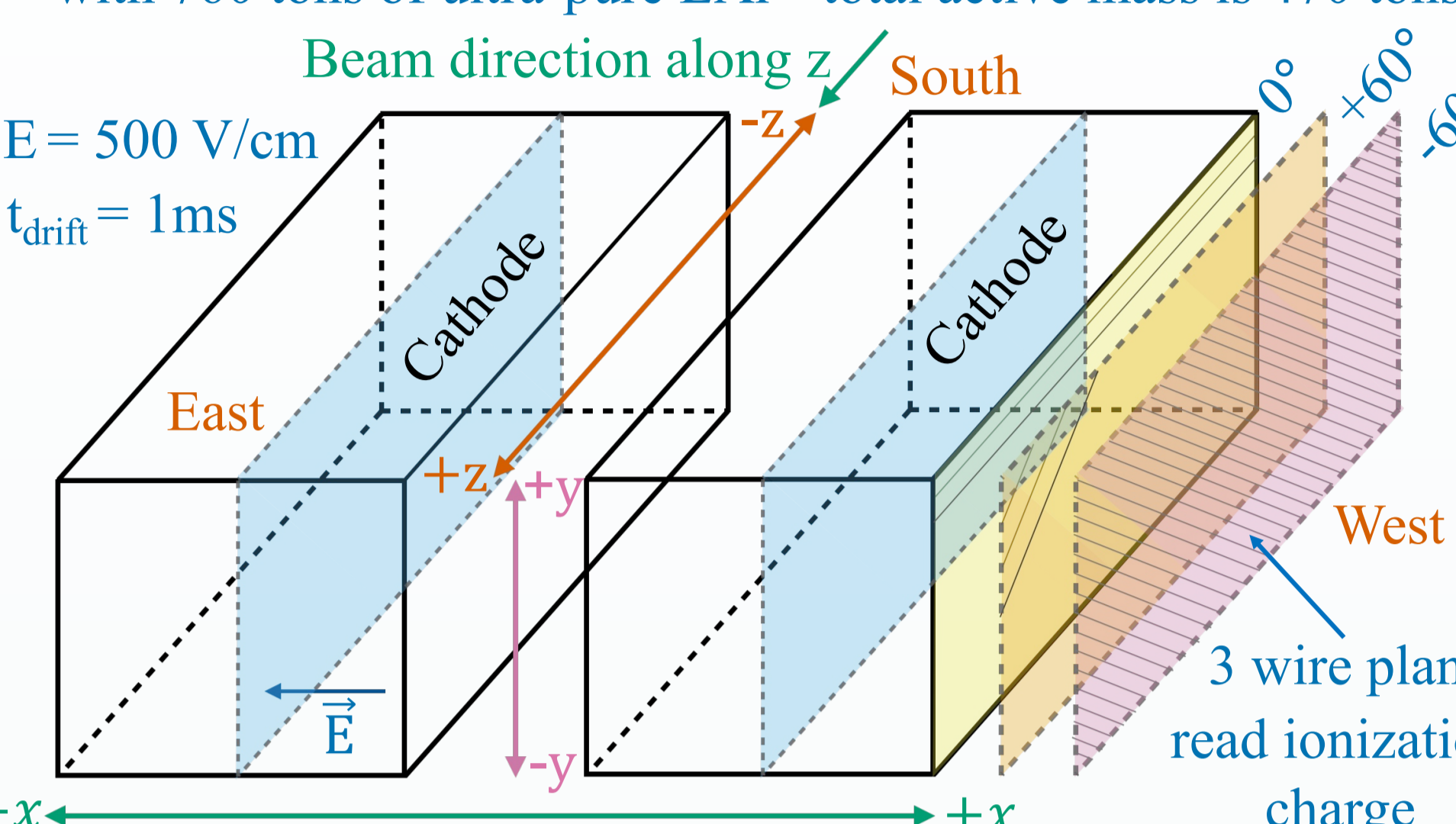


1 The ICARUS experiment in a nutshell

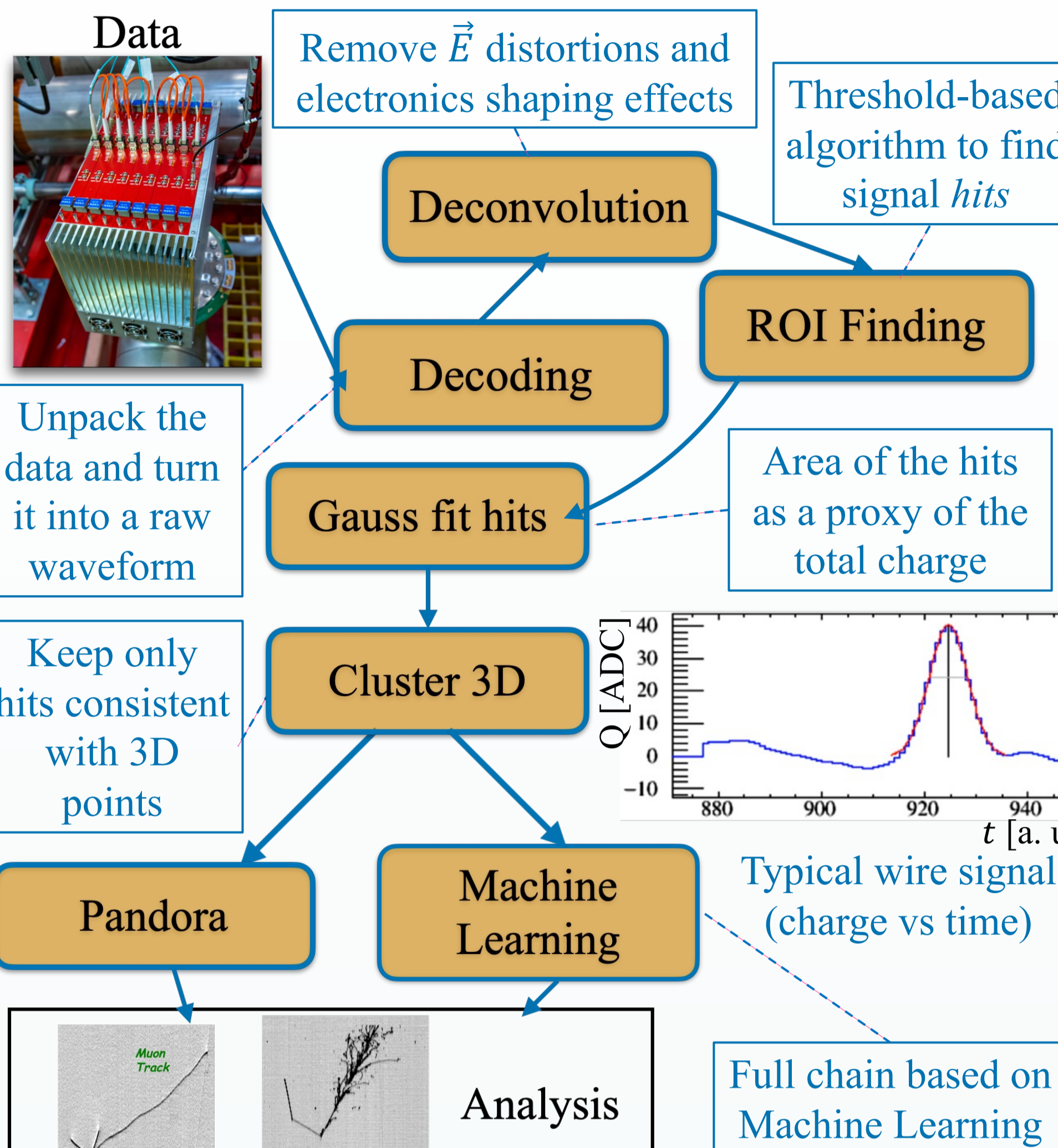
- Liquid Argon Time Projection Chambers (LAR-TPCs)¹: detectors with 3D imaging and calorimetric reconstruction capabilities - ideal for ν interaction studies in a wide E range
- ICARUS-T600² is the first large scale LAR-TPC: two identical cryostats (3.6 x 3.9 x 19.6 m³) housing two TPCs each, filled with 760 tons of ultra-pure LAr - total active mass is 470 tons



360 PMTs for timing/trigger are located behind the wire planes

- Initial operations at LNGS³ (Italy), taking data at FNAL (US) since June 2022 - far detector of the SBN⁴ program
- Rich physics program: search for 1eV mass-scale sterile ν , measurements of ν -Ar cross sections, beyond SM searches

2 The ICARUS TPC event reconstruction chain



Remove \vec{E} distortions and electronics shaping effects

Threshold-based algorithm to find signal hits

Deconvolution

Decoding

ROI Finding

Area of the hits as a proxy of the total charge

Gauss fit hits

Cluster 3D

Typical wire signal (charge vs time)

Full chain based on Machine Learning

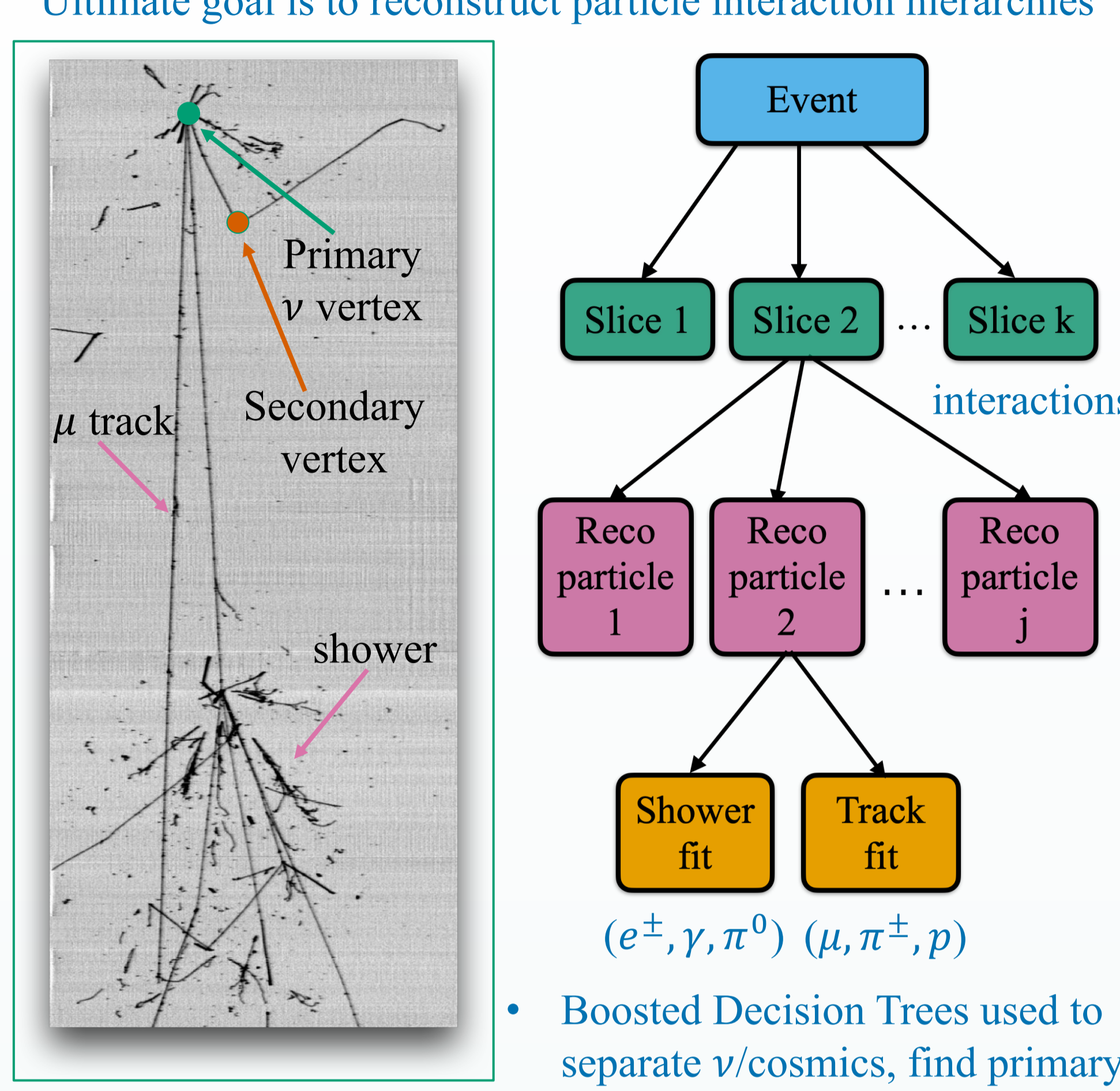
Pandora

Machine Learning

Analysis

3 The Pandora event reconstruction framework

- Multi-algorithm pattern-recognition software⁵ PandoraPFA
- Ultimate goal is to reconstruct particle interaction hierarchies



Event

Slice 1, Slice 2, ..., Slice k

Primary ν vertex

Secondary vertex

μ track

shower

Reco particle 1, Reco particle 2, ..., Reco particle j

interactions

Shower fit, Track fit

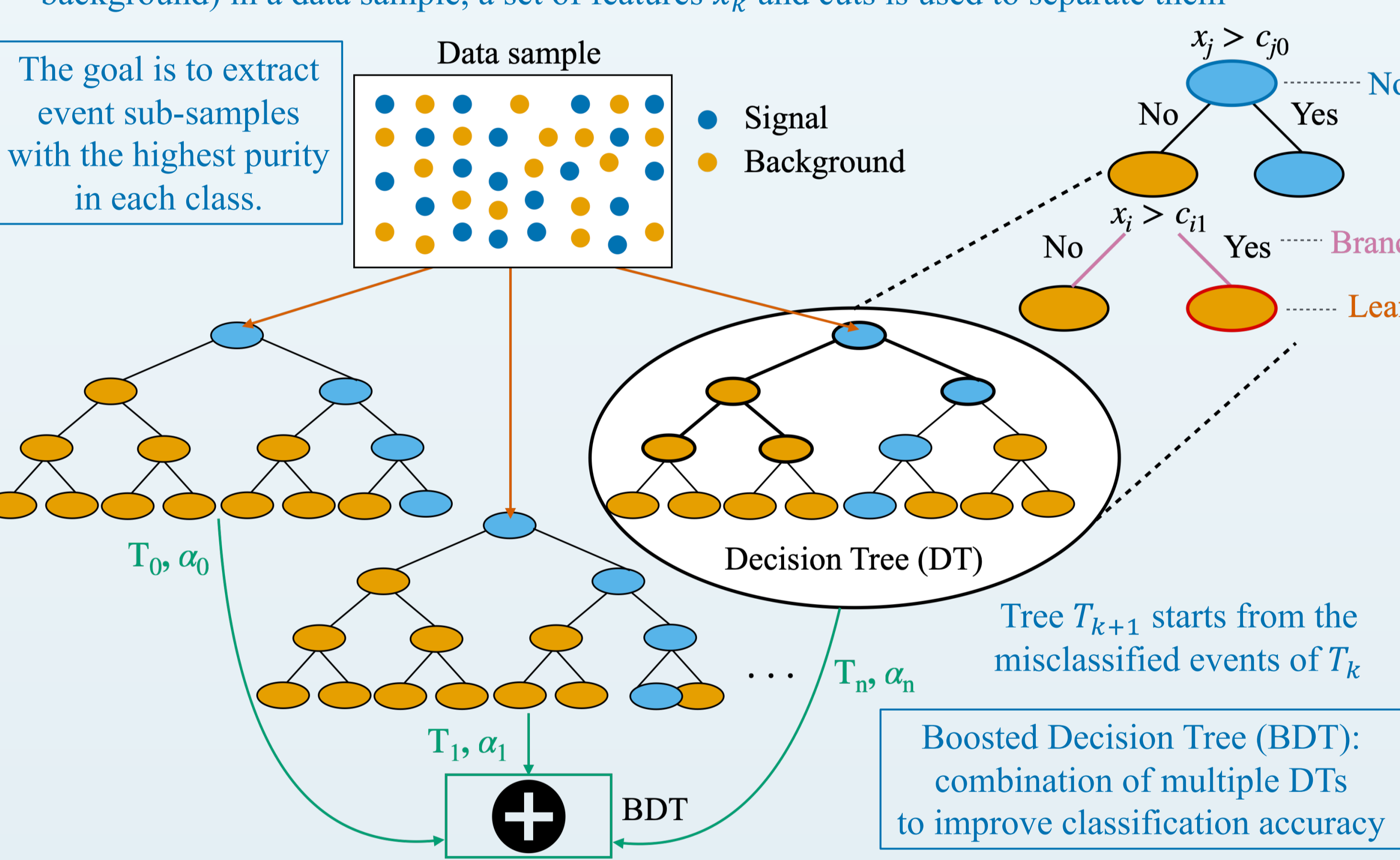
(e^\pm, γ, π^0) (μ, π^\pm, p)

- Boosted Decision Trees used to separate ν /cosmics, find primary vertex, separate tracks/showers
- Common in LAr detectors

4 Boosted Decision Trees (BDTs)

- Machine learning algorithm used for event classification: given n event classes (e.g. signal and background) in a data sample, a set of features x_k and cuts is used to separate them

The goal is to extract event sub-samples with the highest purity in each class.



Data sample

Signal (blue dot), Background (yellow dot)

Node: $x_j > c_{j0}$

Branch: $x_i > c_{i1}$

Leaf

Decision Tree (DT)

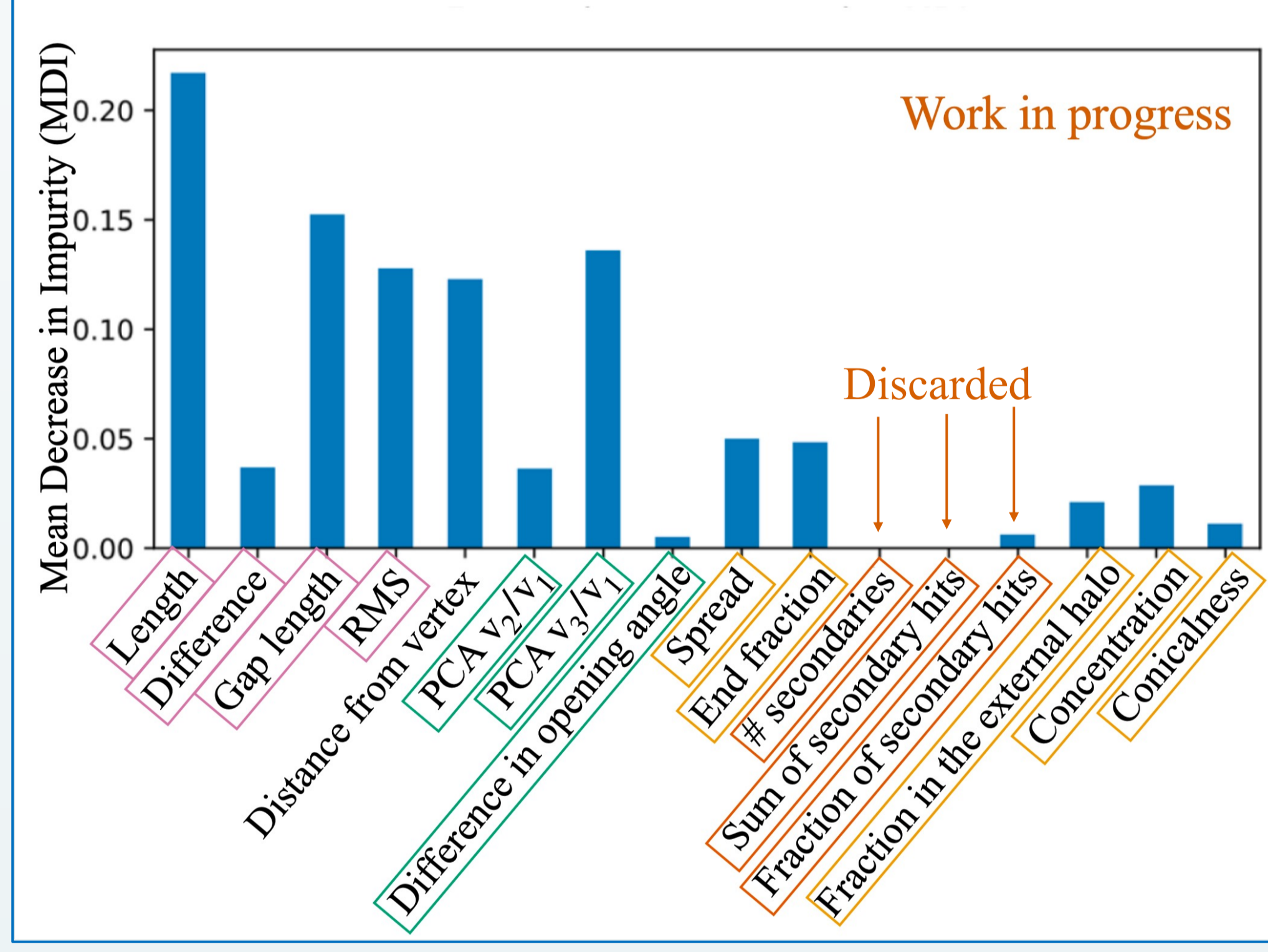
Tree T_{k+1} starts from the misclassified events of T_k

Boosted Decision Tree (BDT): combination of multiple DTs to improve classification accuracy

- Training phase to learn the correct events classification on an input dataset
- Testing phase to classify events from a different sample using the outcome of the previous step

5 The BDT to discriminate tracks and showers in ICARUS

- BDT to separate track-like (μ, π^\pm, p) and shower-like (e^\pm, γ, π^0) particles relies on geometrical and calorimetric features (charge as energy proxy) of the reconstructed particle hits
- The output is a track-score: parameter in [0,0.5] for showers and [0.5,1] for track-like particles
- Previous training on SBND Monte Carlo (MC) sample with outdated simulation/reconstruction



Work in progress

Discarded

Variable type	#	Selected
Linear fit	4	✓
Distance from interaction vertex	1	✓
Principal Component Analysis (PCA)	3	✓
Charge (old)	2	✓
Hierarchy (new)	3	✗
Conical charge (new)	3	✓

- New BDT training with adaptive boosting based on Python library `Scikit learn`
- 6 new features tested,
- Final set includes 13 BDT variables with the highest relative importance

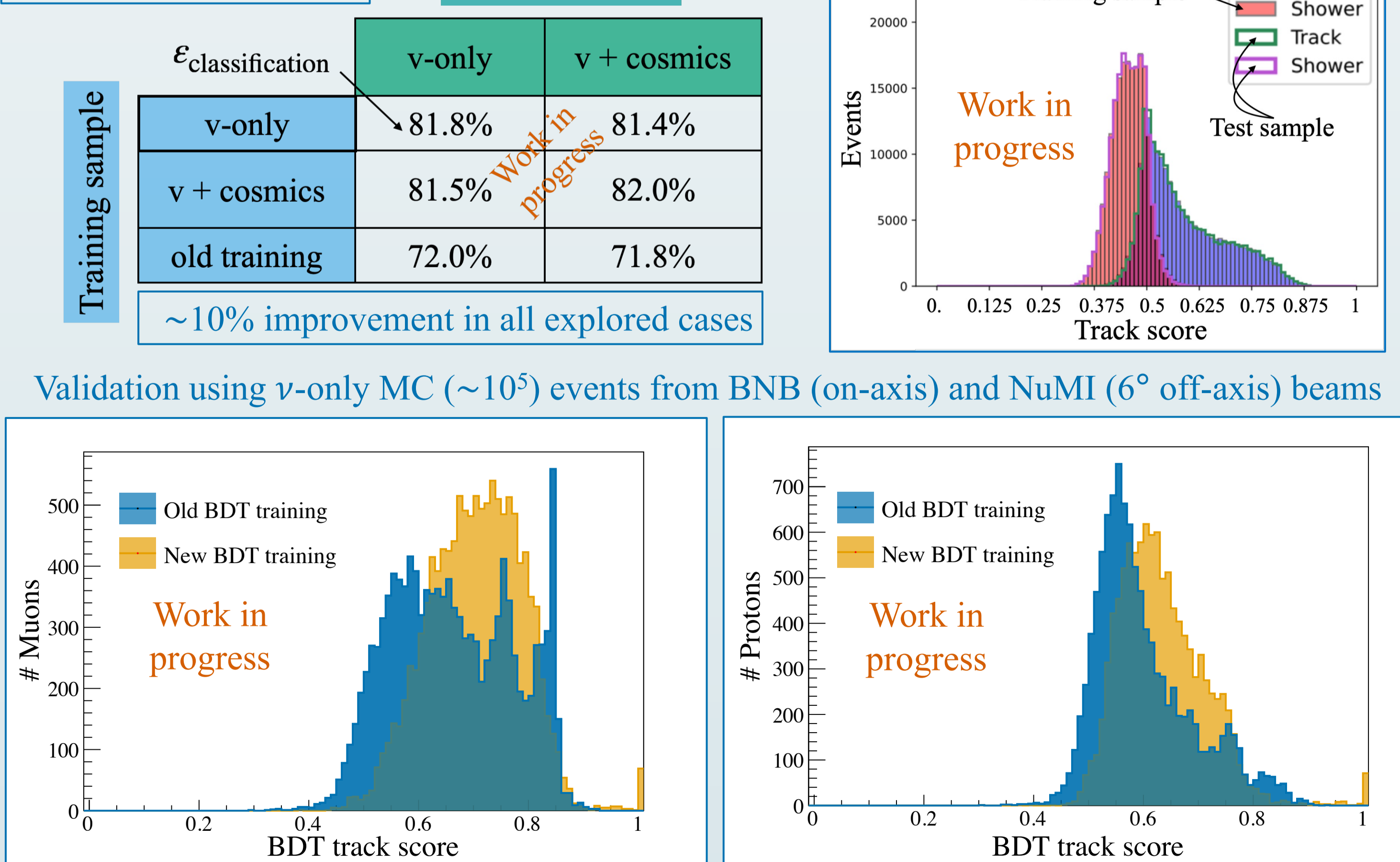
- Different training/testing sample compositions and combinations tested with $O(10^4-10^5)$ MC events from the BNB ν beam – cross validation to mitigate overtraining and optimize parameters

6 New BDT training: results and validation tests

BNB beam MC samples

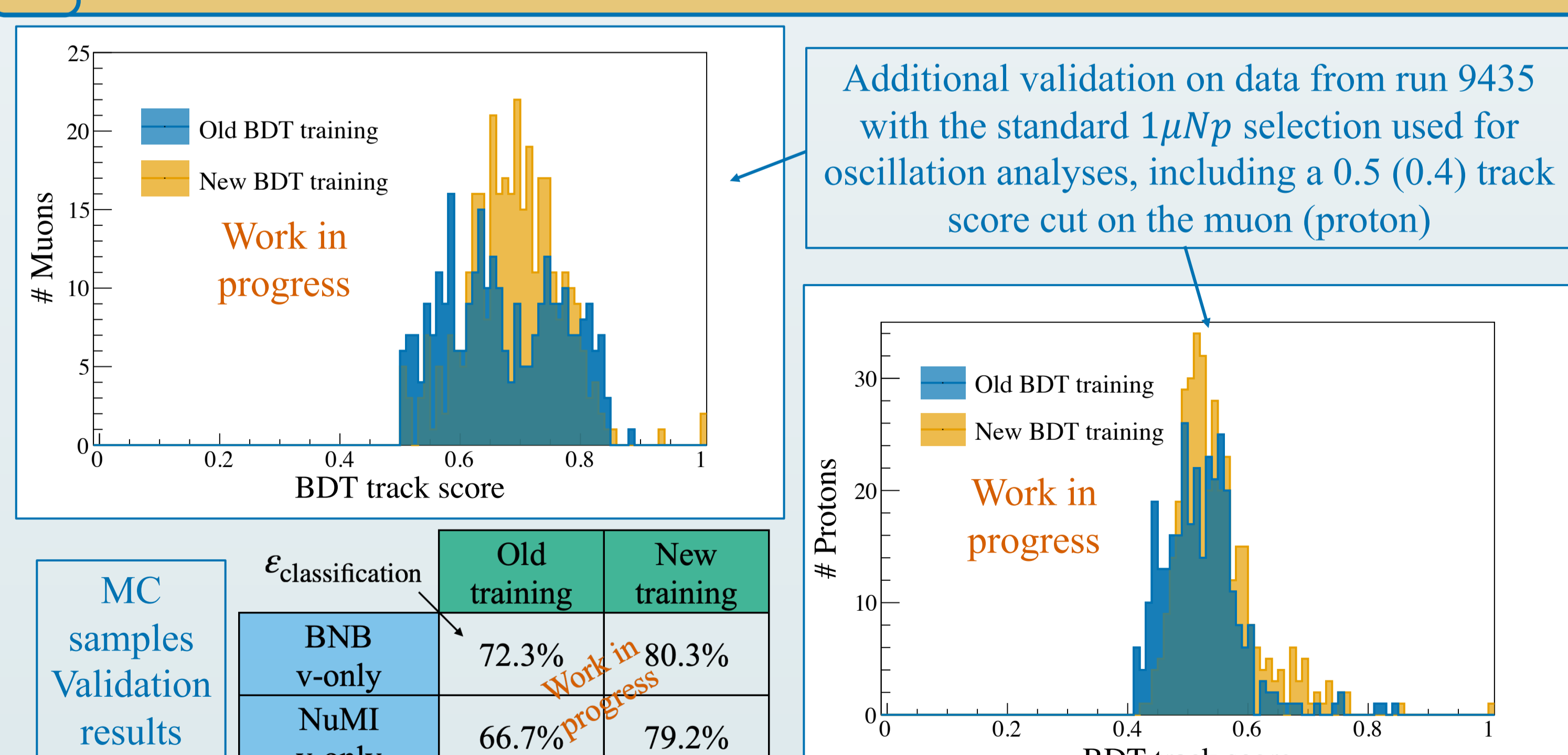
Training sample	Test sample	
	v-only	v + cosmics
v-only	81.8%	81.4%
v + cosmics	81.5%	82.0%
old training	72.0%	71.8%

~10% improvement in all explored cases



Validation using ν -only MC ($\sim 10^5$) events from BNB (on-axis) and NuMI (6° off-axis) beams

7 New training: results and validation tests



Additional validation on data from run 9435 with the standard $1\mu Np$ selection used for oscillation analyses, including a 0.5 (0.4) track score cut on the muon (proton)

MC samples Validation results	$\epsilon_{\text{classification}}$	
	Old training	New training
BNB v-only	72.3%	80.3%
NuMI v-only	66.7%	79.2%

8 Conclusions and perspectives

- Repeat the BDT training on new MC samples to include recent upgrades of the signal processing
- Test different event samples for the training, including a generator-model-agnostic input sample also used within the Machine-Learning-based processing chain

References

[1] C. Rubbia, *CERN-EP, 77-08*, (1977)

[2] P. Abratenko et al. (ICARUS Collaboration), *Eur. Phys. J. C* **83**, 467 (2023)

[3] M. Antonello et al. (ICARUS Collaboration), *Eur. Phys. J. C* **73**, 2599 (2013)

References

[4] P. A. N. Machado, O. Palamara, and D.W. Schmitz, *Annual Review of Nuclear and Particle Science*, **69**, 367–387 (2019)

[5] R. Acciarri et al. (MicroBooNE Collaboration), *Eur. Phys. J. C* **78** (2018)