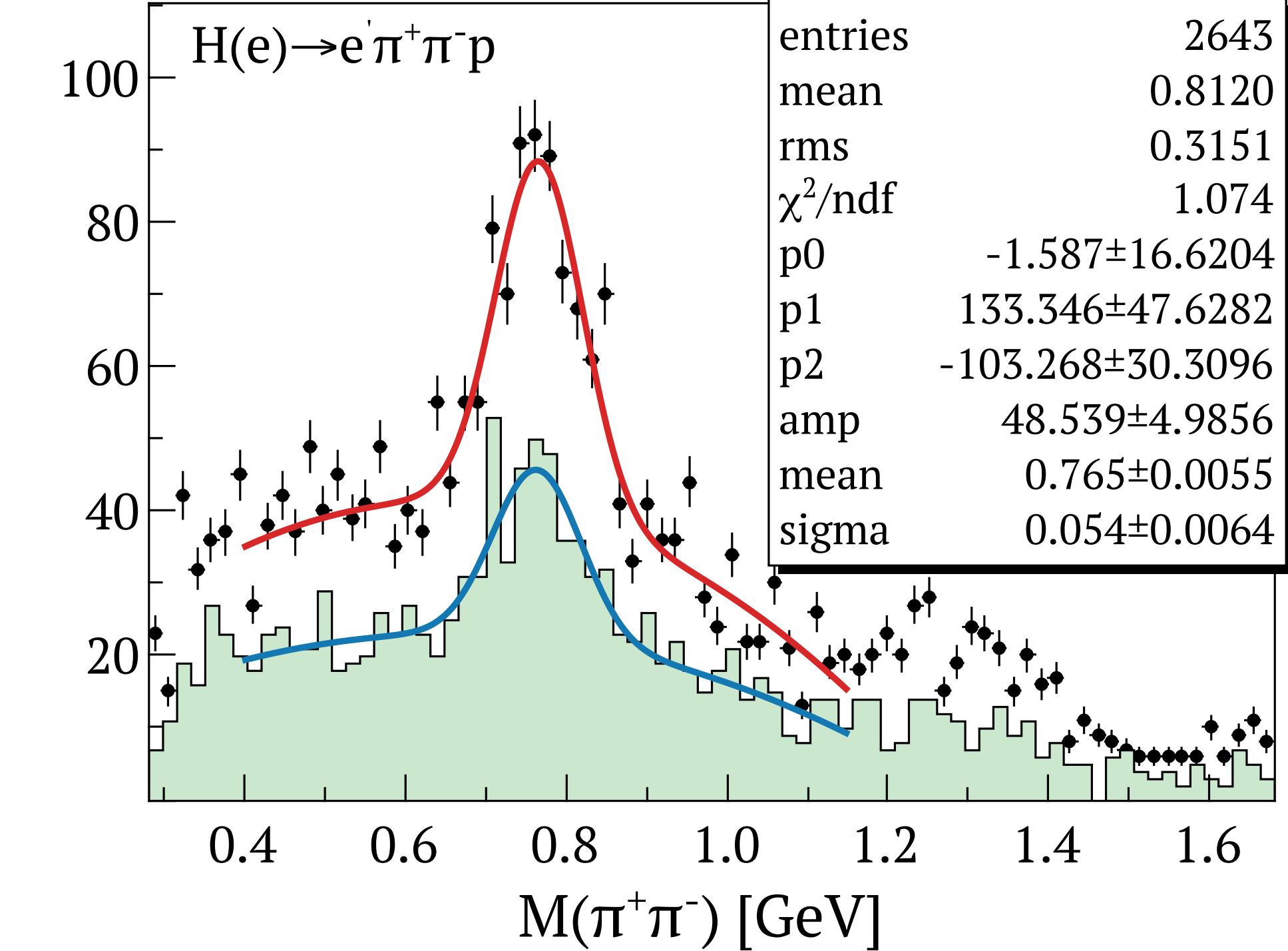
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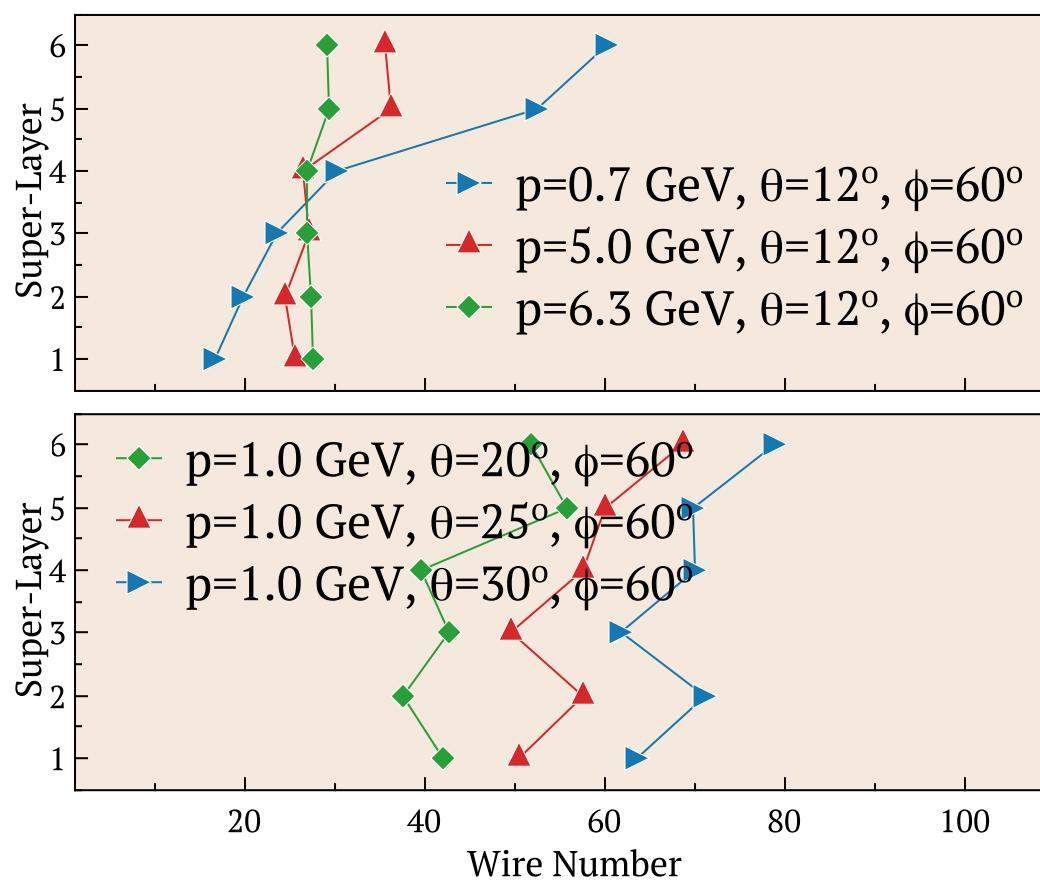
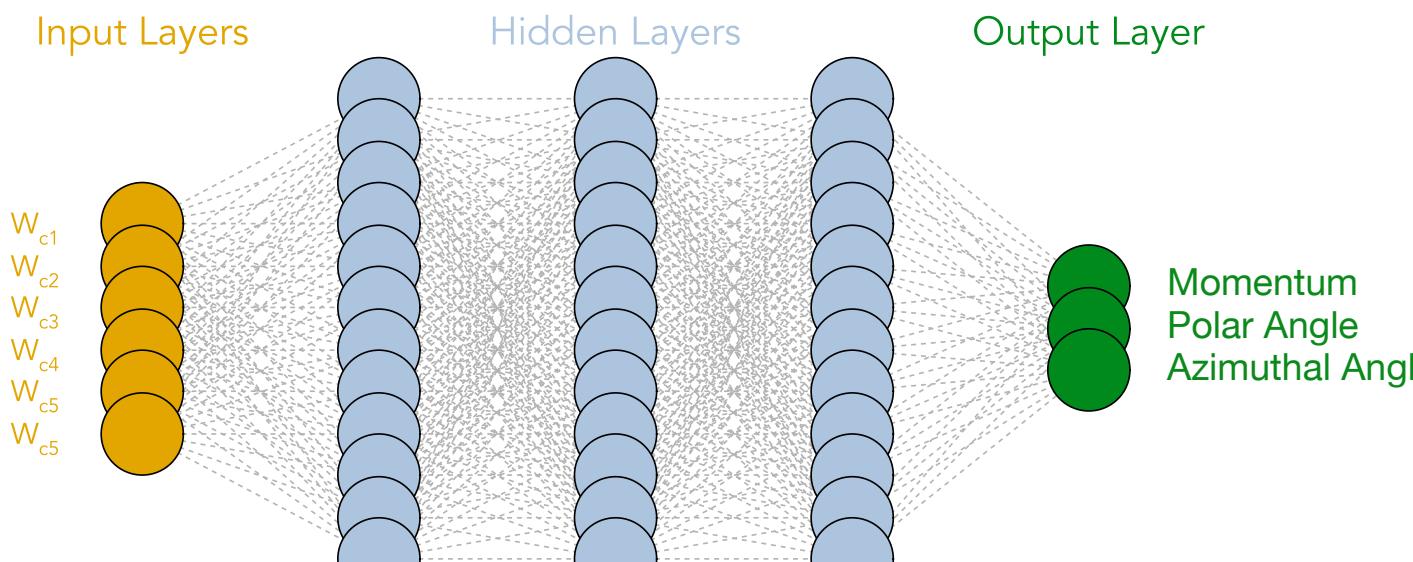


Relative Gain 2.34



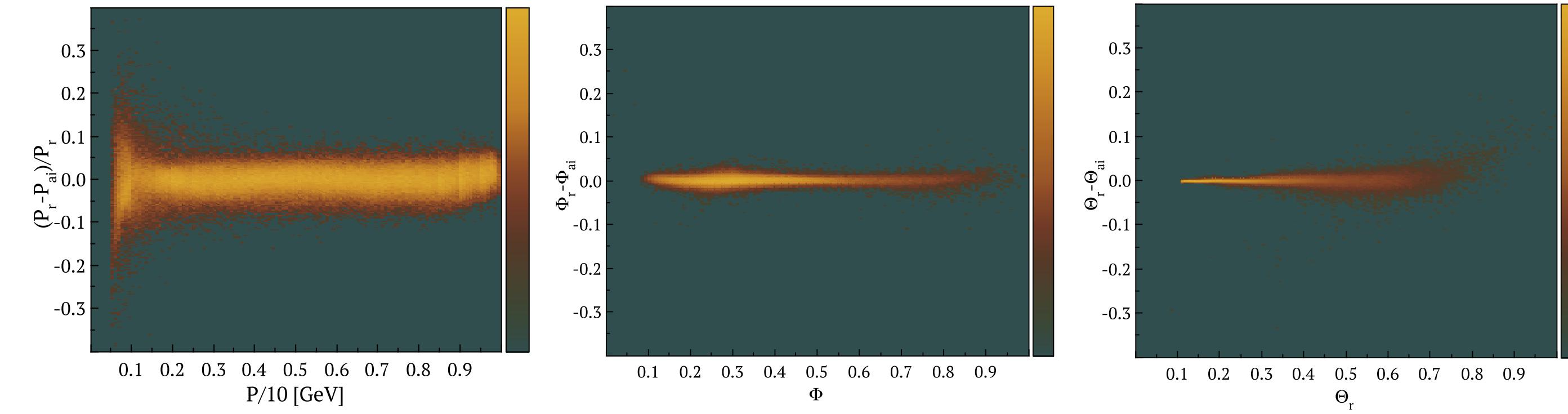
The statistics increase by a factor of 2.34 for exclusive rho mesons when using AI in reconstruction

What do we have now.

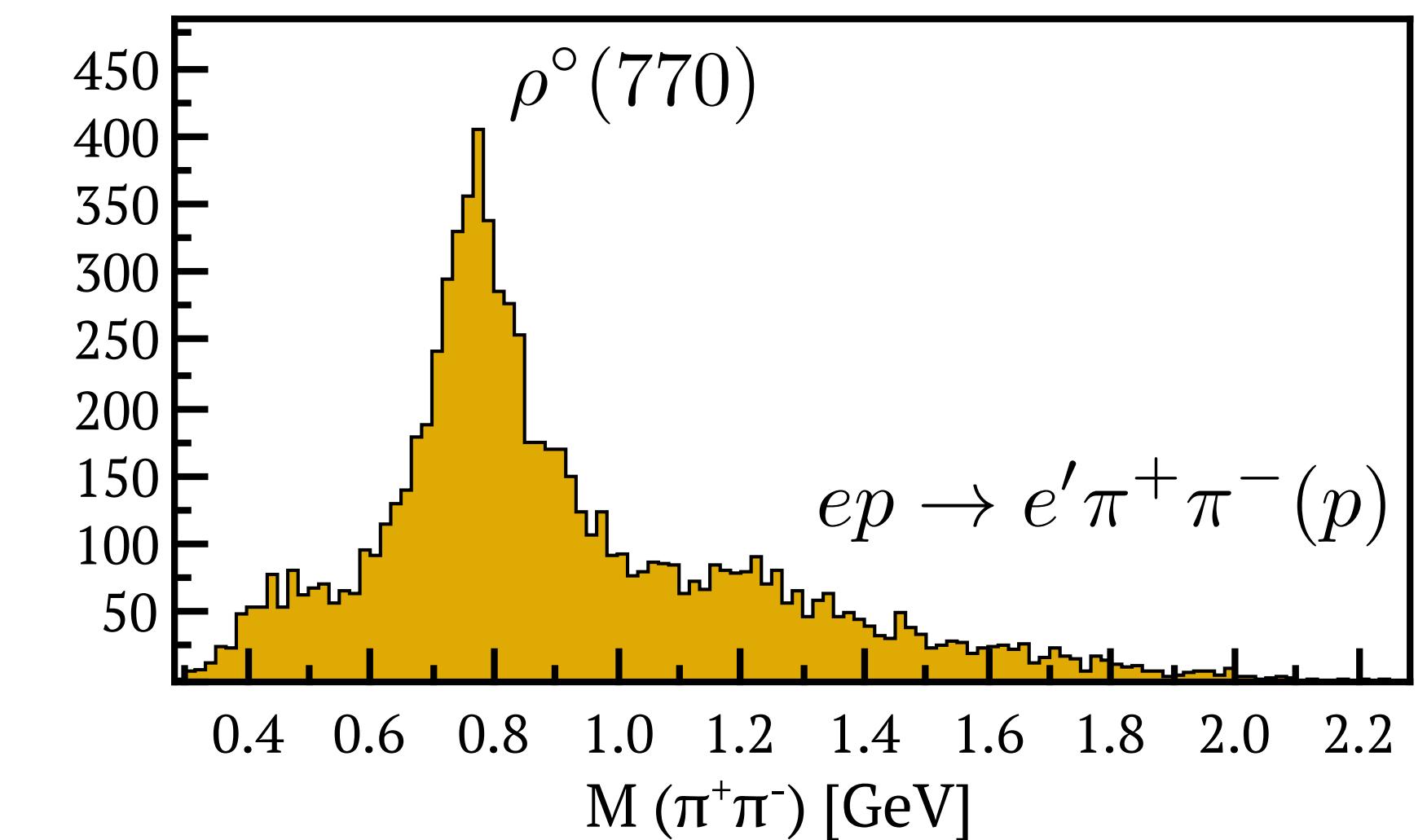
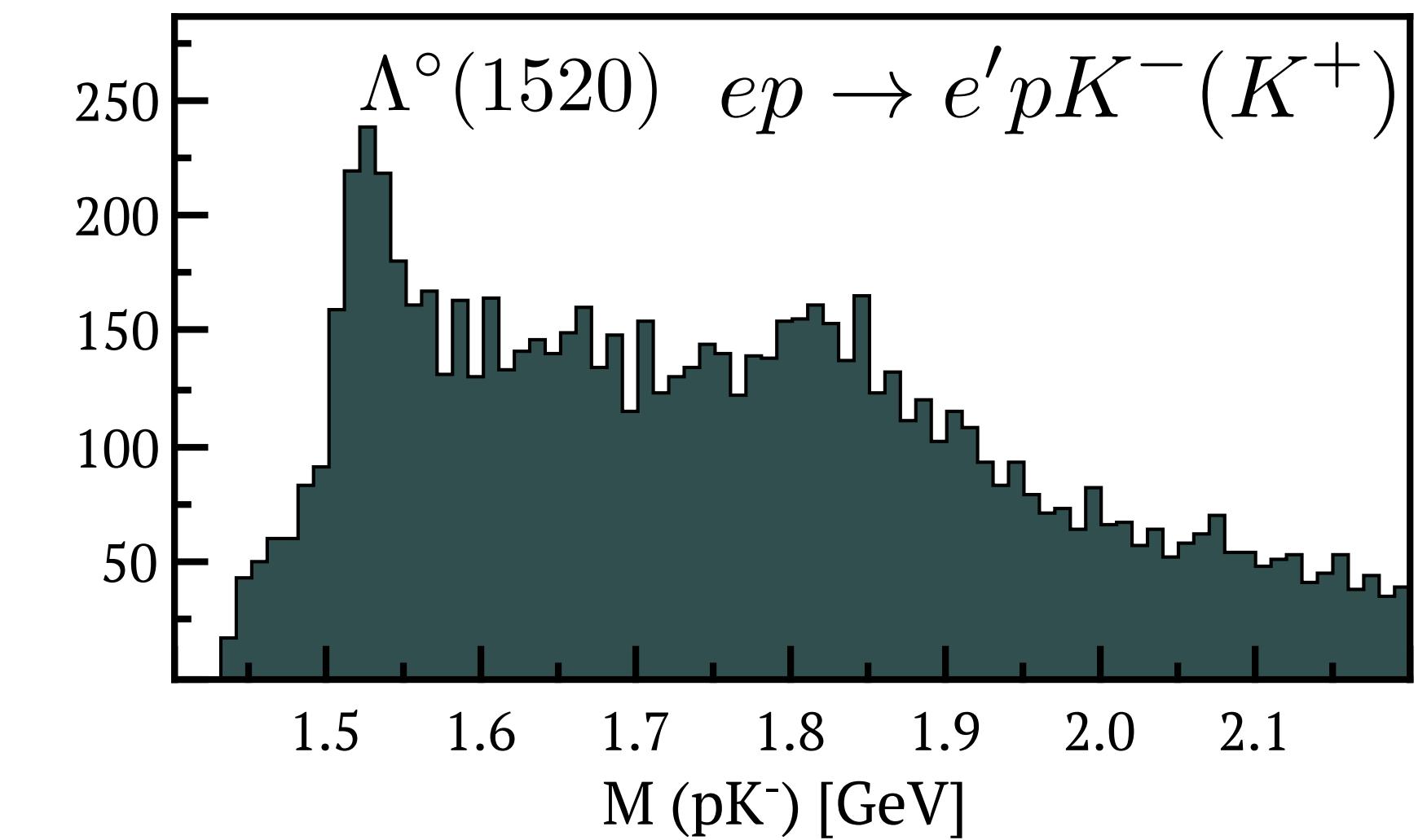


Regression Neural network to predict the track momentum and direction.
The track momentum is reconstructed with an accuracy of 1.4%-1.7%

Physics reactions can be cleanly identified using particle parameters inferred by the neural network.



Distributions calculated from track reconstruction from RAW Drift Chamber hits
Inference speed 96 kHz on a laptop (MacBook M3)



Conclusion

■ AI approach

- Do fast reconstruction of the track with AI, and isolate rho
- The AI reconstruction is faster than DAQ, making it feasible
- Analyse rho meson angular observables (asymmetries, modulations, etc..)
- Determine how the extraction systematic errors change as a function of beam current
- Build an AI model that can predict the optimal beam conditions for the experiment.