

PANDORA: LARTPC EVENT RECONSTRUCTION

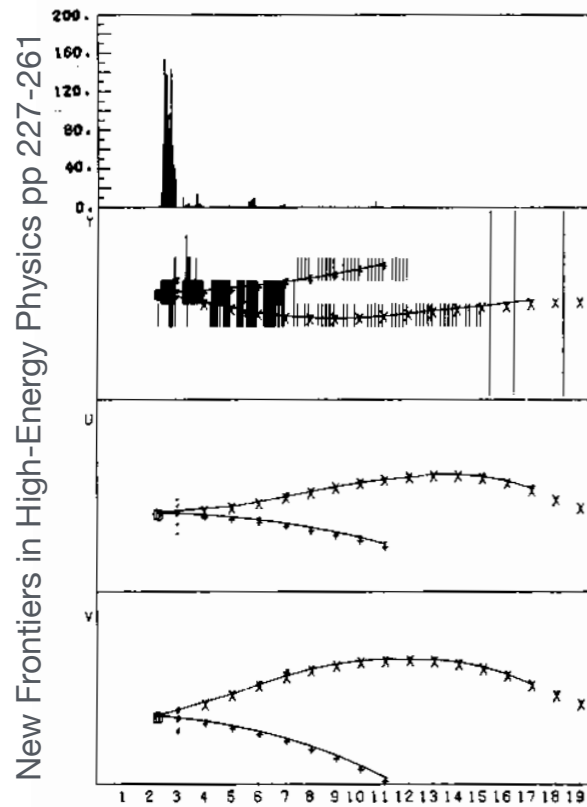
John Marshall for the Pandora Team

16th September 2020

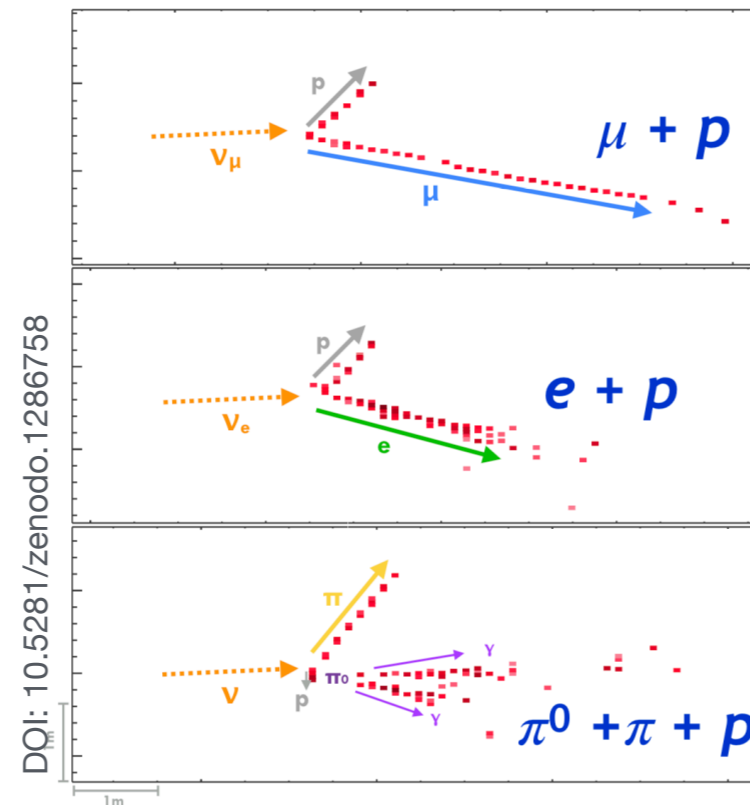
- **LArTPC detectors and event reconstruction**
- **Details of key Pandora LArTPC pattern-recognition algorithms**
- Handling LArTPCs w/ multiple volumes and cosmic-ray backgrounds
- Pattern-recognition performance

Key references: **Eur. Phys. J. C (2018) 78: 82**
and **Eur. Phys. J. C (2015) 75: 439**

Neutrino Detectors

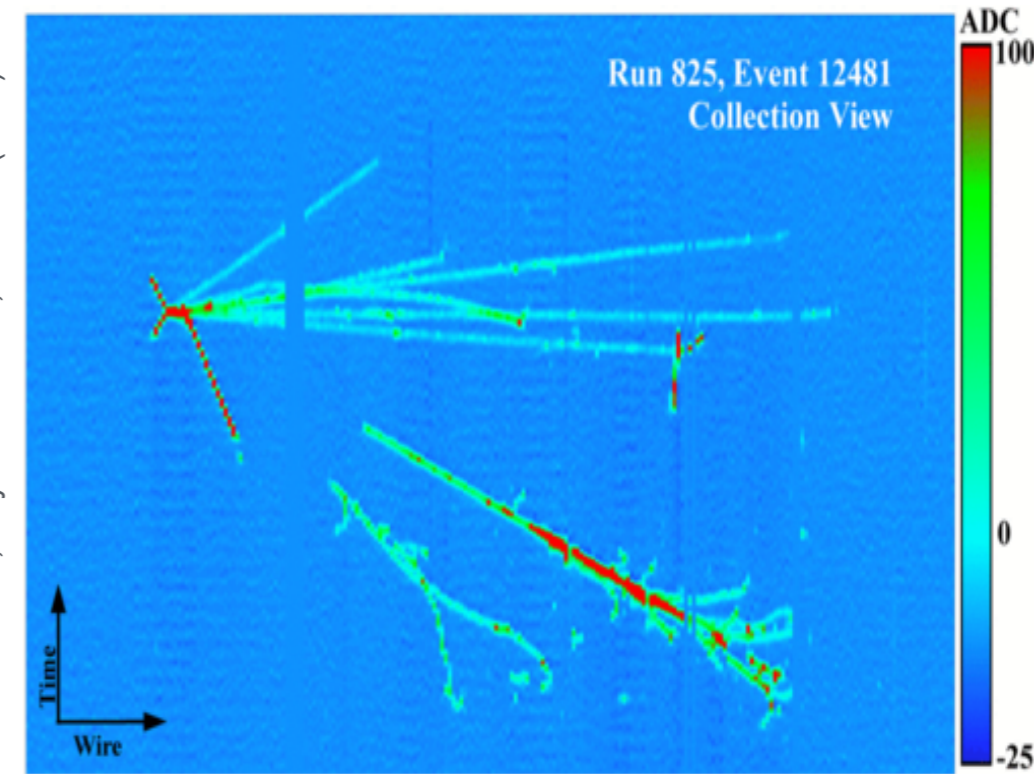


CDHS



NoVA

R. Acciari et al, Phys. Rev. D 95, 072005 (2017)



ArgoNeuT

- Evolving detector technologies, with general trend towards **imaging** neutrino interactions
 - Emphasis on identifying and characterising individual visible particles
- Physics sensitivity now depends critically on both **hardware** and **software**
 - Need a sophisticated event reconstruction to harness information in the images
- Aim to reconstruct hierarchy of particles of identified types, with measured four-momenta
 - “Particle flow” reconstruction

LArTPC Detectors

Liquid-Argon Time-Projection Chamber (LArTPC)

- Charged particles, e.g. produced in neutrino interactions, deposit ionisation trails in liquid argon.
- Ionisation electrons drift in an applied electric field.
- In a **single-phase** LArTPC, the electrons are detected by a series of wire planes.
- Single-phase LArTPC detectors:
 - Past: ICARUS, ArgoNeuT
 - Current: MicroBooNE, ProtoDUNE, LArIAT
 - Coming soon: SBND, ICARUS@SBN

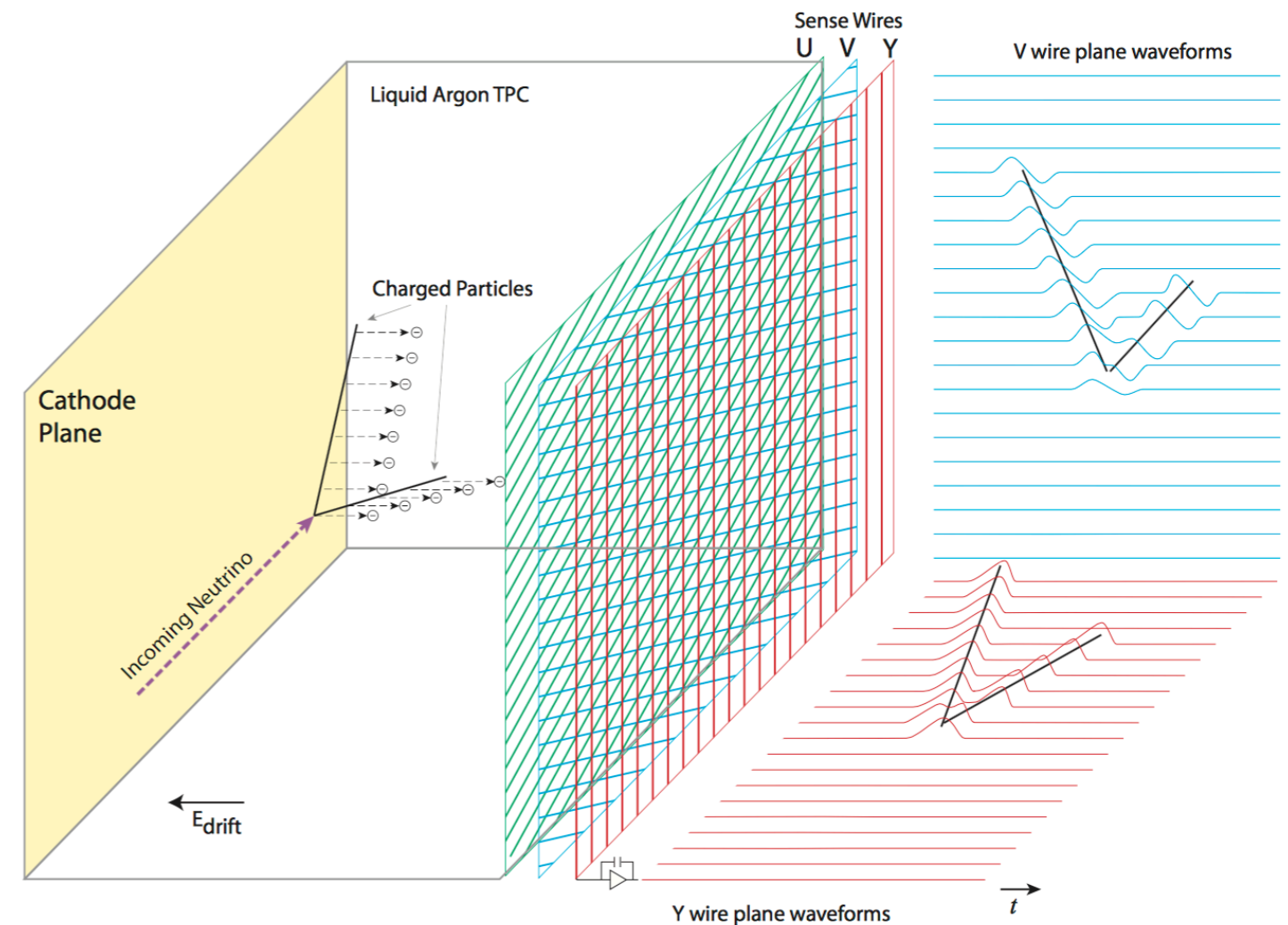
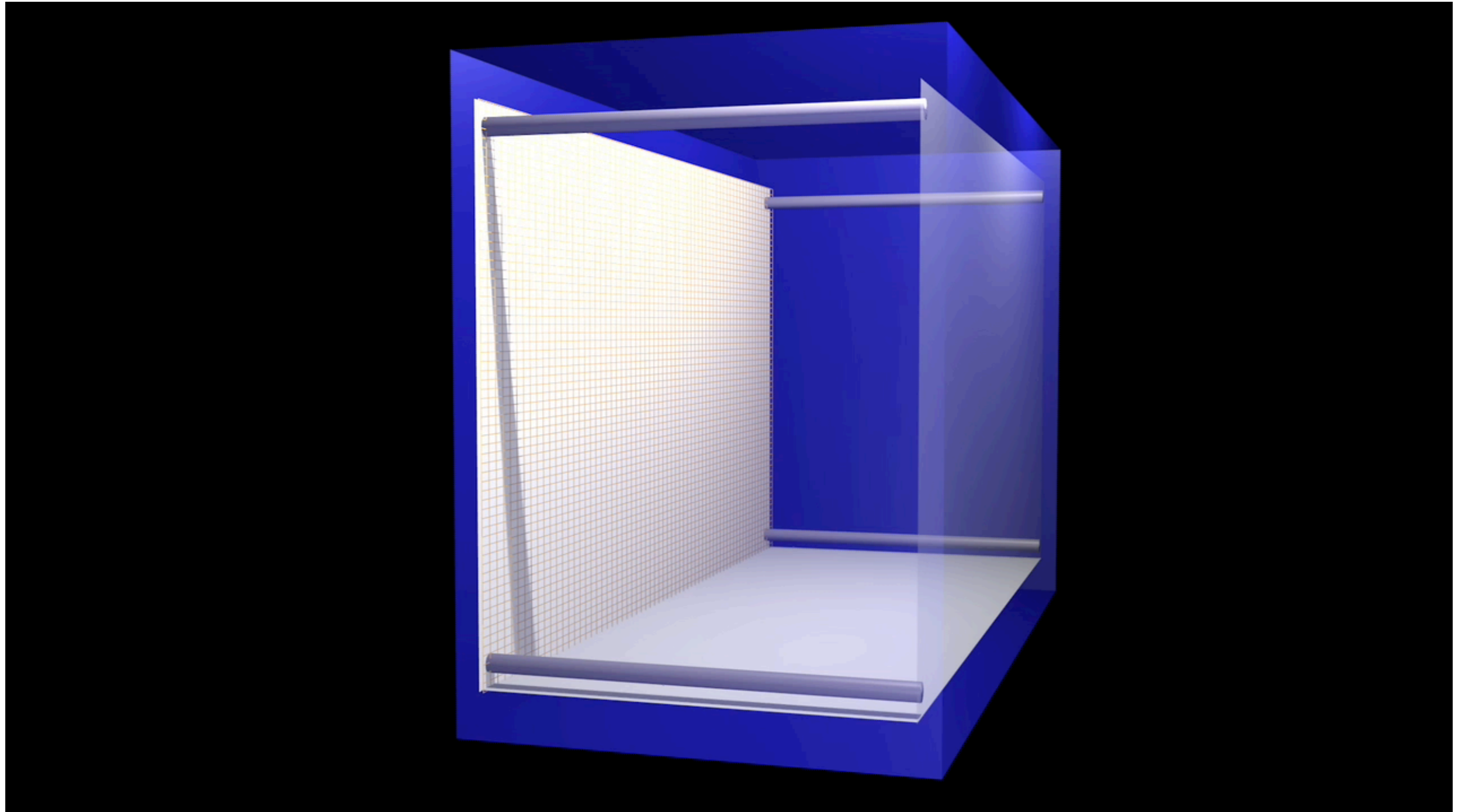


Figure: arXiv:1612.05824

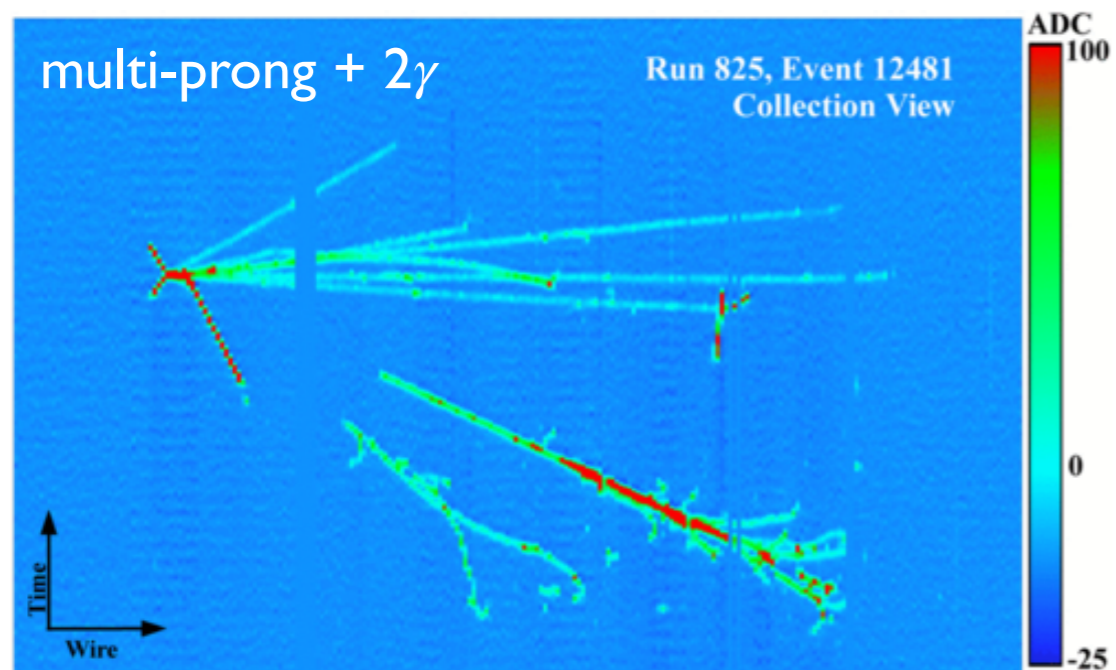
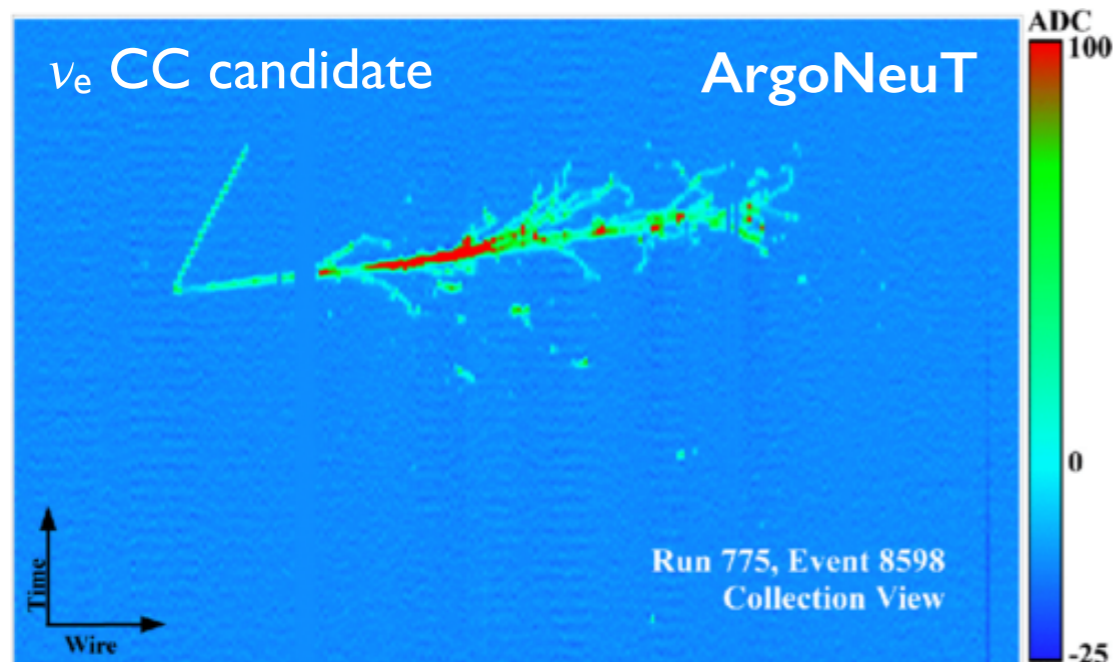
LArTPC Detectors

LArTPCs: a somewhat slicker introduction (assuming embedded video works...)



Video: Fermilab

Why Liquid Argon?



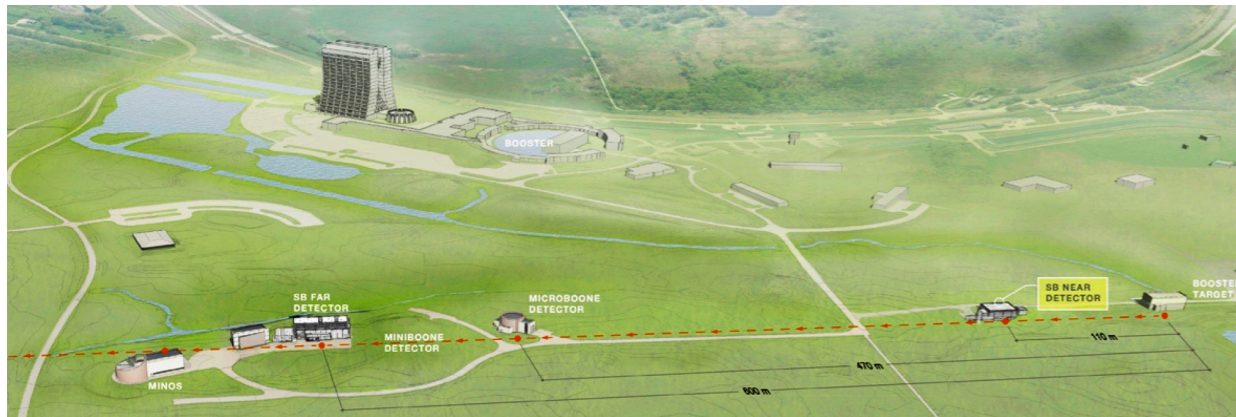
- LArTPC detectors are **fully active** and **fine grain**, offering superb spatial and calorimetric resolution:
 - Reconstruction of multi-prong final states.
 - Particle identification:
 - ▶ $\mu/p/K$ in particle tracks
 - ▶ e/γ in electromagnetic showers
- Potential for high efficiency and low backgrounds in most channels
- Scalable to multi-kiloton masses.

R. Acciari *et al*, Phys. Rev. D 95, 072005 (2017)

LArTPC Programme

One of the key technologies in the current and future neutrino physics programmes is the Liquid-Argon Time-Projection Chamber (LArTPC)

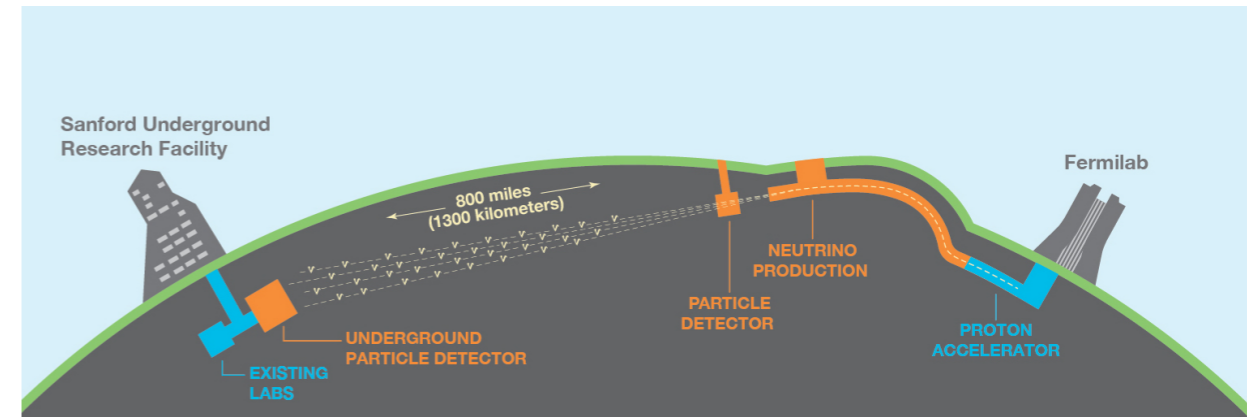
Short-baseline programme



ICARUS MicroBooNE SBND

- Three LArTPC detectors located along the Booster Neutrino Beam (BNB) at Fermilab
- Main goal is to investigate the potential sterile neutrino signals from LSND and MiniBooNE
- Precision cross-section measurements for neutrino interactions on argon

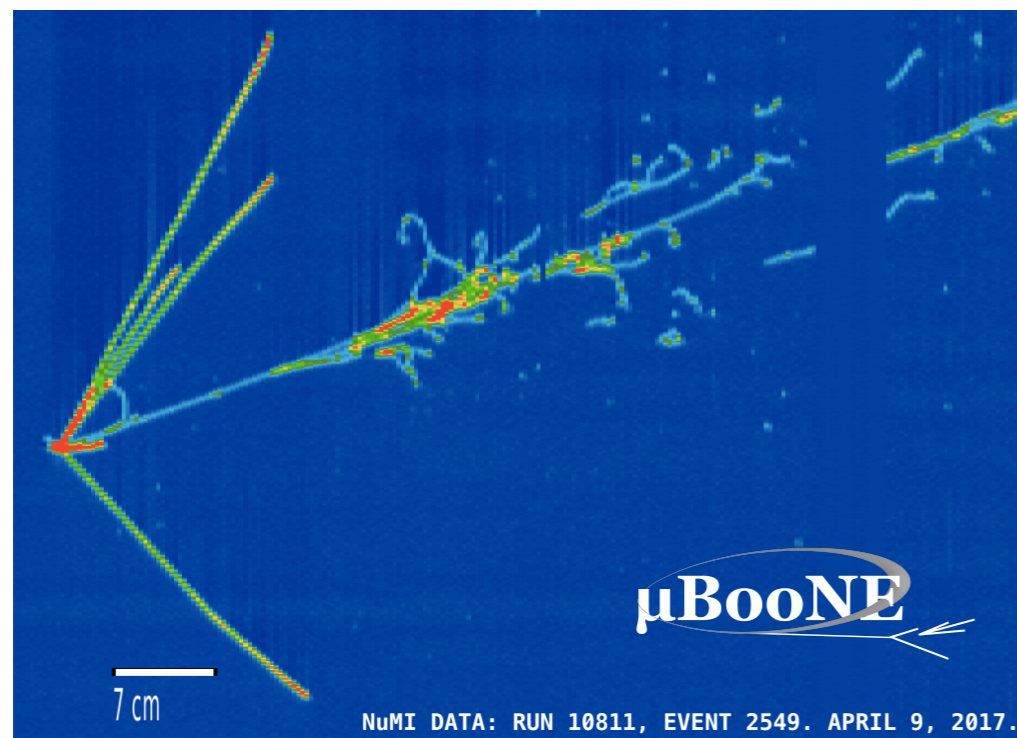
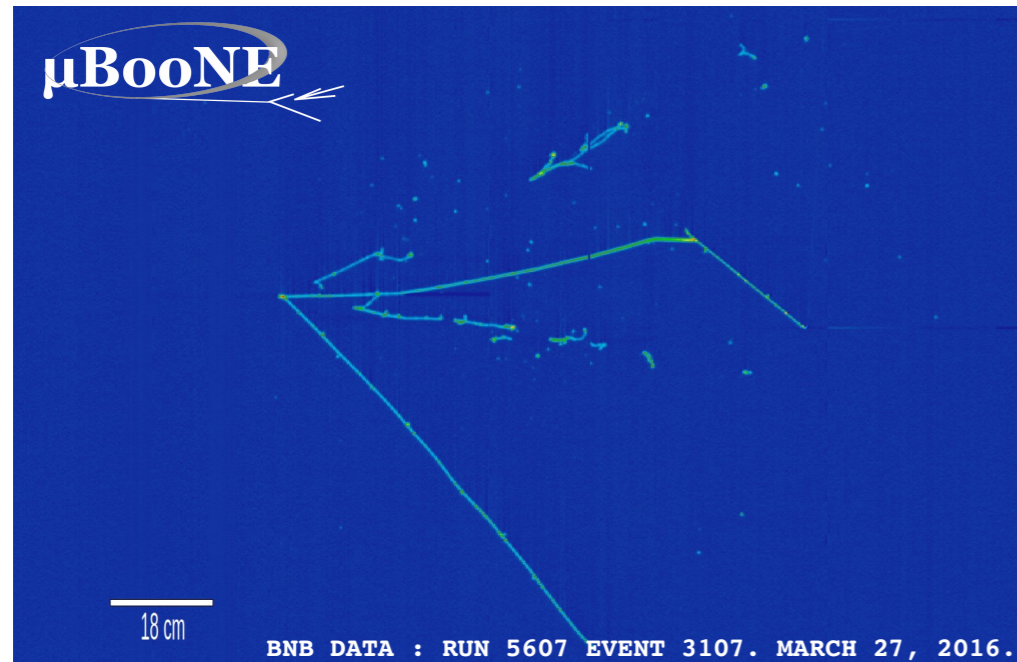
Long-baseline programme



ProtoDUNE_s DUNE

- Neutrino oscillation physics:
 - Discover CP violation in the leptonic sector
 - Resolve mass hierarchy
 - Test three-flavour paradigm
 - Precision parameter measurement
- Proton decay
- Supernova neutrinos

LArTPC Event Reconstruction

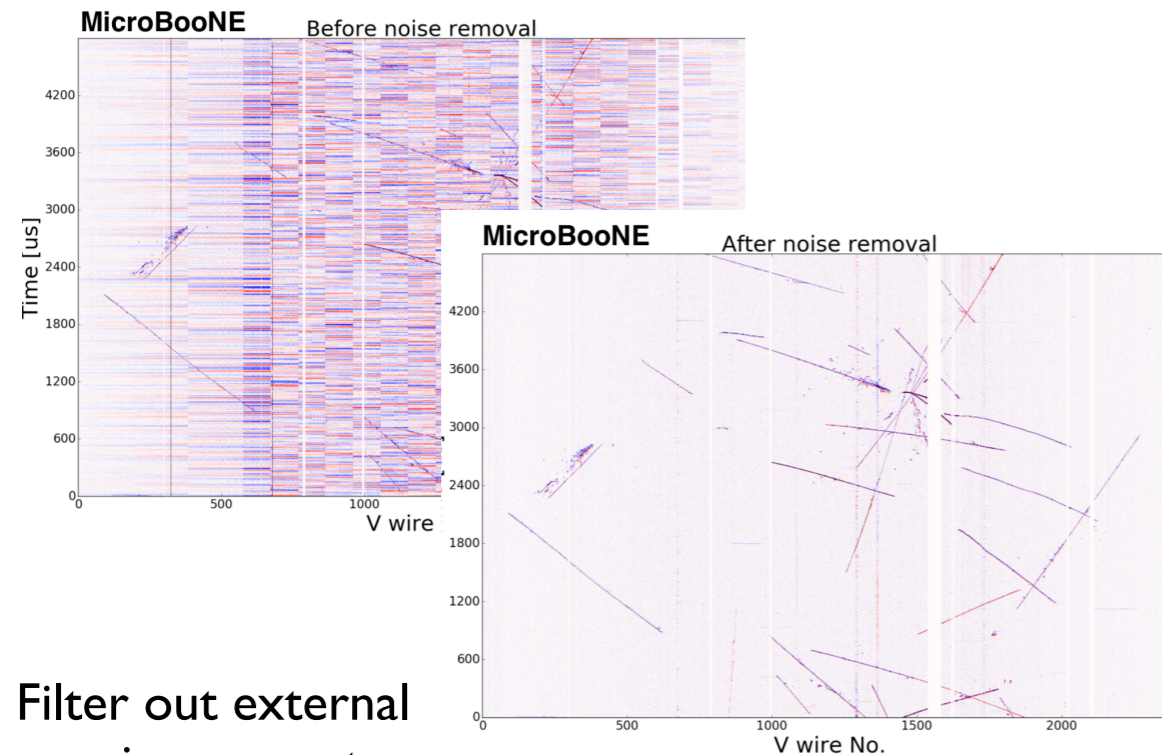


The conversion of raw LArTPC images into analysis-level physics quantities:

- **Low-level steps:**
 - Noise filtering
 - Signal processing
- **Pattern recognition:**
 - The bit you do by eye!
 - Turn images into sparse 2D hits
 - Assign 2D hits to clusters
 - Match features between planes
 - Output a hierarchy of 3D particles
- **High-level characterisation:**
 - Particle identification
 - Neutrino flavour and interaction type
 - Neutrino energy, etc...

Low-Level Steps

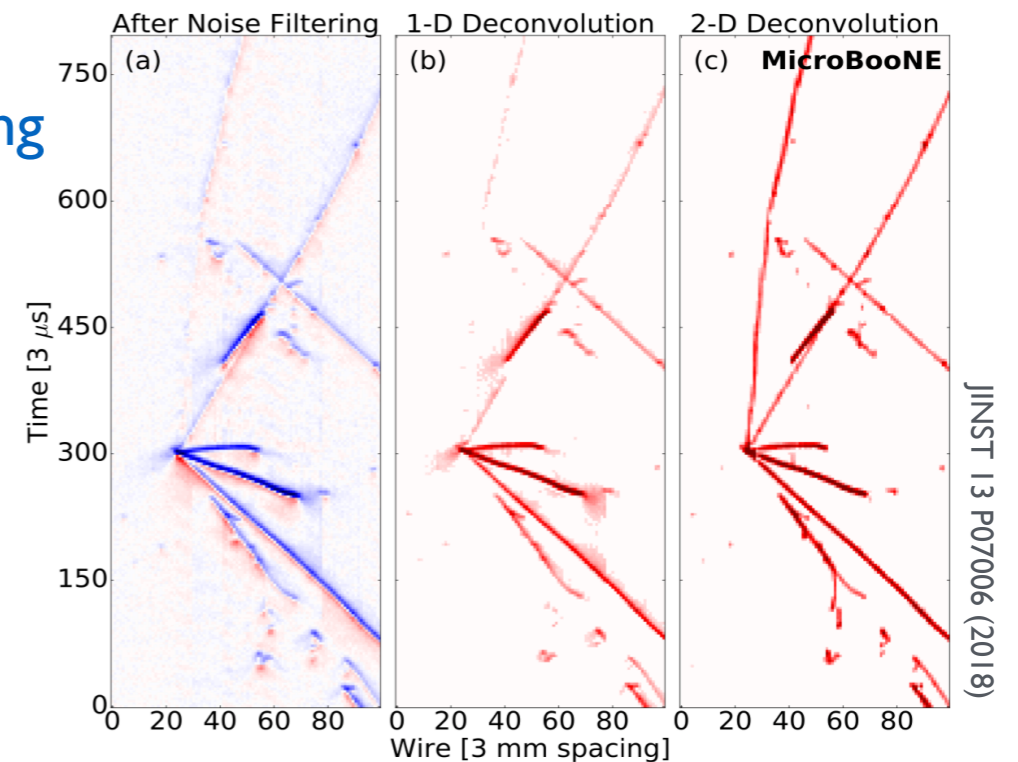
Noise Filtering



Filter out external noise, correct some distortions

JINST 12 P08003 (2017)

Signal Processing

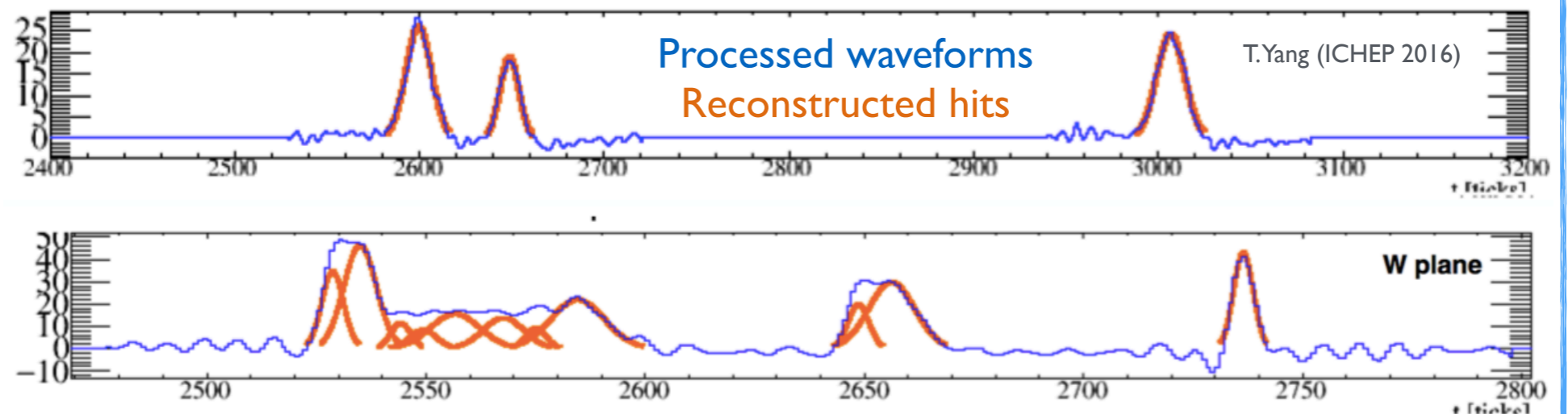


Convert digitised waveform to number of ionization electrons passing through a wire plane at a given time

JINST 13 P07006 (2018)

Hit Finding

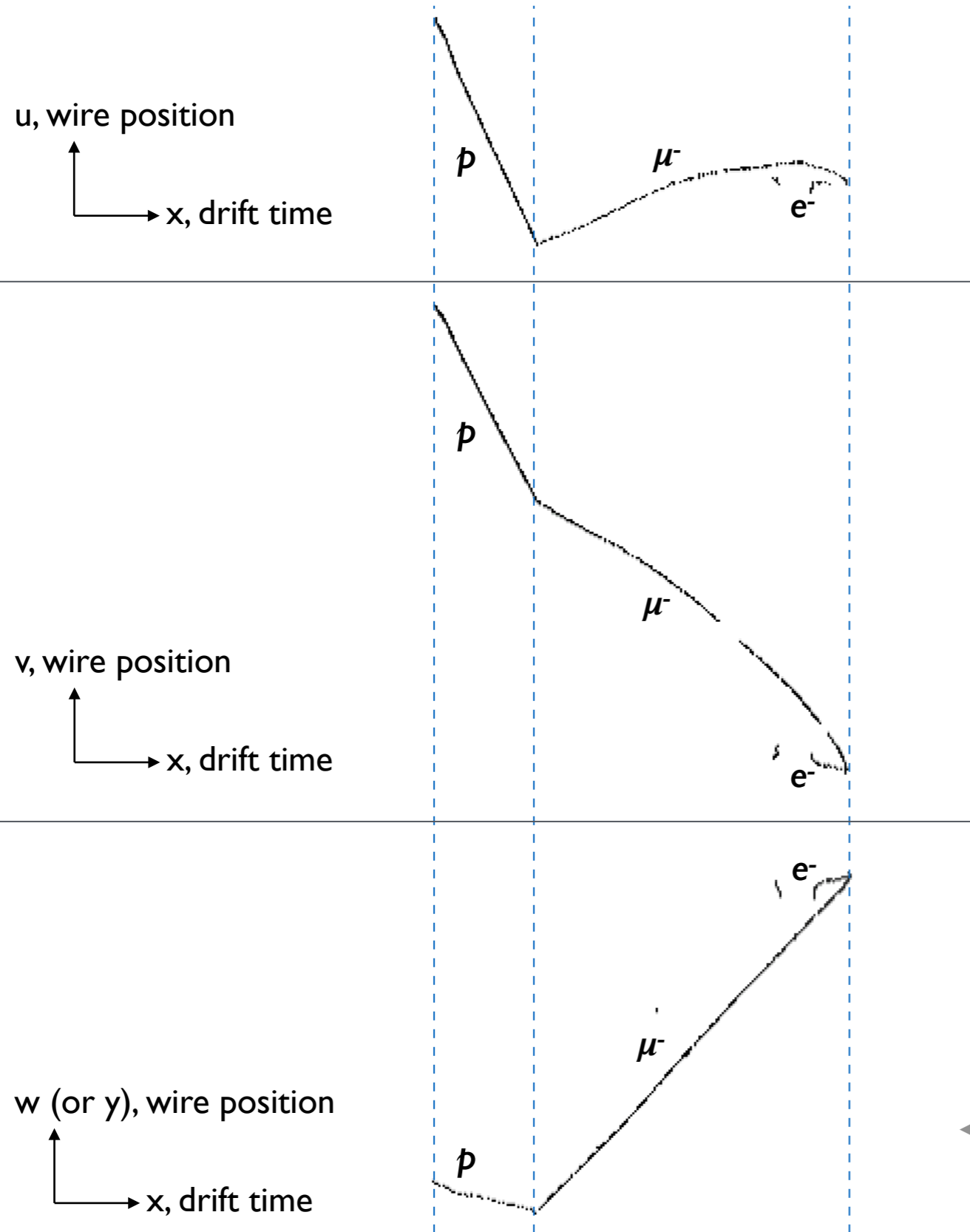
Fit clean waveform with N Gaussians, where N is number of peaks in pulse. Each Gaussian represents a hit.



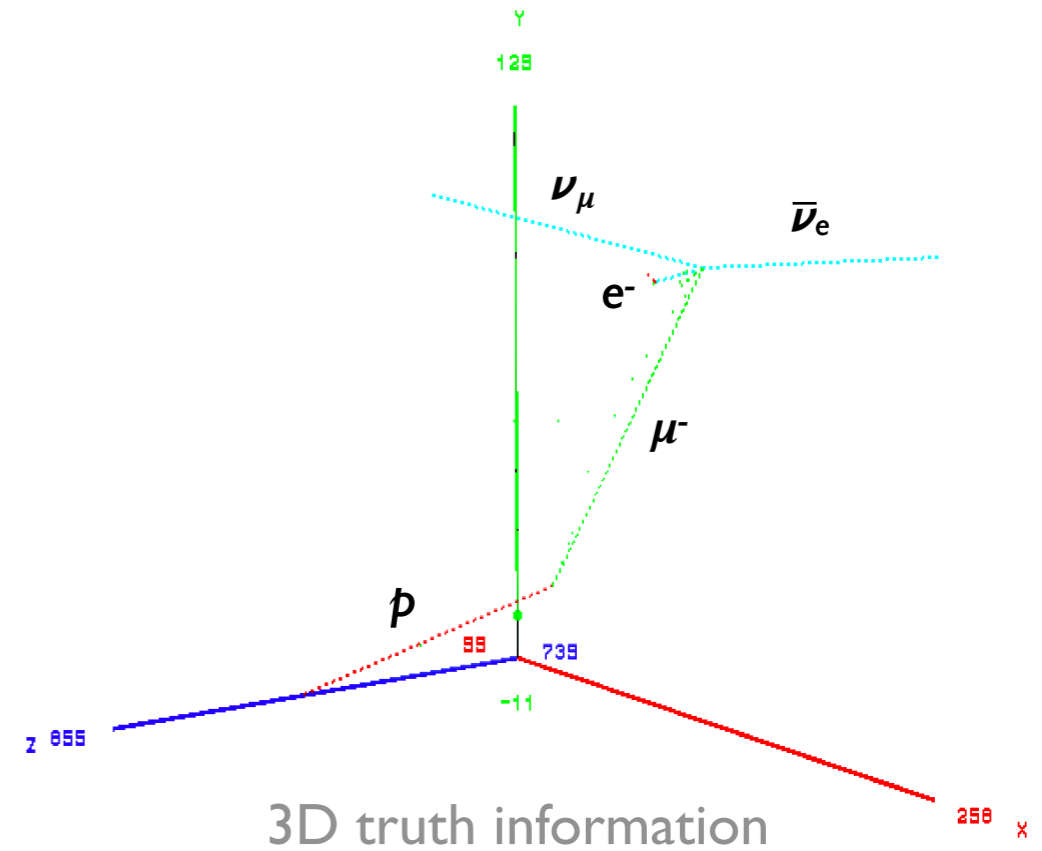
T.Yang (ICHEP 2016)

W plane

Inputs to Pattern Recognition



E.g. CC QE: $\nu_\mu + Ar \rightarrow p + \mu^-$

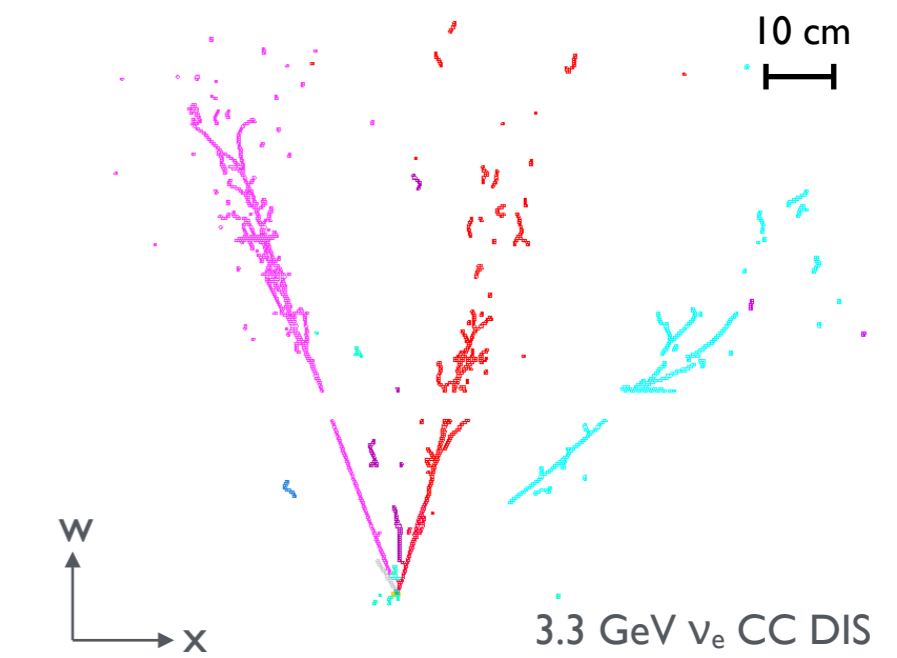
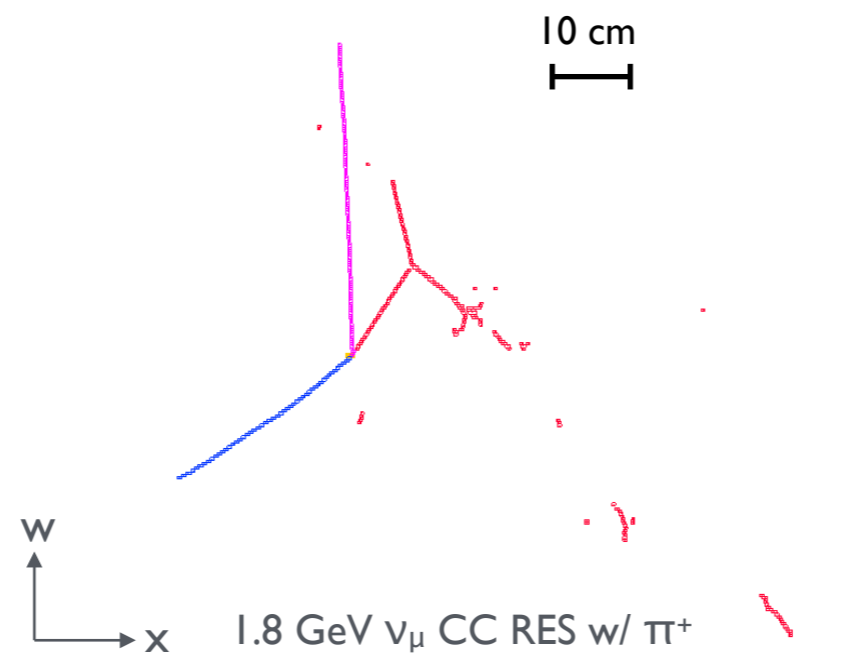
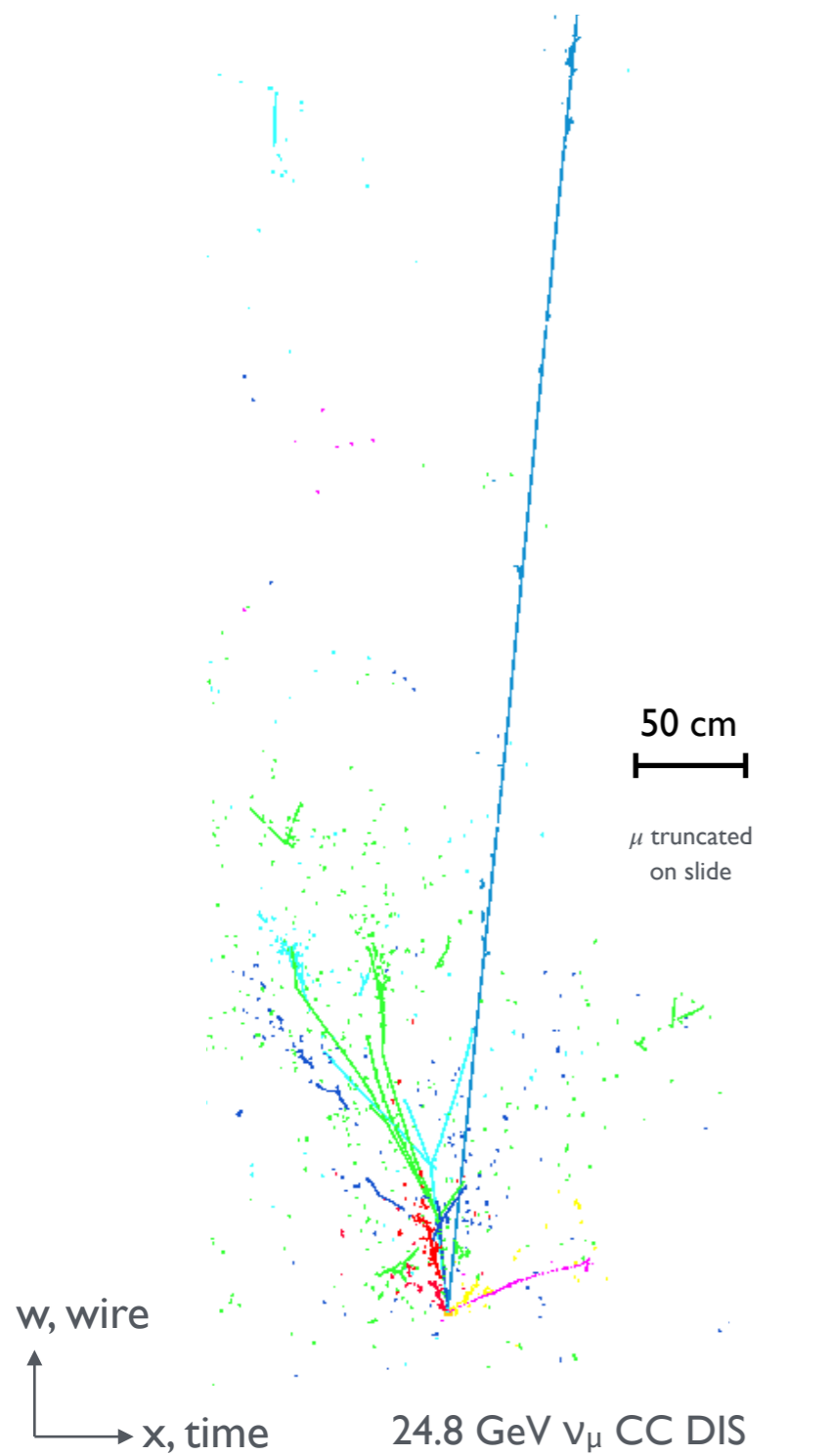


← Three 2D representations with common x coordinate, derived from drift time

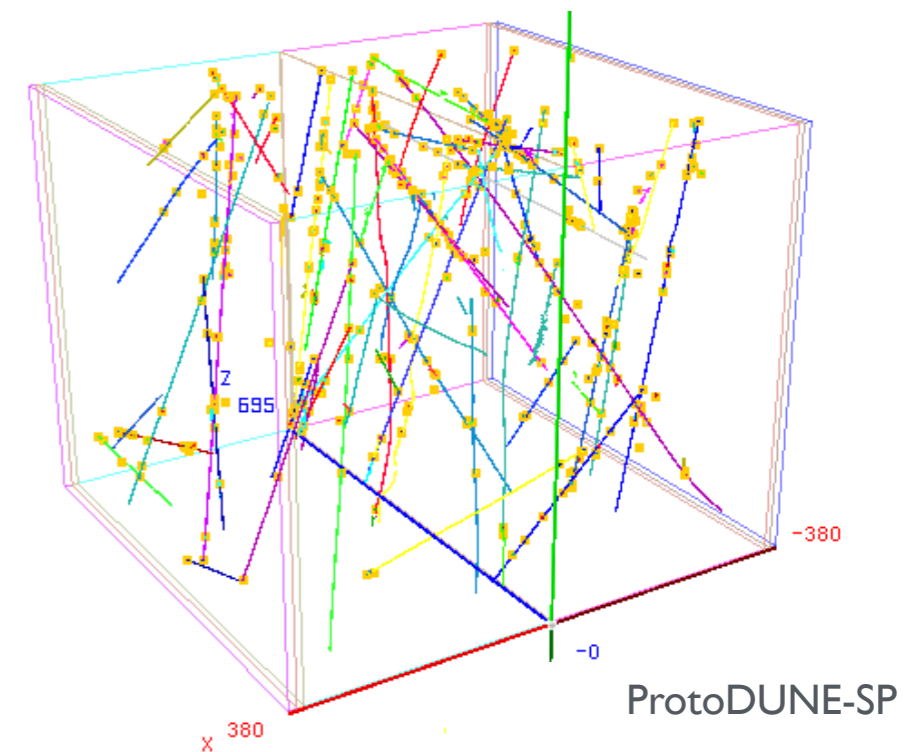
Pattern-Recognition Challenges

It is a significant challenge to develop automated, algorithmic LArTPC pattern recognition

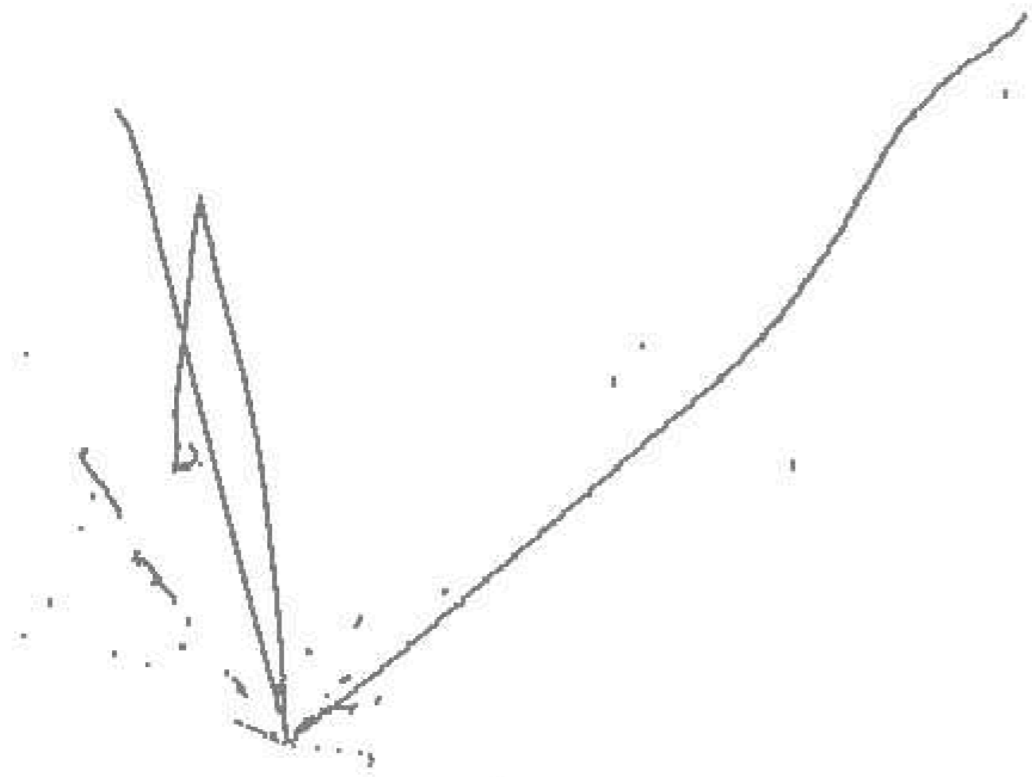
- Complex, diverse topologies:



- Also, LArTPCs have long exposures, due to lengthy drift times (up to few ms).
- Significant cosmic-ray muon background in surface-based detectors.



Examples: Collection Plane Images



2.8 GeV ν_μ CC DIS

w, wire
↑
x, time
→



3.2 GeV ν_e CC DIS

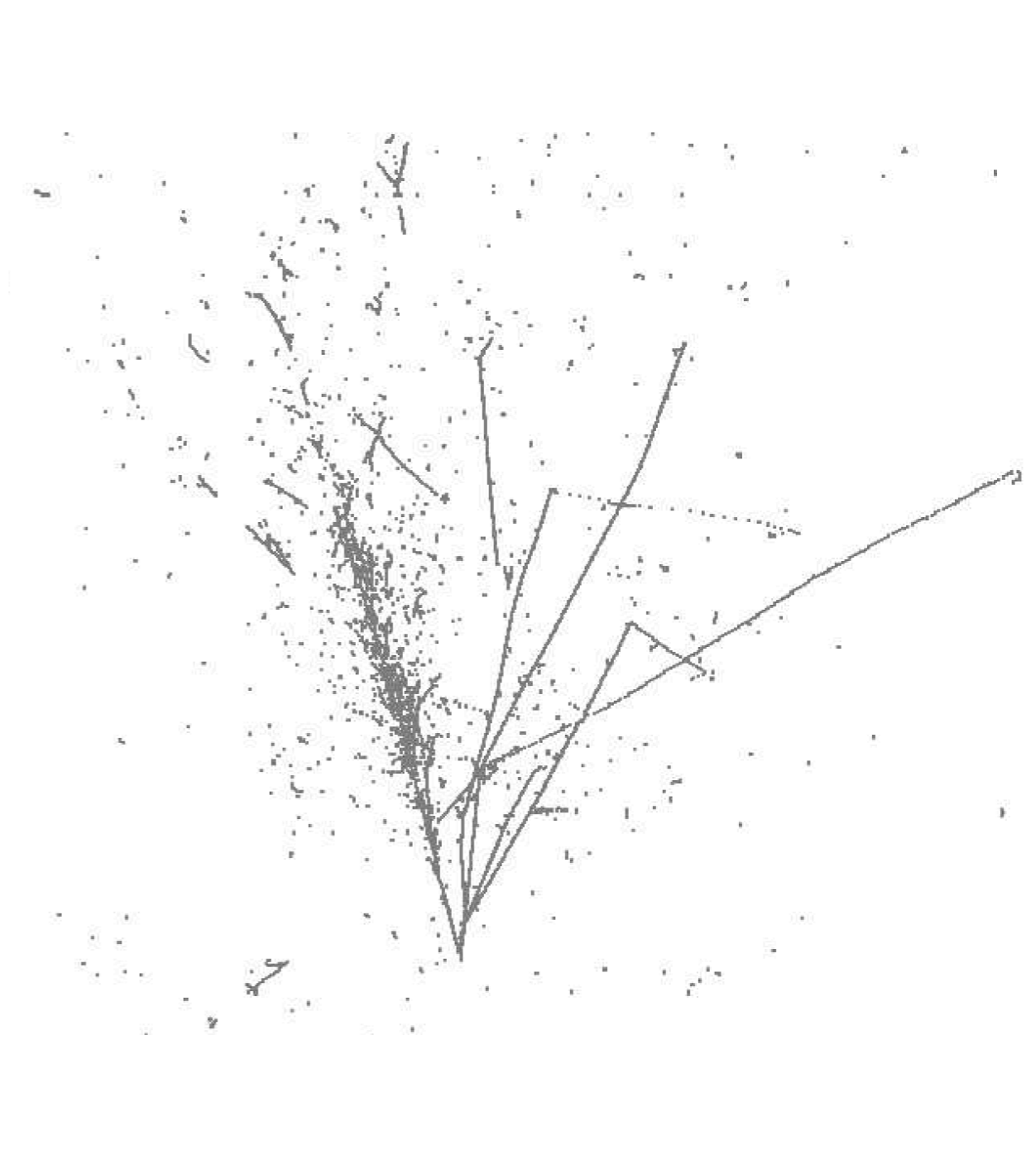
w, wire
↑
x, time
→

Examples: Collection Plane Images



26.8 GeV ν_μ CC DIS

w, wire
↑
x, time
→

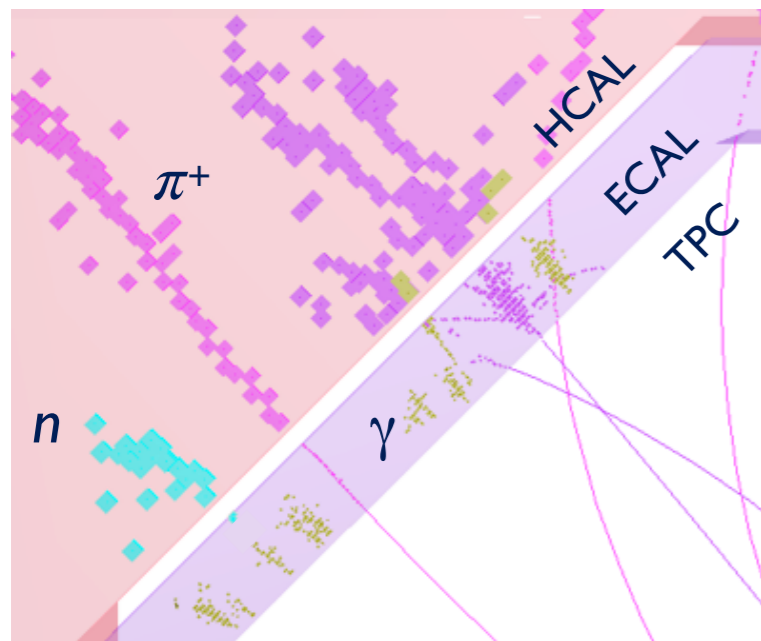


12.7 GeV ν_e CC DIS

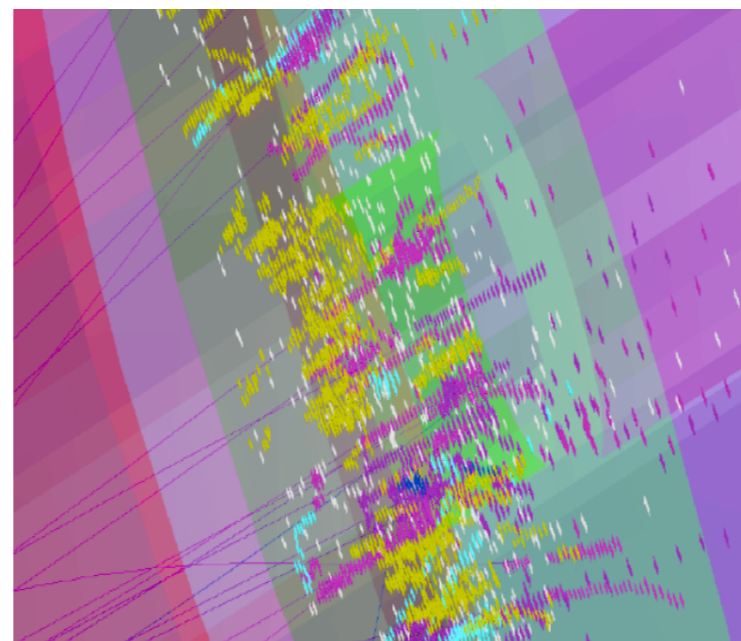
w, wire
↑
x, time
→

Pandora Multi-Algorithm Approach

- Single clustering approach is unlikely to work for such complex topologies:
 - Mix of track-like and shower-like clusters
- **Pandora** project has tackled similar problems before, using a multi-algorithm approach:
 - Build up events gradually
 - Each step is incremental - aim not to make mistakes (undoing mistakes is hard...)
 - Deploy more sophisticated algorithms as picture of event develops
 - Build physics and detector knowledge into algorithms

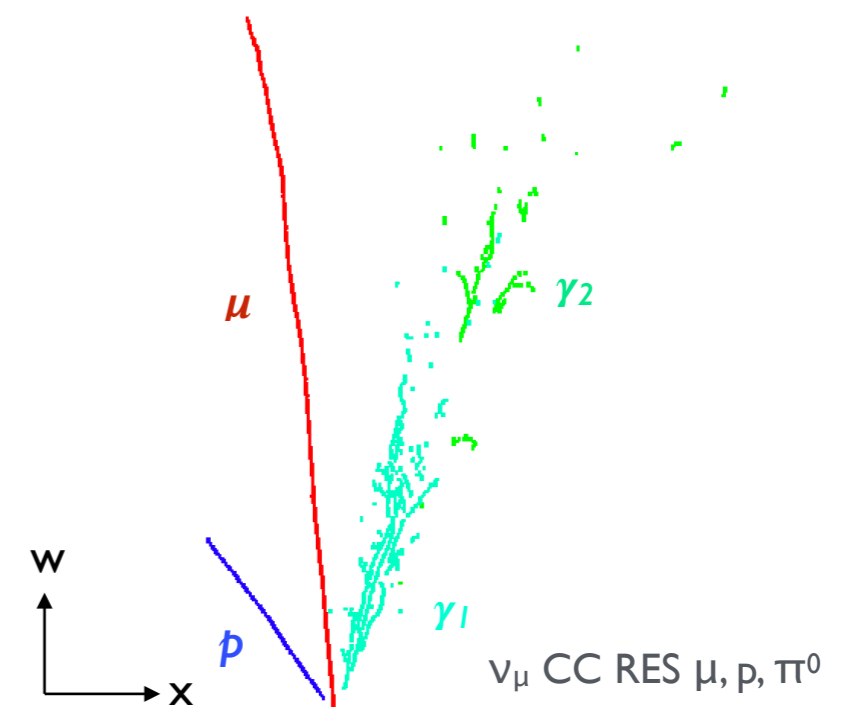


Typical ILC event topologies - 3D
NIMA.2009.09.009 NIMA.2012.10.038



Typical showers in CMS HGCal - 3D
LHCC-P-008

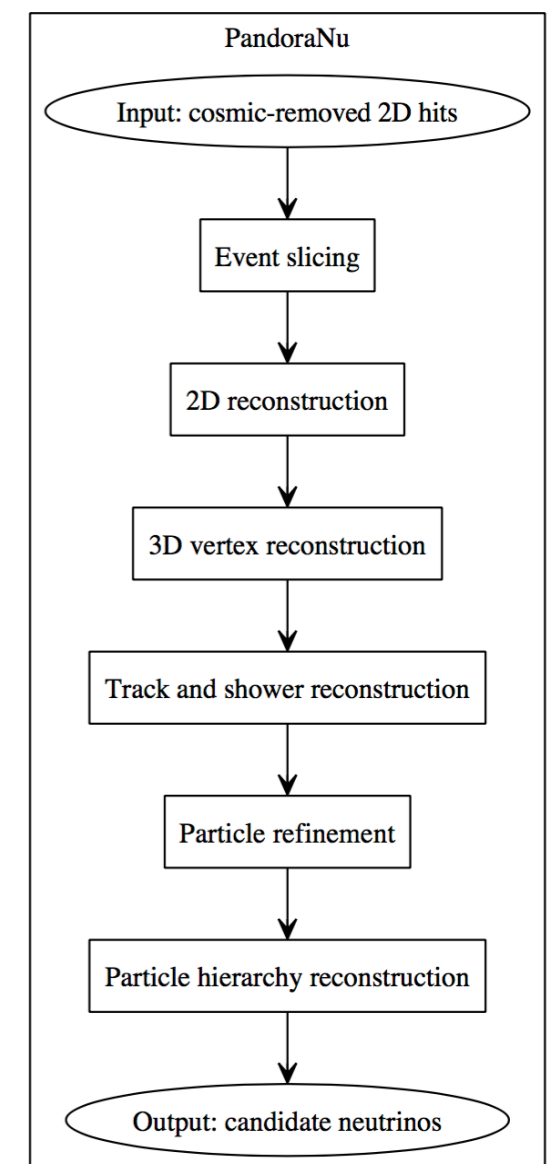
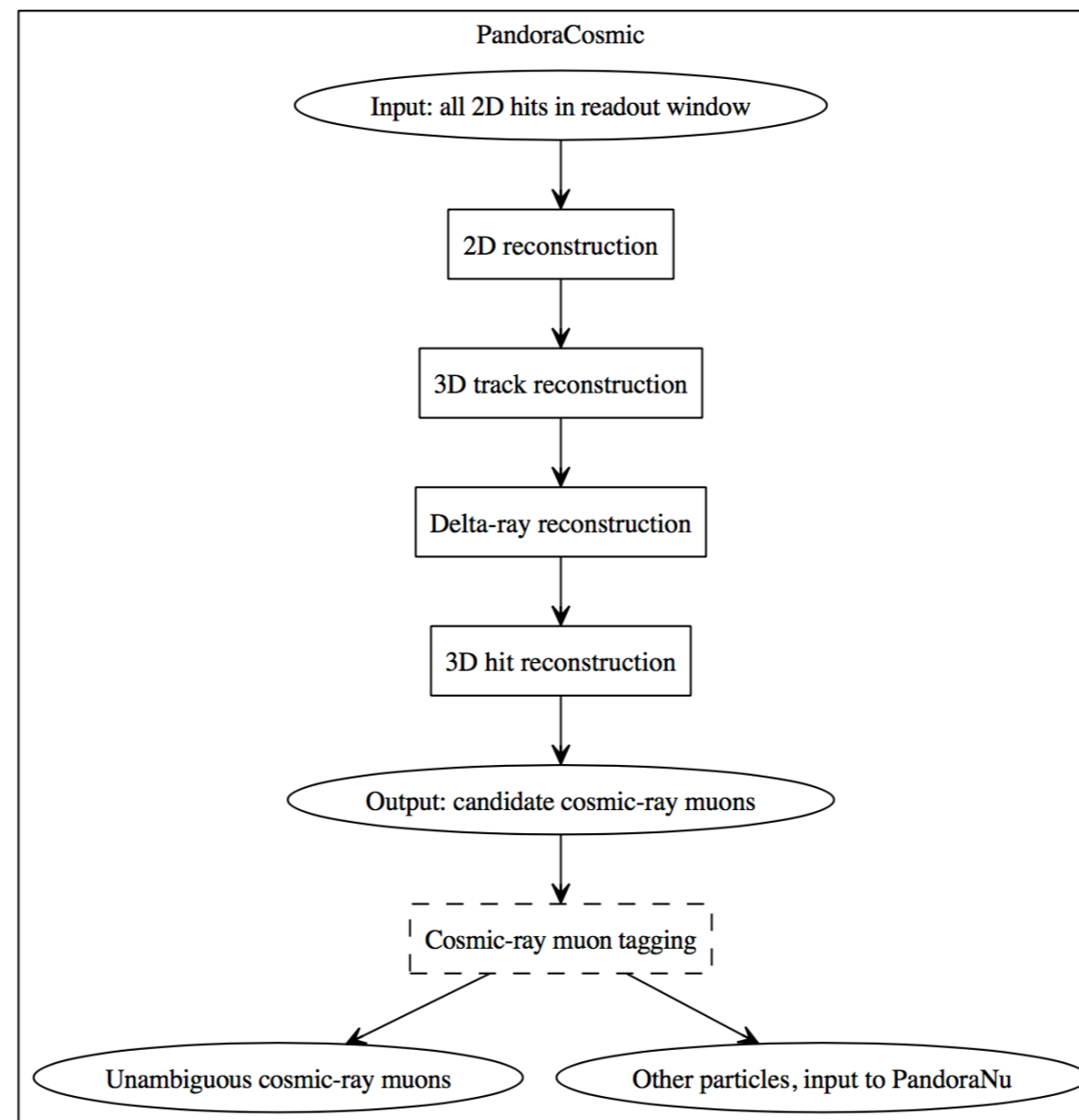
BNB interaction at MicroBooNE - 3 x 2D



Pandora Algorithm Chains

- Two algorithm chains (>140 algs) created for LArTPC use, with many algs in common:
 - **PandoraCosmic**: strongly track-oriented; showers assumed to be delta rays, added as daughters of primary muons; muon vertices at track high-y coordinate.
 - **PandoraNu**: finds neutrino interaction vertex and protects all particles emerging from vertex position. Careful treatment to address track/shower tensions.

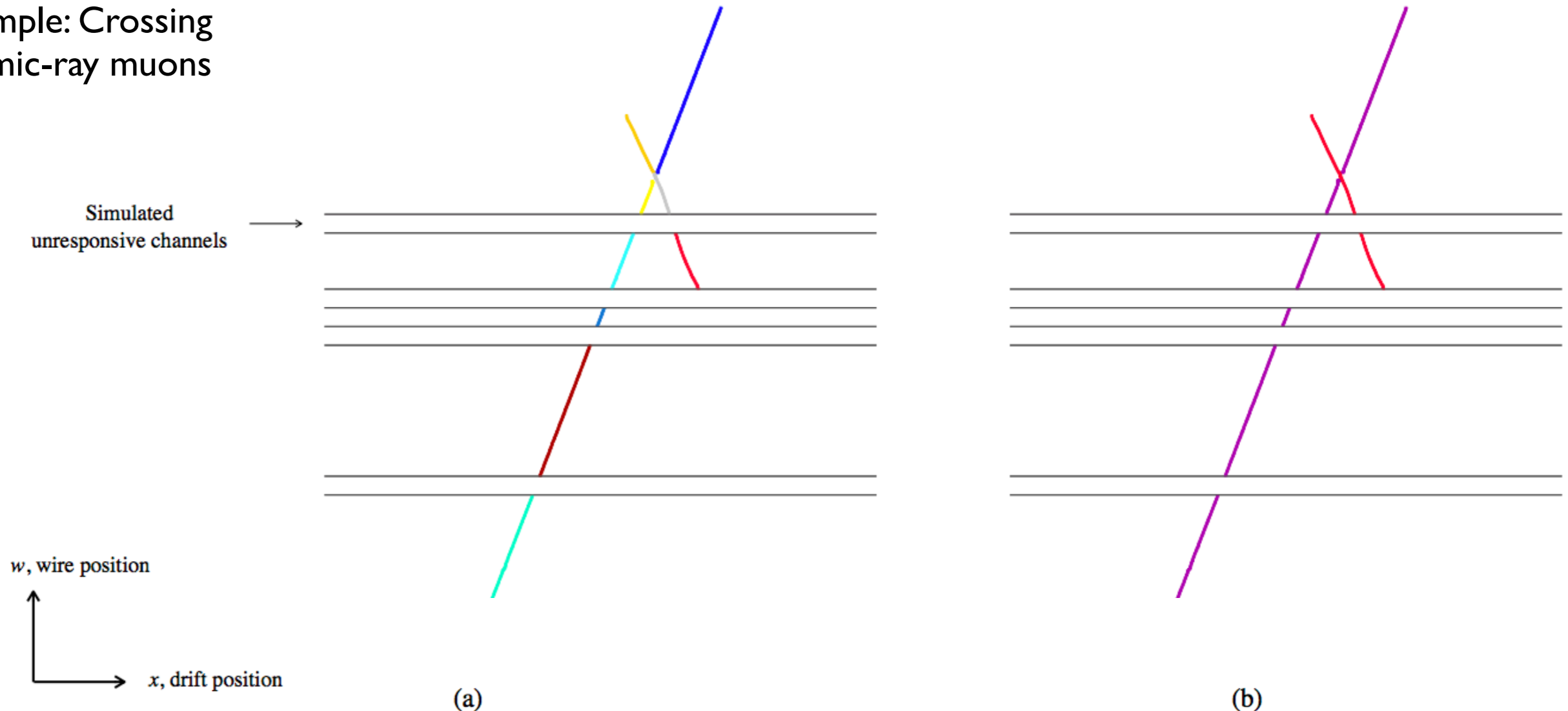
Initially use a two-pass approach:
Input to PandoraNu excludes hits from unambiguous cosmic rays.



Cosmic-Ray Muon Reconstruction - 2D

- For each plane, produce list of 2D clusters that represent continuous, unambiguous lines of hits:
 - Separate clusters for each structure, with clusters starting/stopping at each branch or ambiguity.
- Clusters refined by series of 16 cluster-merging and cluster-splitting algs that use **topological info**.

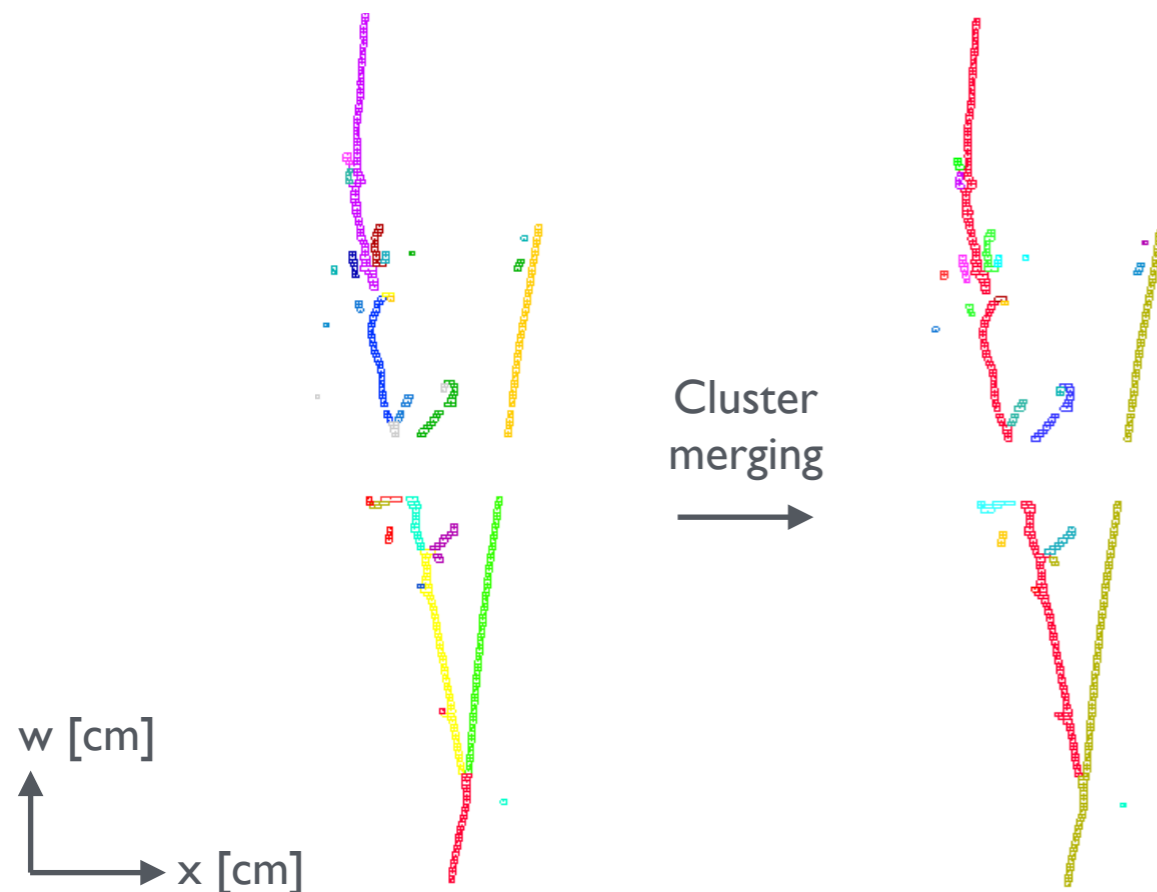
Example: Crossing cosmic-ray muons



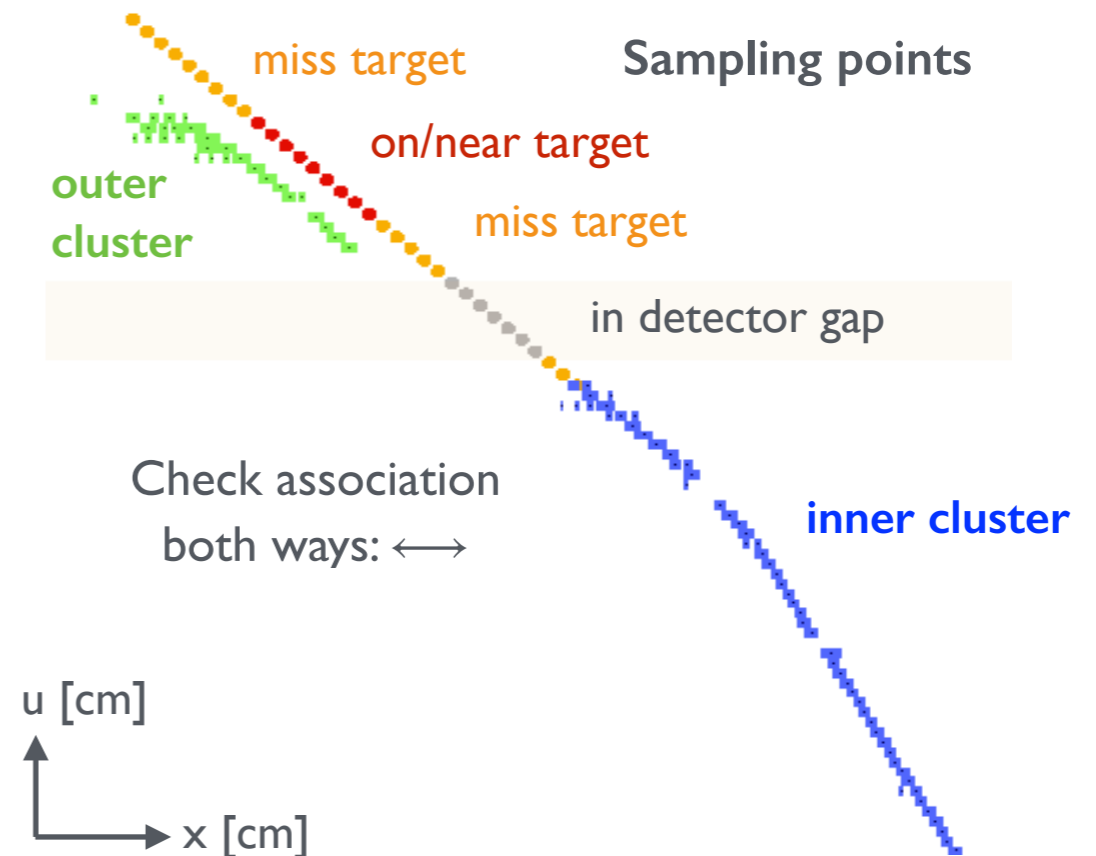
Topological Association - 2D

- **Cluster-merging algorithms identify associations between multiple 2D clusters and look to grow the clusters to improve completeness, without compromising purity.**
 - The challenge for the algorithms is to make cluster-merging decisions in the context of the entire event, rather than just considering individual pairs of clusters in isolation.
 - Typically need to provide a definition of association (for a given pair of clusters), then navigate forwards and backwards to identify chains of associated clusters that can be safely merged.

E.g. LongitudinalAssociation

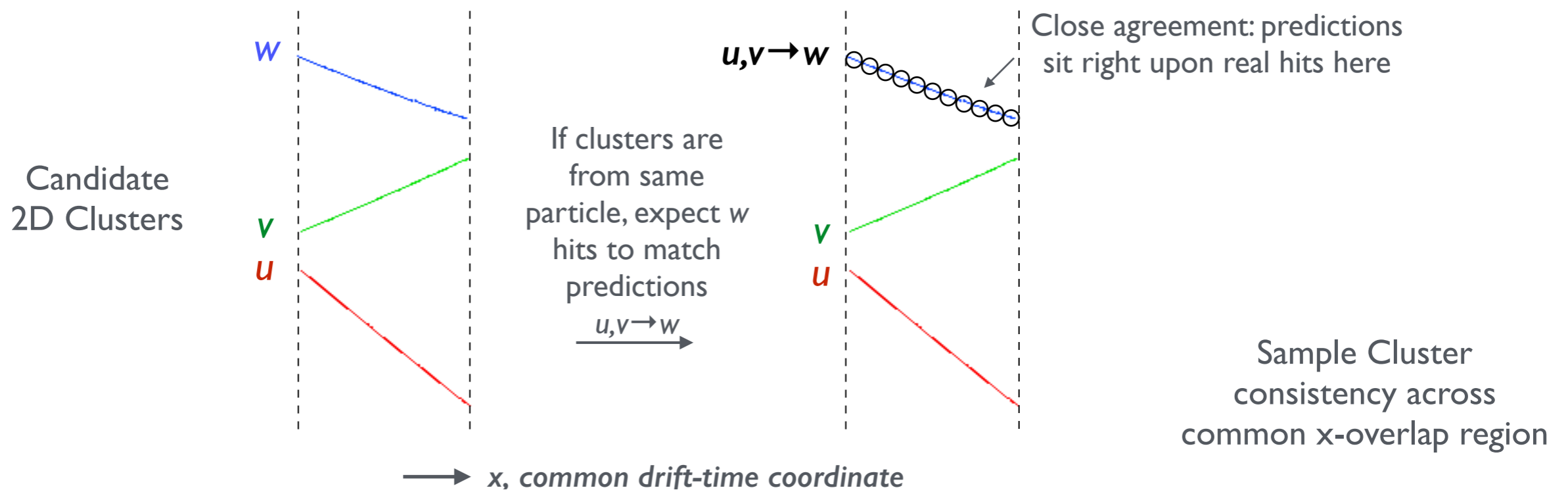


E.g. CrossGapsAssociation



Track Pattern Recognition - 3D

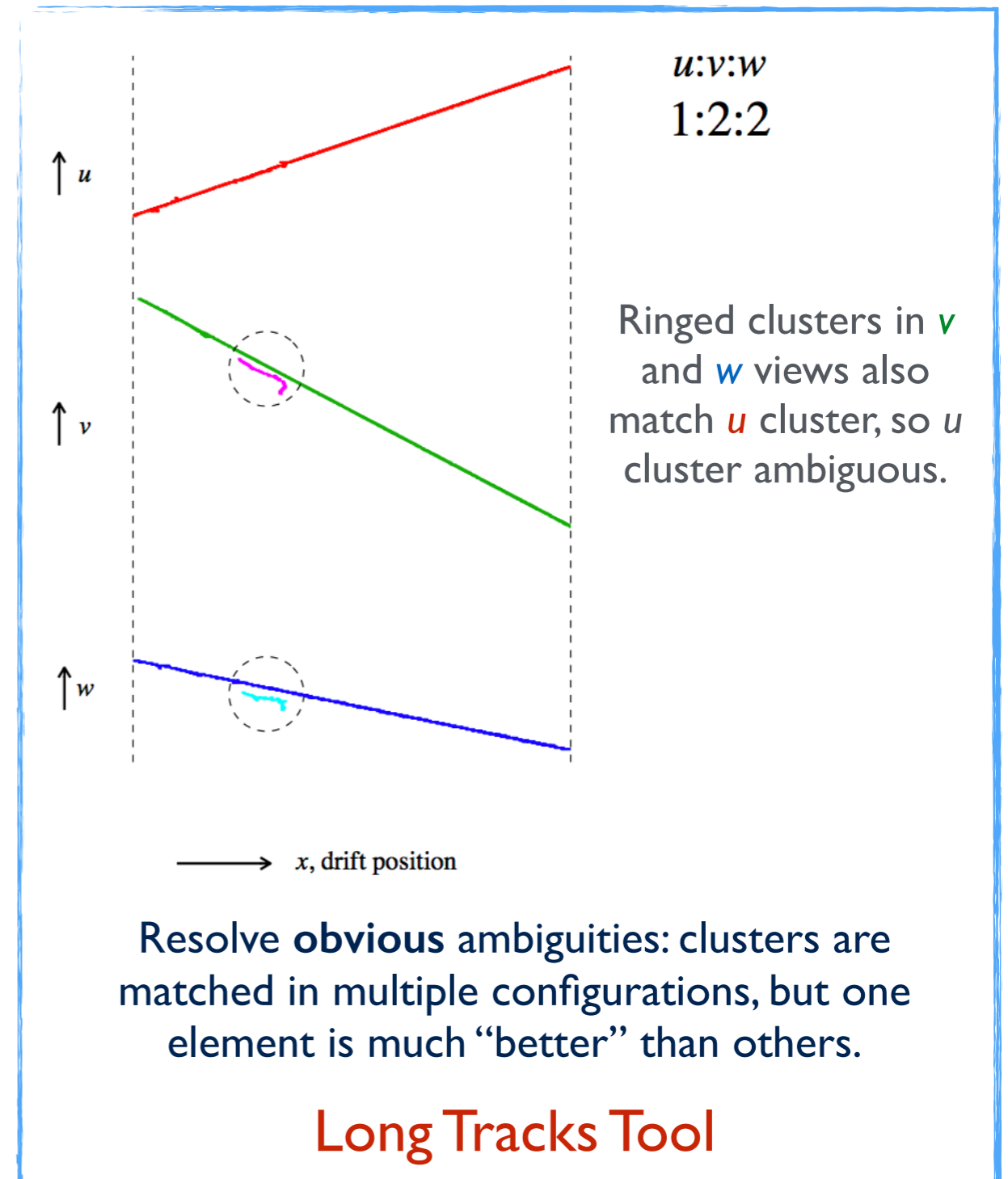
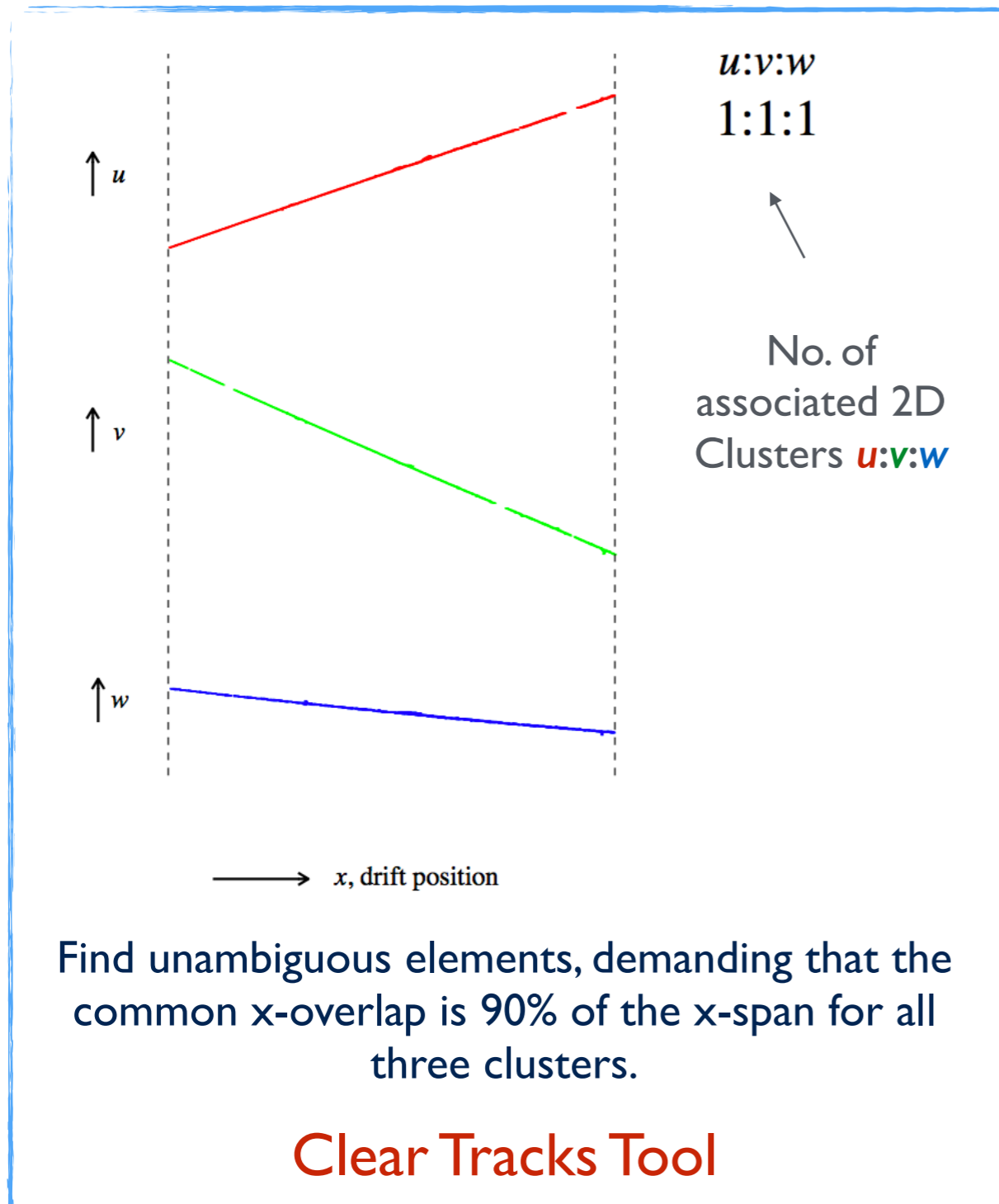
- Our original input was 3x2D images of charged particles in the detector.
- Should now have reconstructed three separate 2D clusters for each particle:
 - Compare 2D clusters from u, v, w planes to find the clusters representing same particle.
 - Exploit common drift-time coordinate and our understanding of wire plane geometry.
 - At given x , compare predictions $\{u, v \rightarrow w; v, w \rightarrow u; w, u \rightarrow v\}$ with cluster positions, calculating χ^2



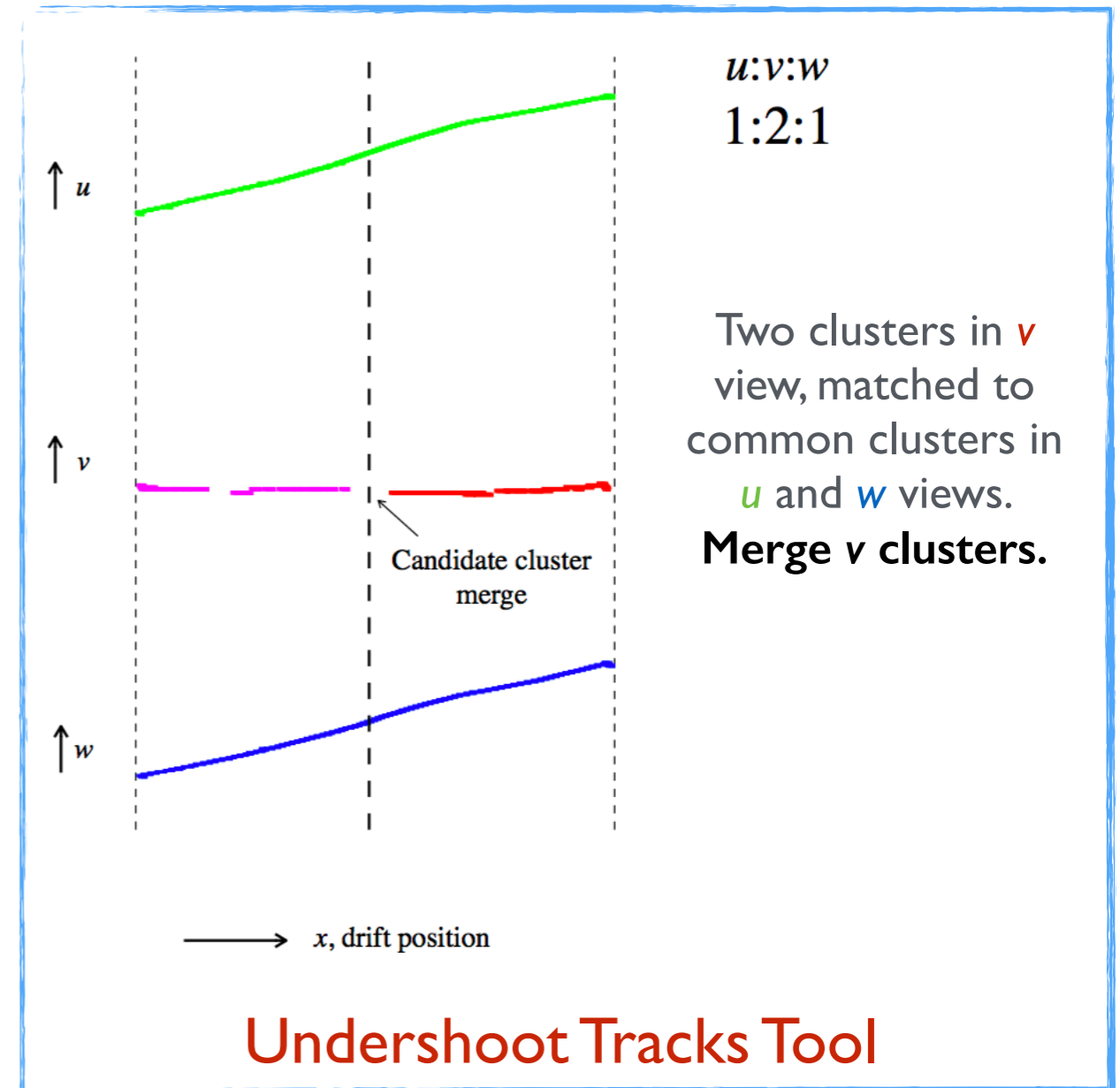
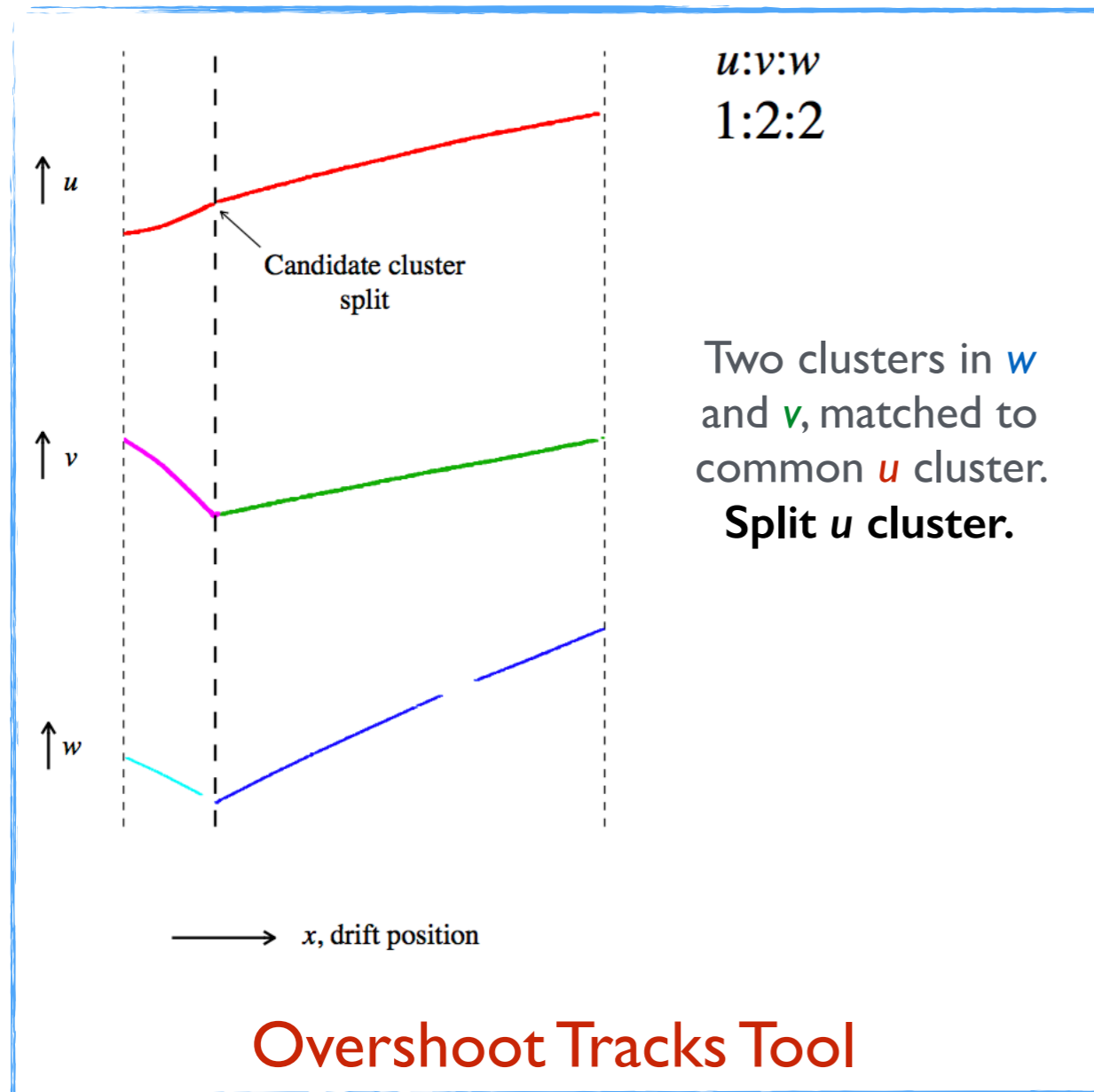
Store all results in a 3D array, recording x -overlap span, no. of sampling points, no. of "matched" sampling points and a χ^2 - documents all 2D cluster-matching ambiguities.

Track Pattern Recognition - 3D

3D array stores overlap details for all trios of 2D clusters. Tools make 2D reco changes to **resolve any ambiguities**. If a tool makes a change (e.g. splits a cluster), all tools run again.



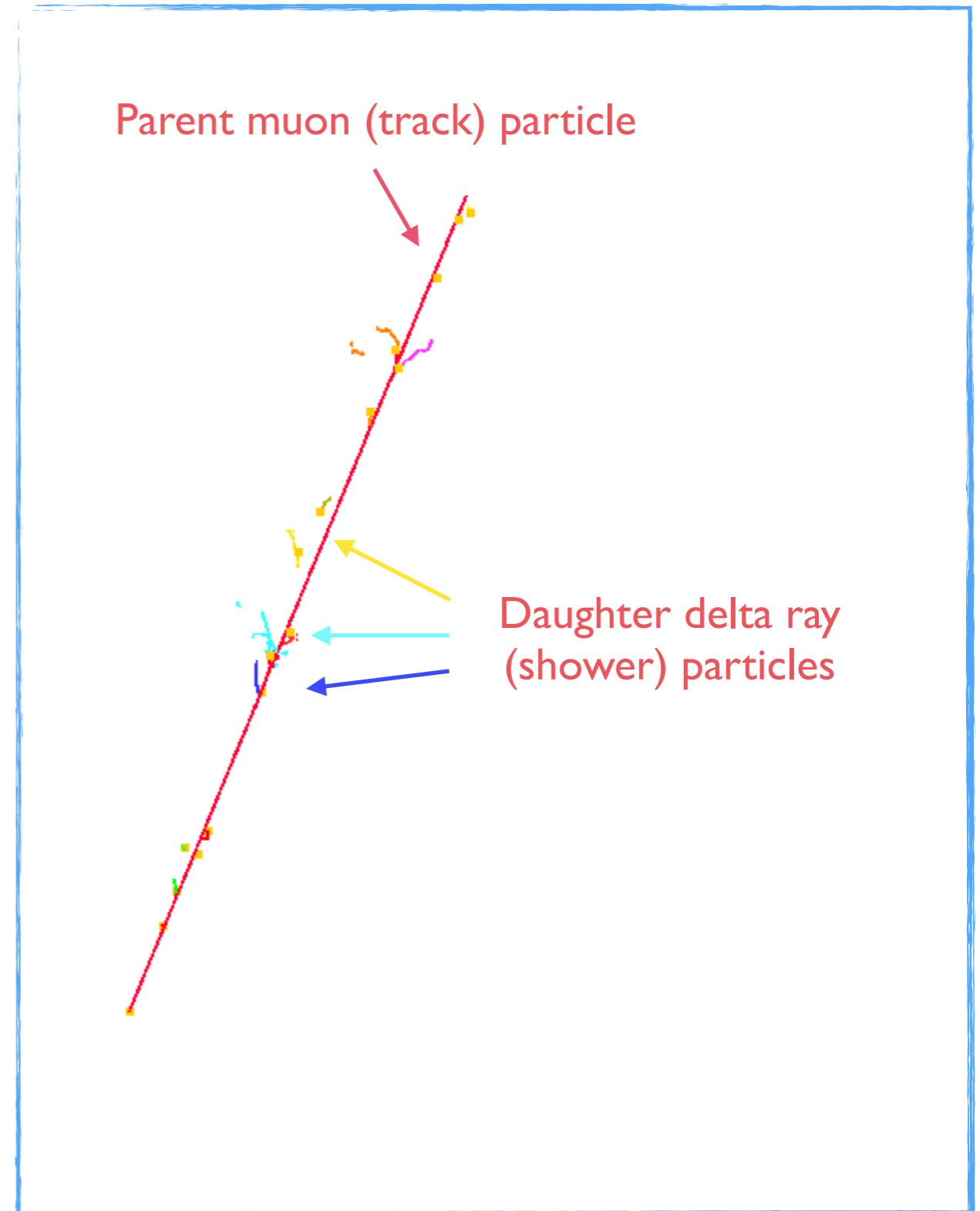
Track Pattern Recognition - 3D



- Use all connected clusters to assess whether this is a true 3D kink topology.
- Modify 2D clusters as appropriate (i.e. merge or split) and update cluster-matching details.
- Initial ClearTracks tool then able to identify unambiguous groupings of clusters and form particles.

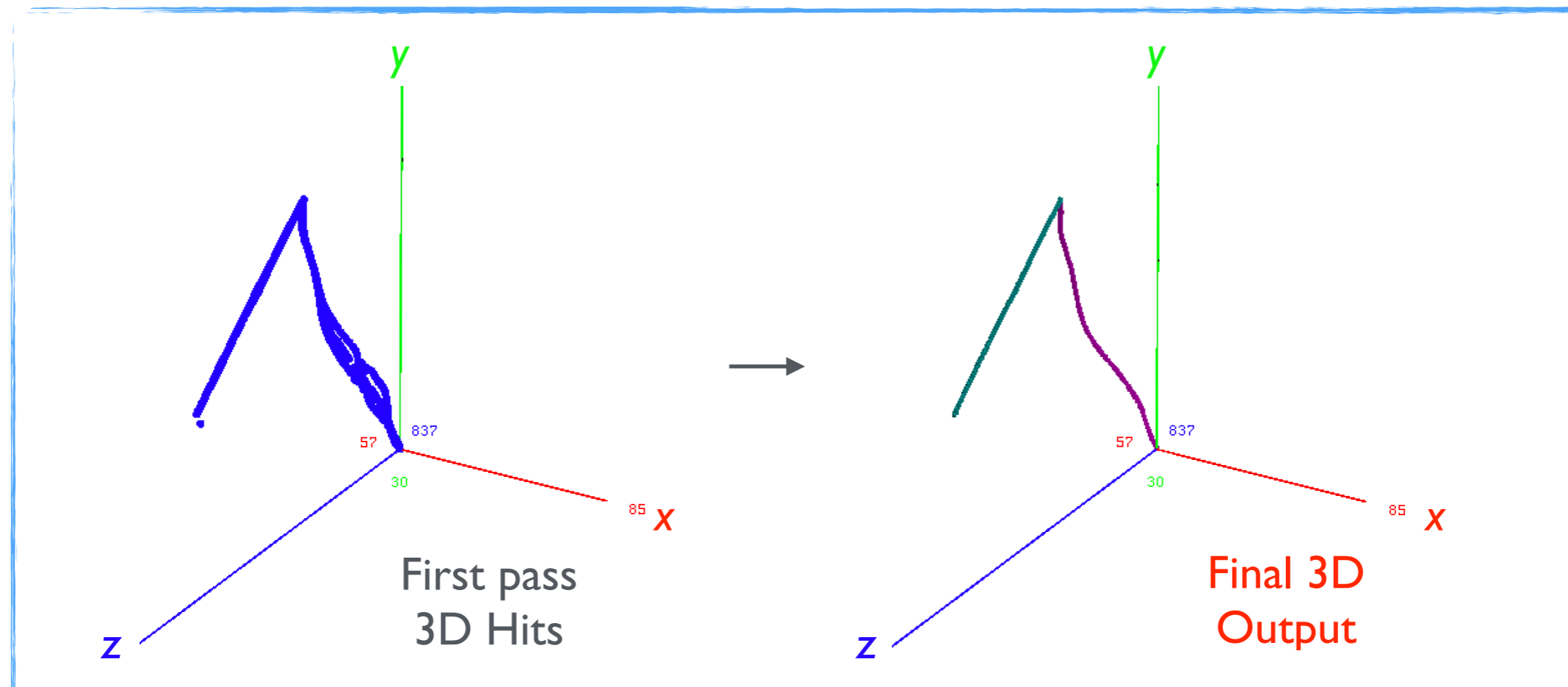
Delta-Ray Reconstruction - 2D, 3D

- Assume any 2D clusters not in a track particle are from delta-ray showers:
 - Simple proximity-based reclustering of hits, then topological association algs.
 - Delta-ray clusters matched between views, creating delta-ray shower particles.
 - Parent muon particles identified and delta-ray particles added as daughters.

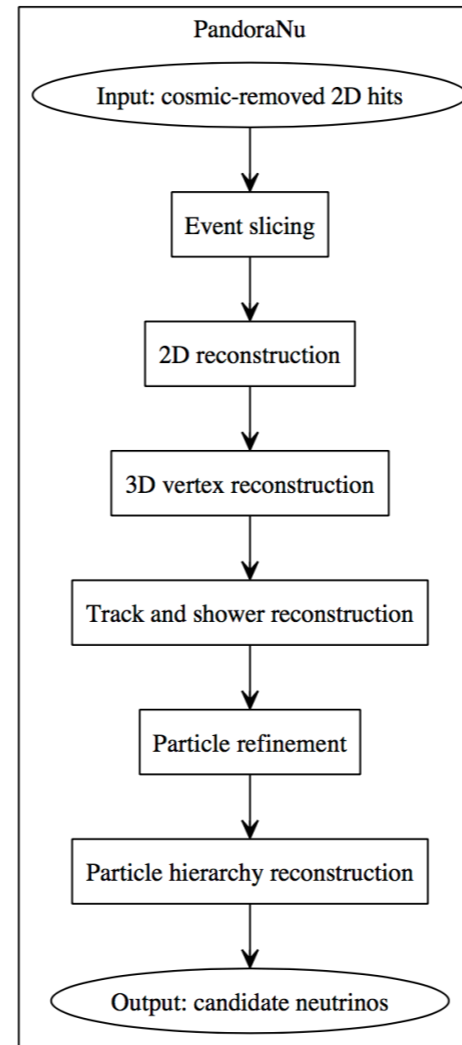
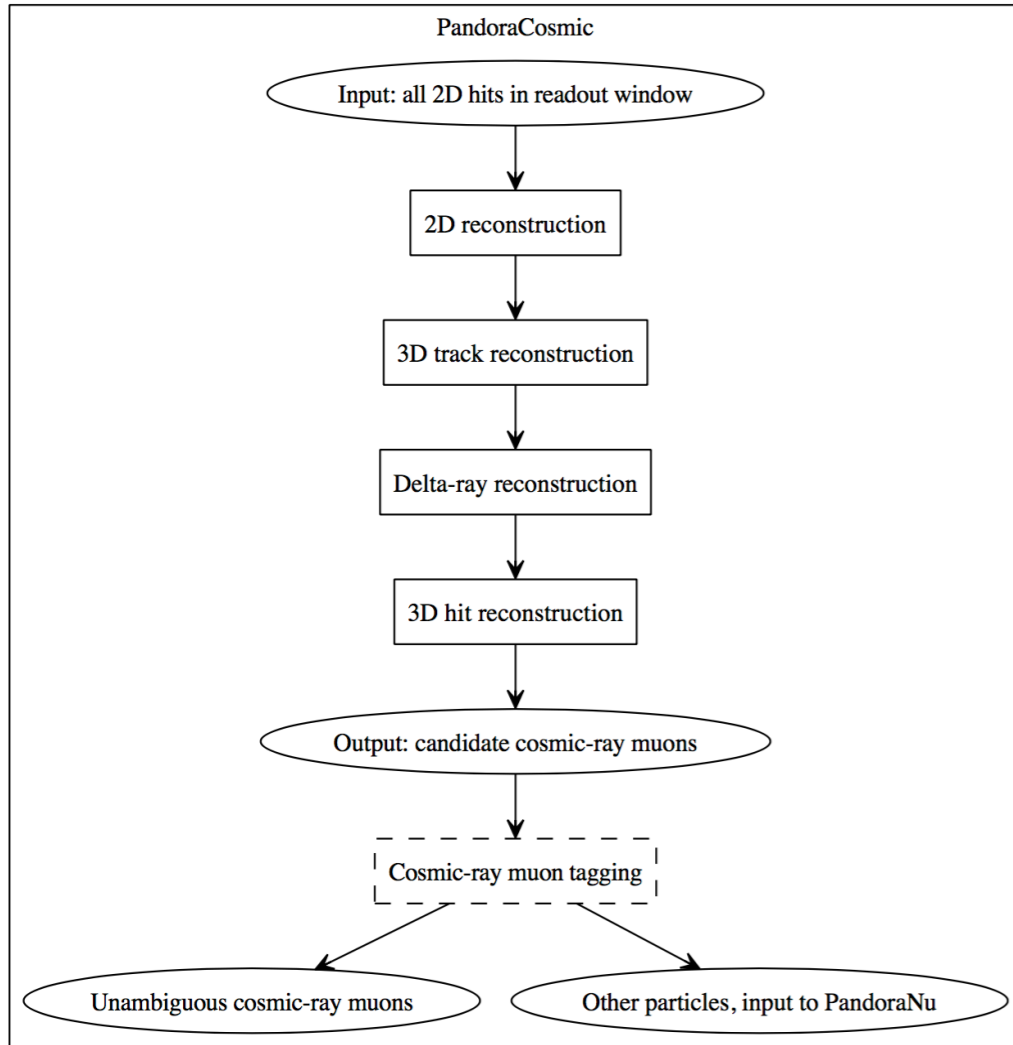


3D Hit/Cluster Reconstruction

- For each 2D Hit, sample clusters in other views at same x , to provide u_{in} , v_{in} and w_{in} values
- Provided u_{in} , v_{in} and w_{in} values don't necessarily correspond to a specific point in 3D space
- Analytic expression to find 3D space point that is *most consistent* with given u_{in} , v_{in} and w_{in}
 - $\chi^2 = (u_{out} - u_{in})^2 / \sigma_u^2 + (v_{out} - v_{in})^2 / \sigma_v^2 + (w_{out} - w_{in})^2 / \sigma_w^2$
 - Write in terms of unknown y and z , differentiate wrt y, z and solve
 - Can iterate, using fit to current 3D hits (extra terms in χ^2) to produce smooth trajectory

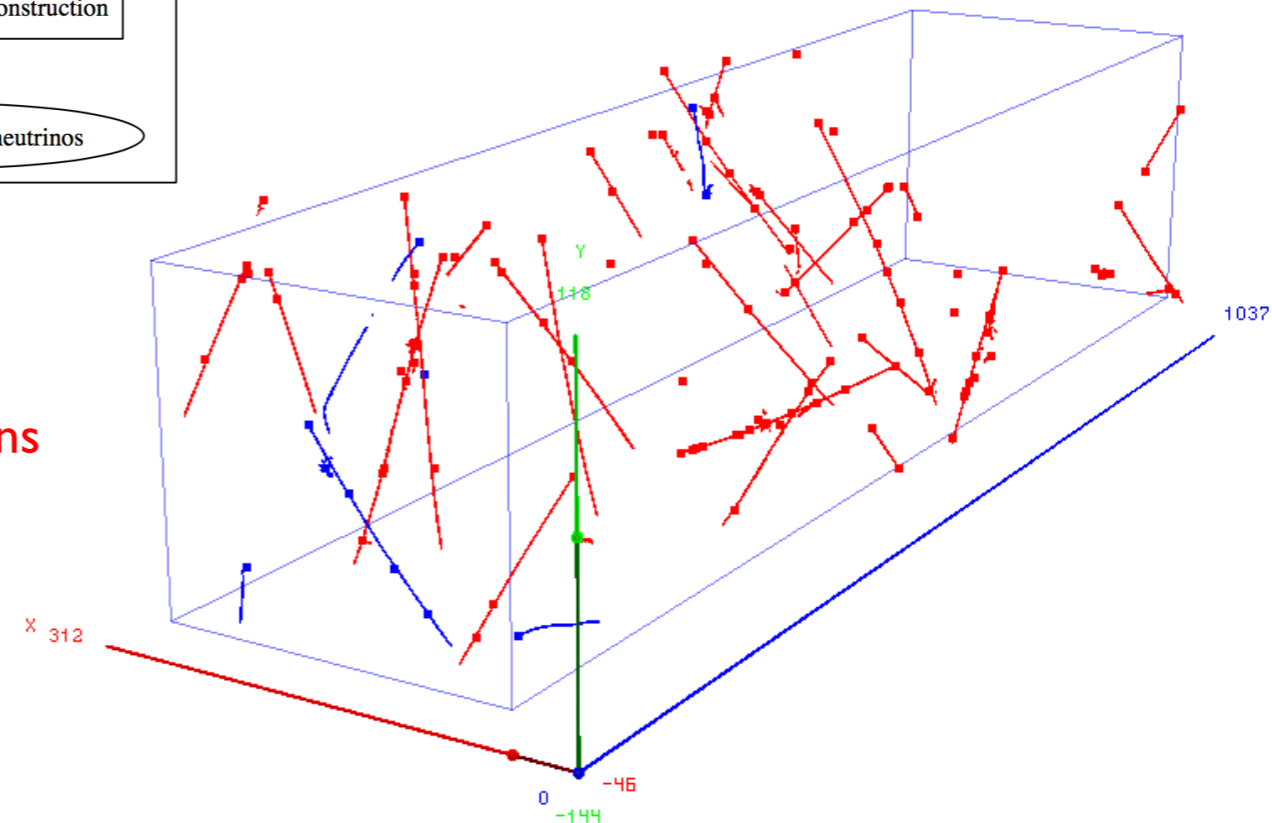


PandoraCosmic → PandoraNu



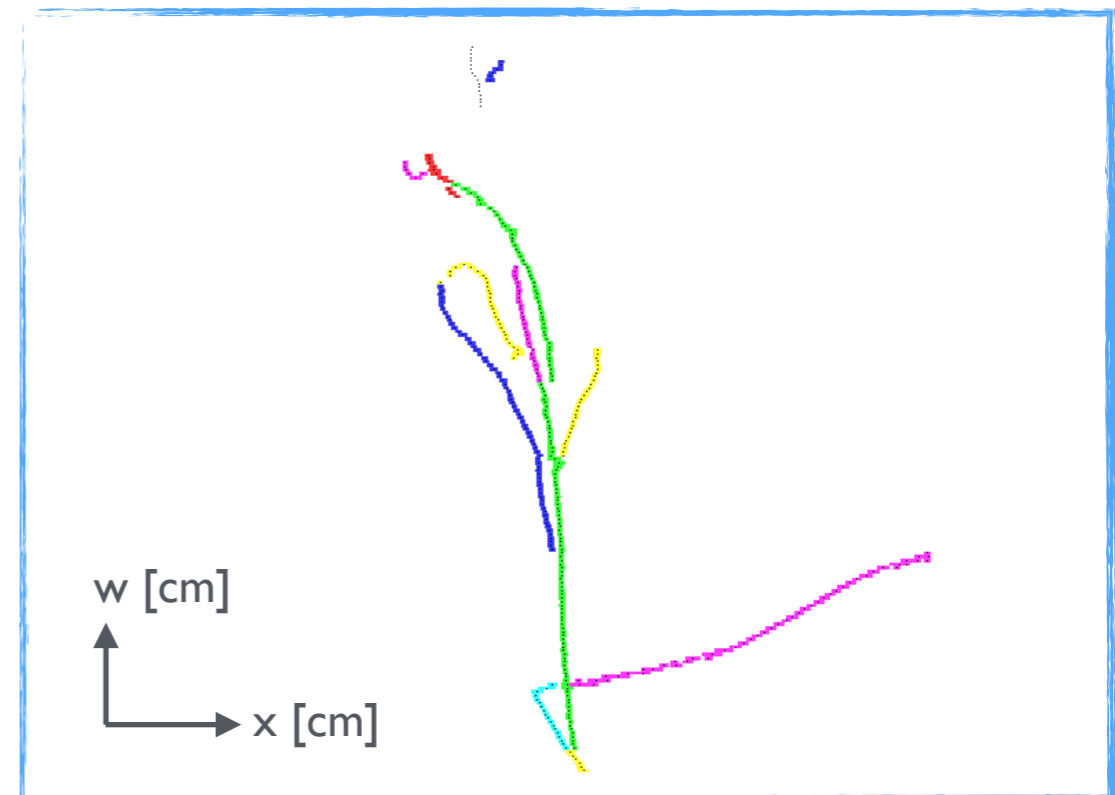
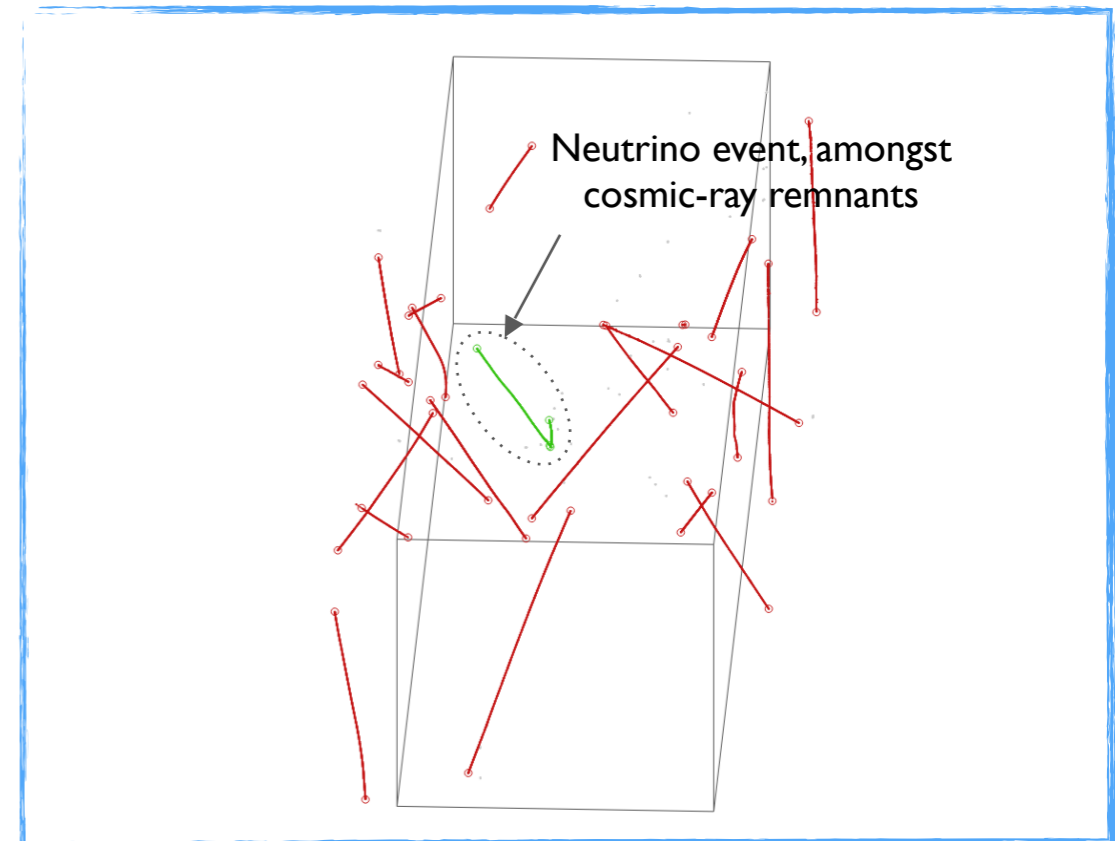
Unambiguous
cosmic-ray muons

Other particles, input
to PandoraNu



Neutrino Reconstruction

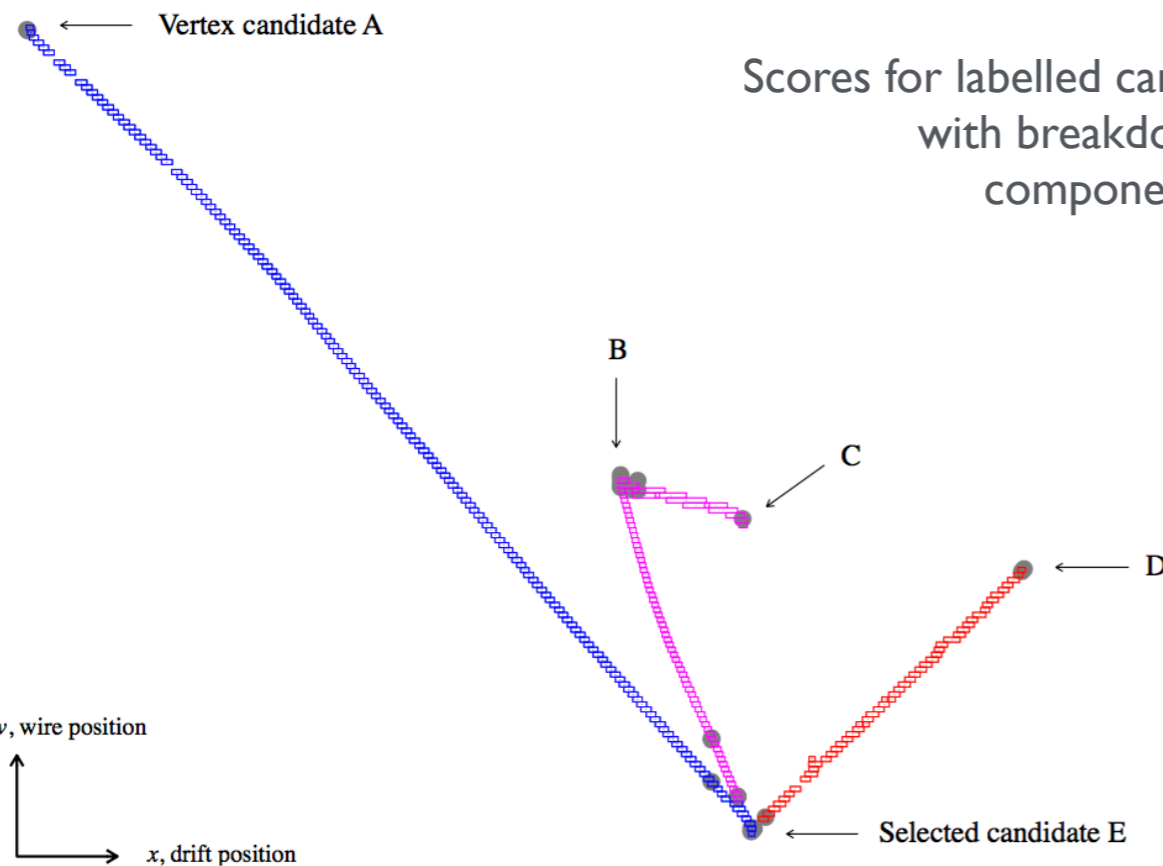
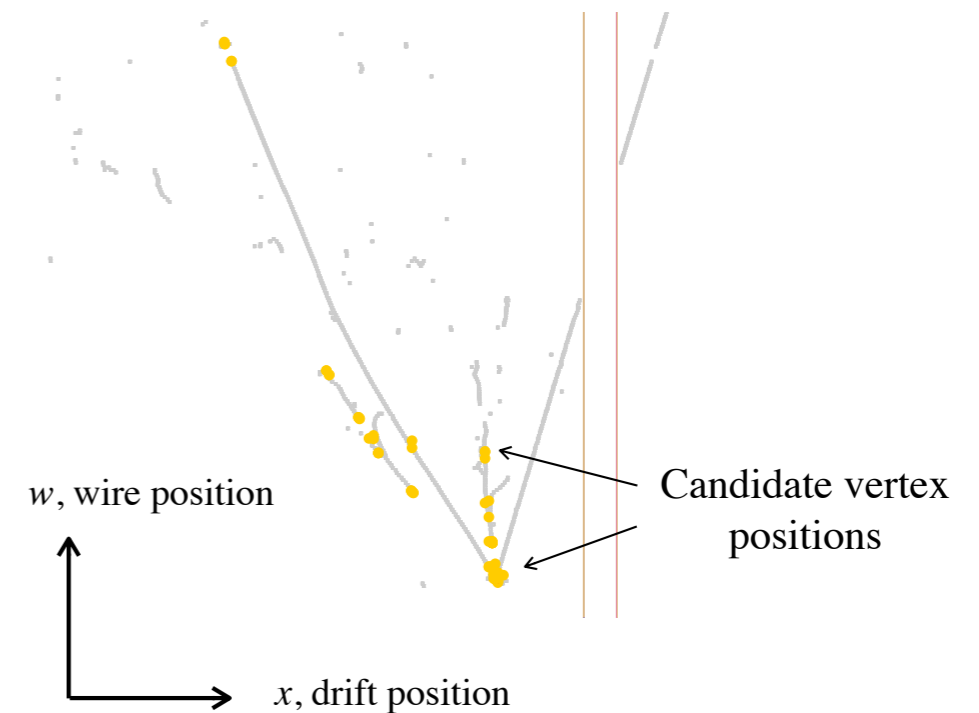
- **Must be able to deal with presence of any cosmic-ray muon remnants.**
 - Run fast version of reconstruction, up to 3D hit creation
 - “Slice” 3D hits into separate interactions, processing each slice in isolation.
 - Each slice \Rightarrow candidate neutrino particle.
- **Neutrino pass reuses track-oriented clustering and topological association.**
 - Topological association algs must handle rather more complex topologies.
 - Specific effort to reconstruct neutrino interaction vertex.
 - More sophisticated efforts to reconstruct showers.



Vertex Reconstruction - 3D

Search for neutrino interaction vertex:

1. Use pairs of 2D clusters to produce list of possible 3D vertex candidates.
2. Examine candidates, calculate a score for each and select the best.
 - Selection uses Boosted Decision Trees and a Convolutional Neural Network.



Scores for labelled candidates, with breakdown into component parts:

Candidate	S	$S_{\text{energy kick}}$	$S_{\text{asymmetry}}$	$S_{\text{beam dweight}}$
A	4.9E-07	3.5E-06	1.00	0.14
B	1.3E-02	3.1E-02	0.99	0.42
C	1.1E-03	2.4E-03	0.95	0.46
D	5.7E-10	1.1E-09	1.00	0.52
E	9.0E-01	9.0E-01	1.00	0.99

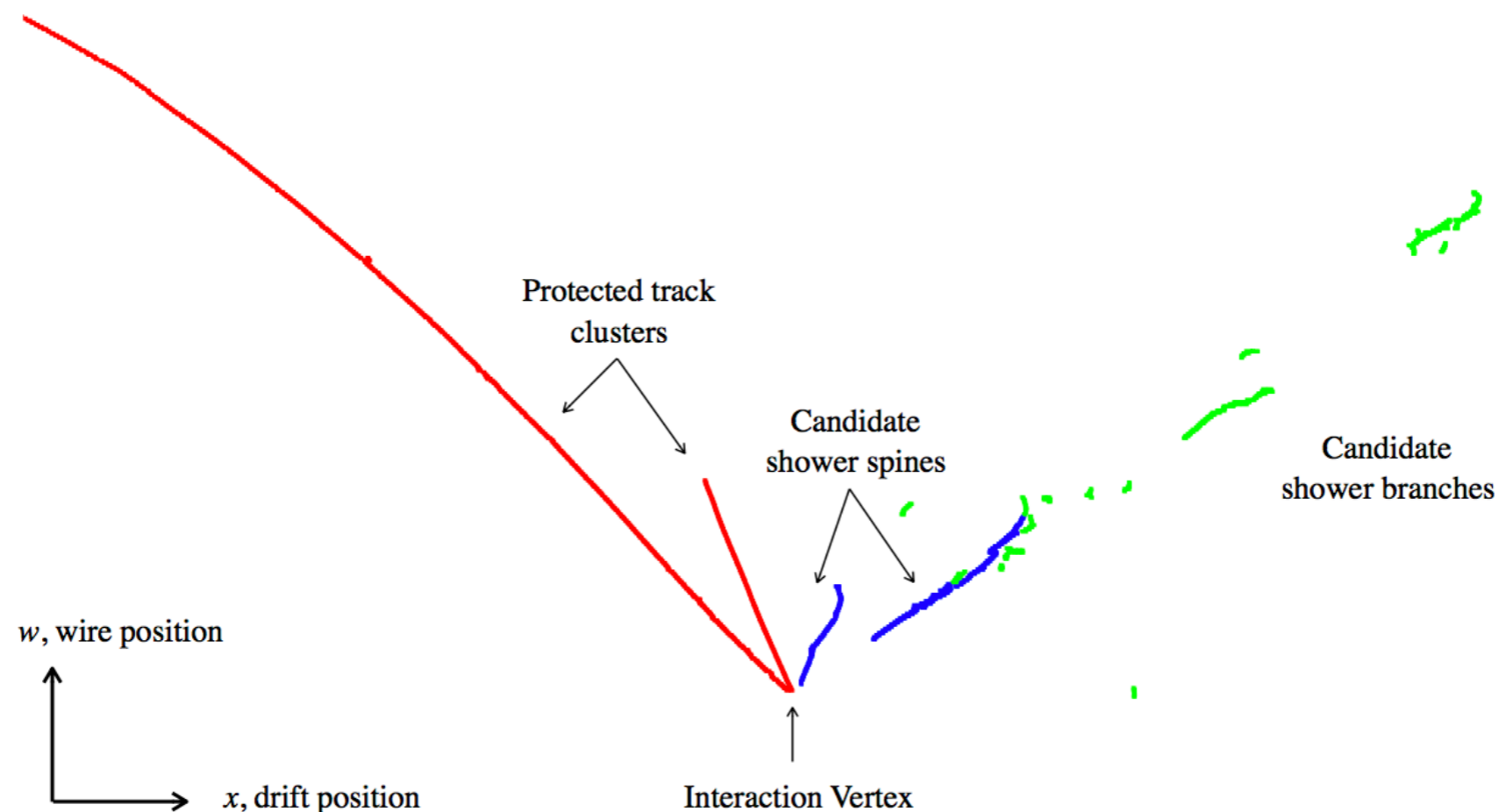
Downstream usage:

- Split 2D clusters at projected vertex position.
- Use vertex to protect primary particles when growing showers.

Shower Reconstruction - 2D

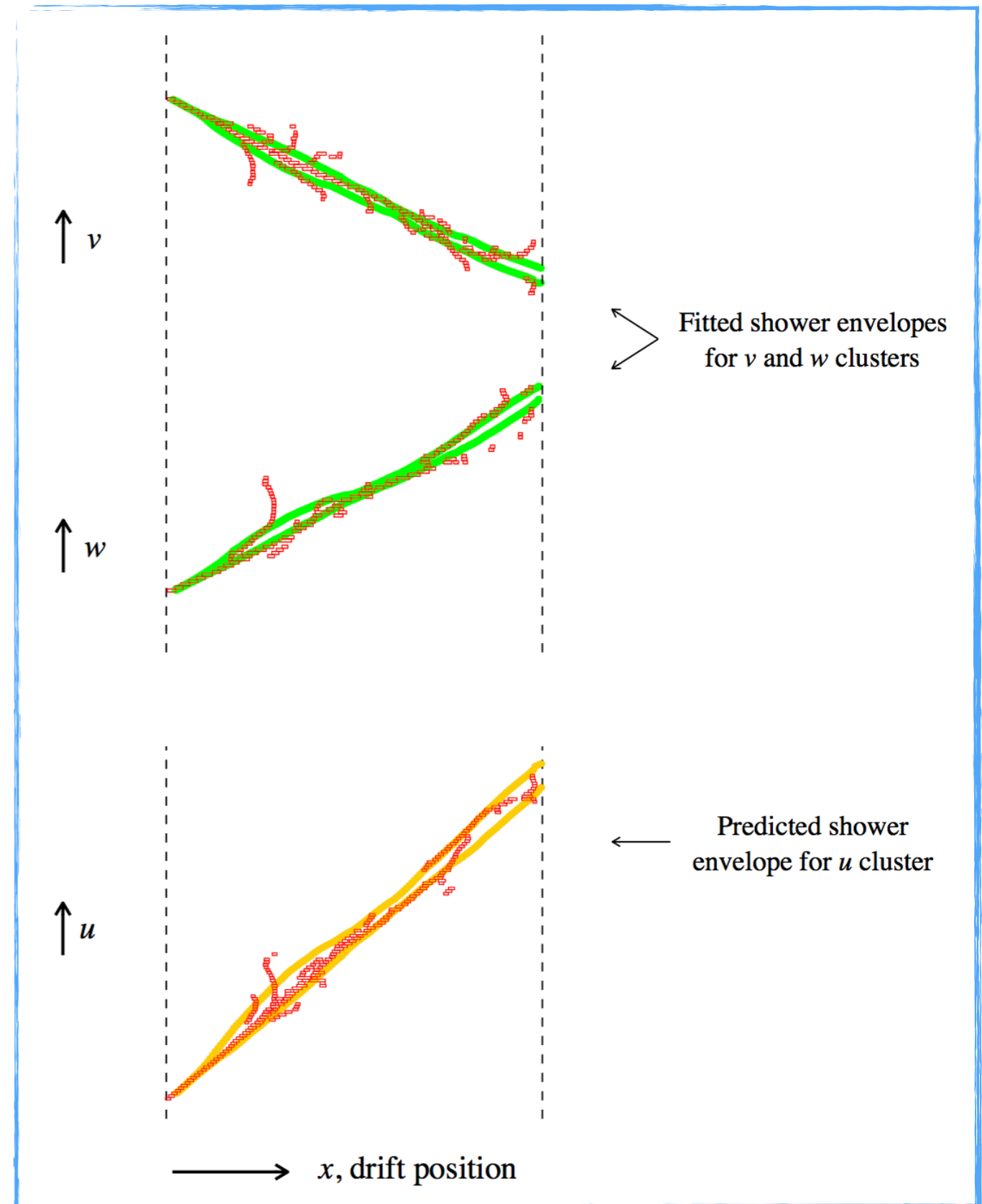
Track reconstruction exactly as in PandoraCosmic, but now also attempt to reconstruct primary electromagnetic showers, from electrons and photons:

- Characterise 2D clusters as track-like or shower-like, and use topological properties to identify clusters that might represent shower spines.
- Add shower-like branch clusters to shower-like spine clusters. Recursively identify branches on the top-level spine candidate, then branches on branches, etc.



Shower Reconstruction - 3D

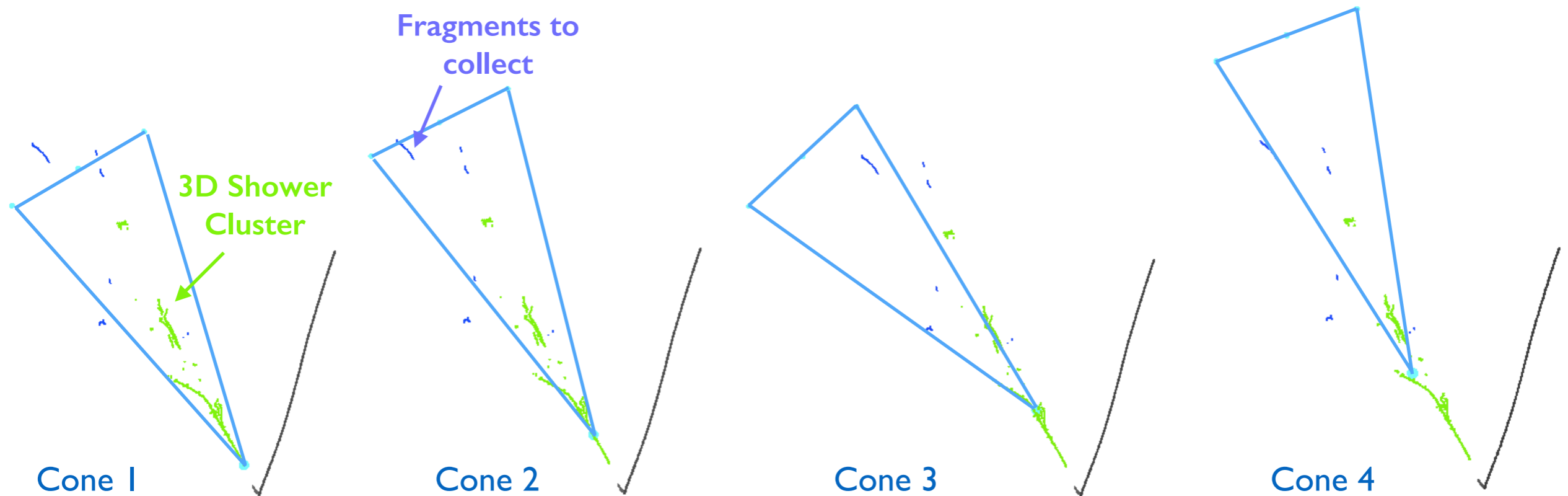
- Reuse ideas from track reco to match 2D shower clusters between views:
 - Build a 3D array to store cluster overlap and relationship information.
 - Overlap information collected by fitting shower envelope to each 2D cluster.
 - Shower edges from two clusters used to predict envelope for third cluster.



Particle Refinement - 2D, 3D

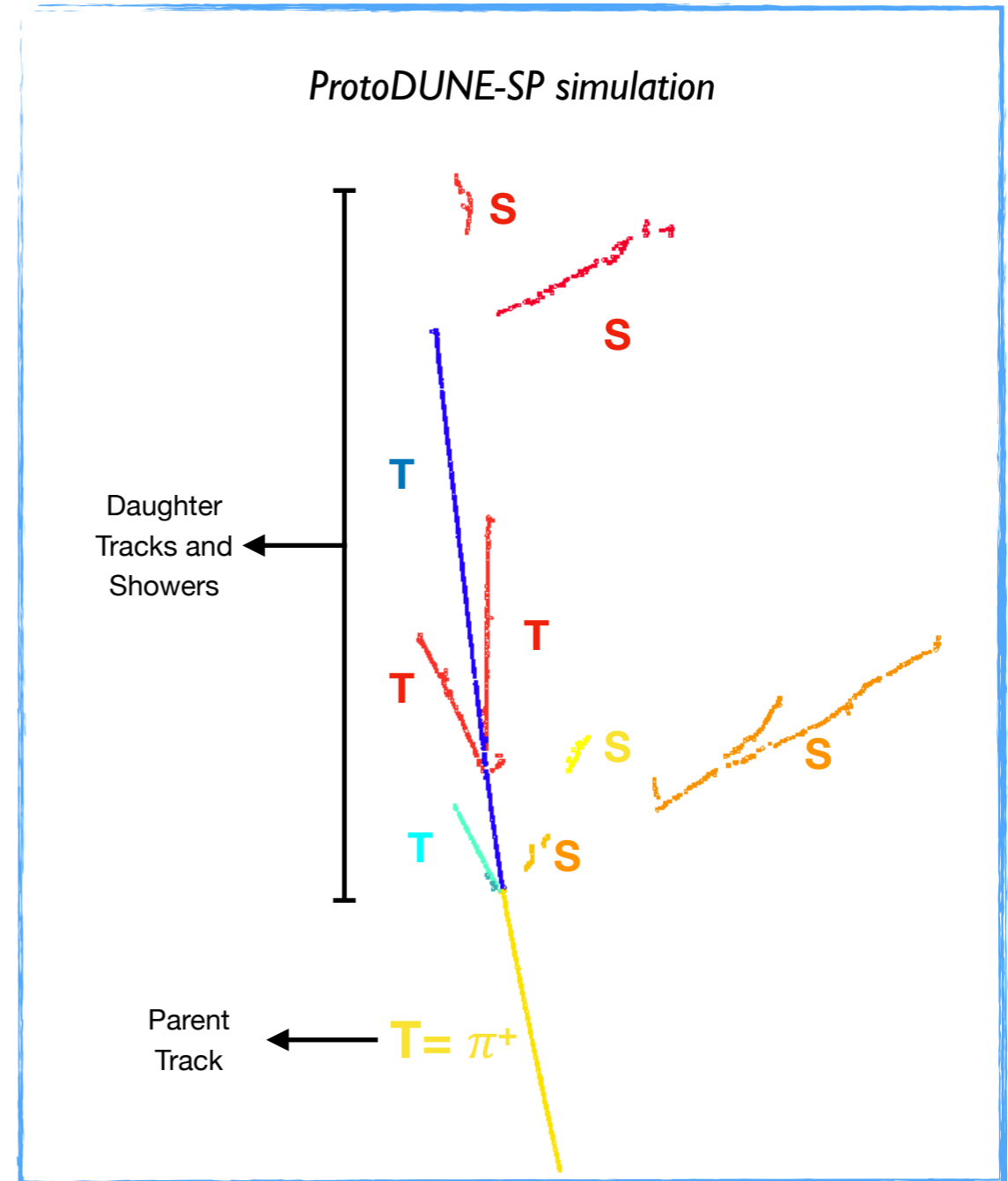
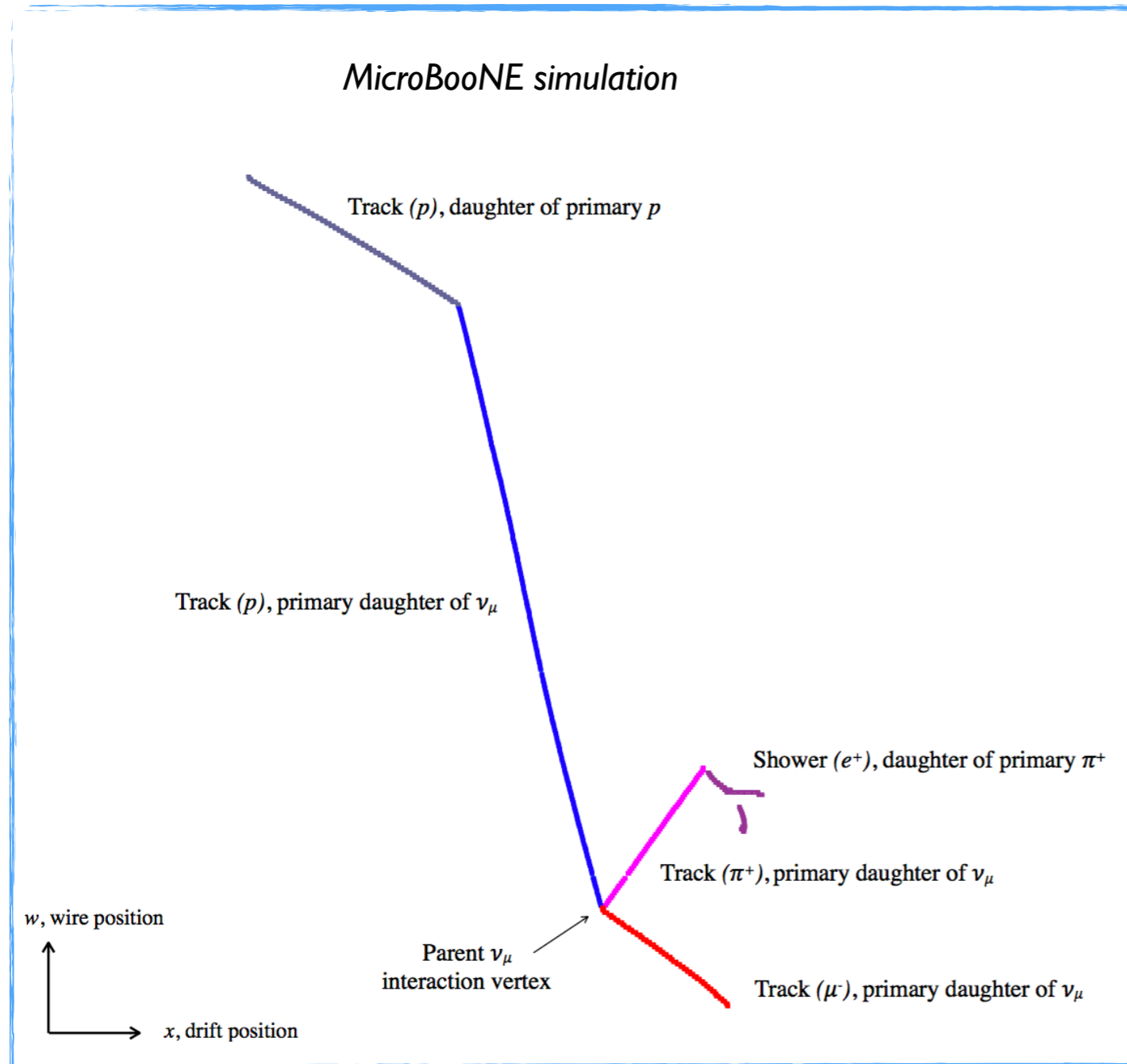
Series of algs deal with remnants to improve particle completeness (esp. sparse showers):

- Pick up small, unassociated clusters bounded by the 2D envelopes of shower-like particles.
- Use sliding linear fits to 3D shower clusters to define cones for merging small downstream shower particles, or picking up additional unassociated clusters.
- If anything left at end, dissolve clusters and assign hits to nearest shower particles in range.



Particle Hierarchy Reconstruction - 3D

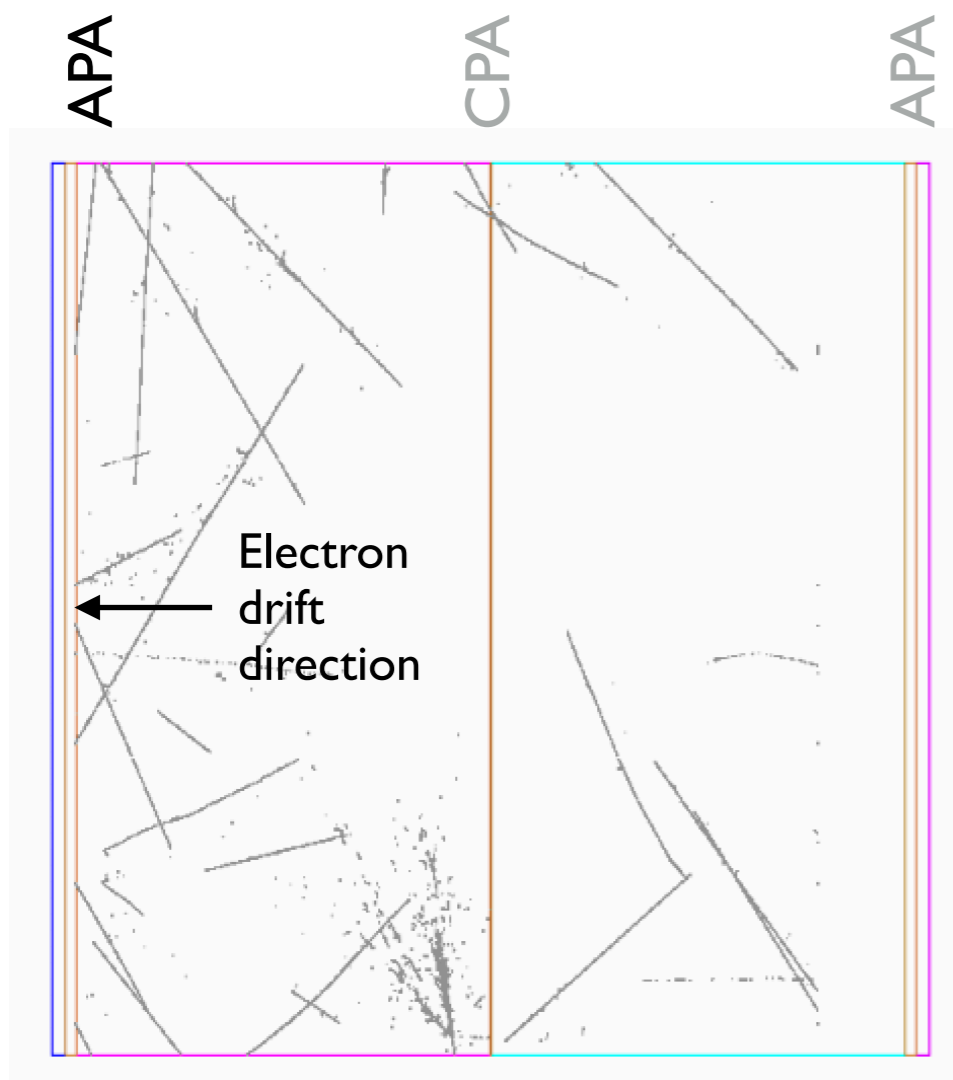
Use 3D clusters to organise particles into a hierarchy, working outwards from interaction vtx:



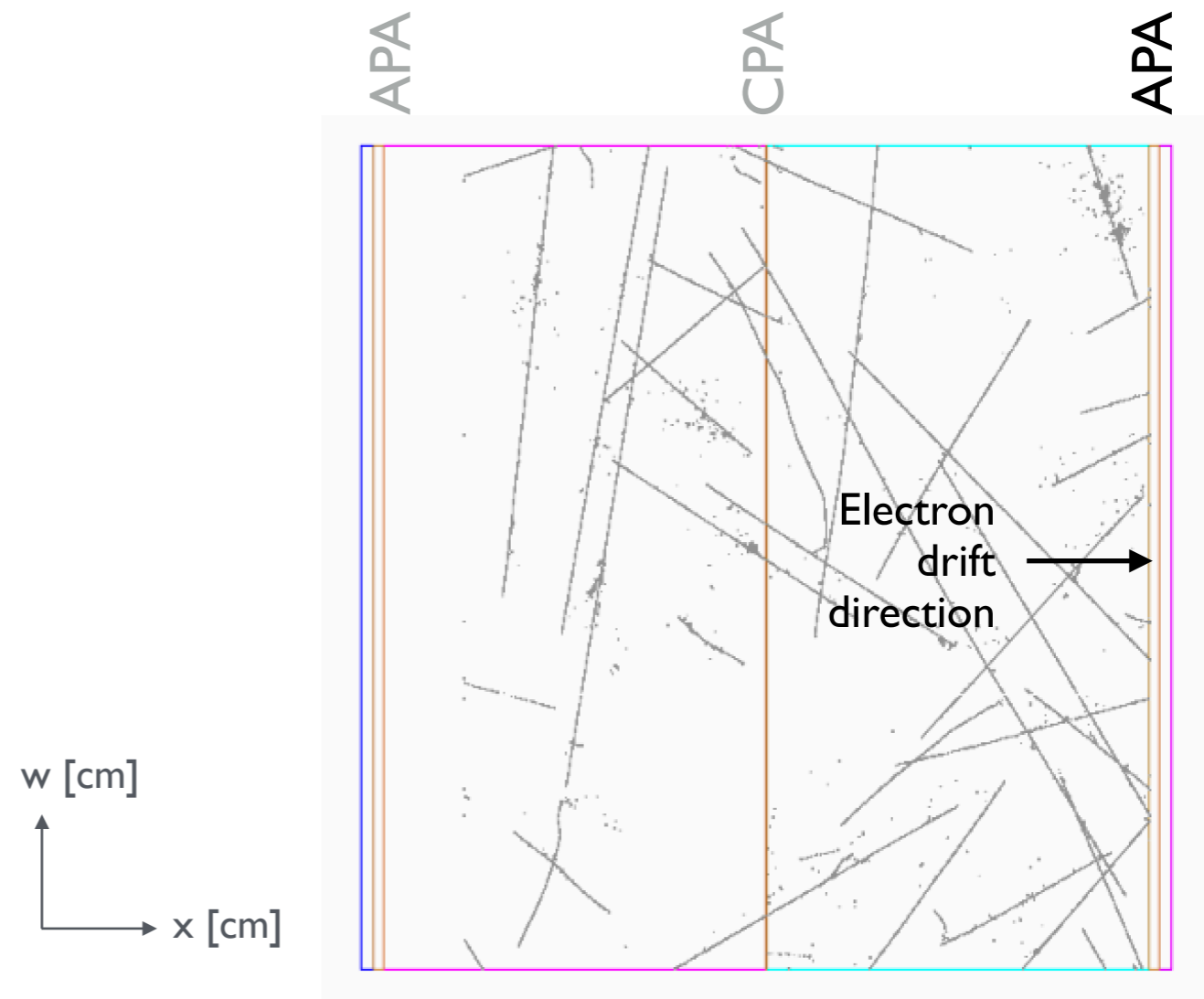
EPJC (2018) 78:82

Example: Reconstruction at ProtoDUNE-SP

- Single Phase DUNE Far Detector prototype, exposed to test beam at CERN
- Multiple “drift volumes”, complex topologies and significant cosmic-ray backgrounds:
 - An ideal testing ground for LArTPC pattern recognition



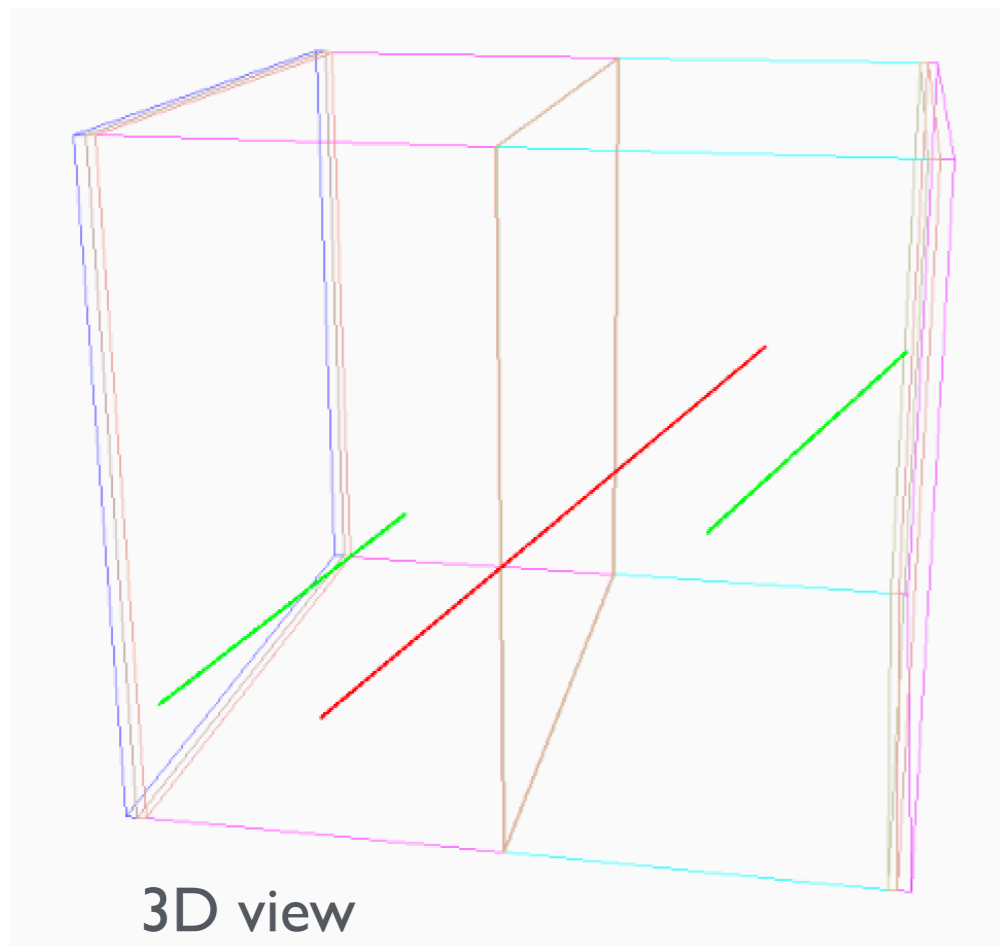
APA: Anode Plane Assembly
CPA: Cathode Plane Assembly



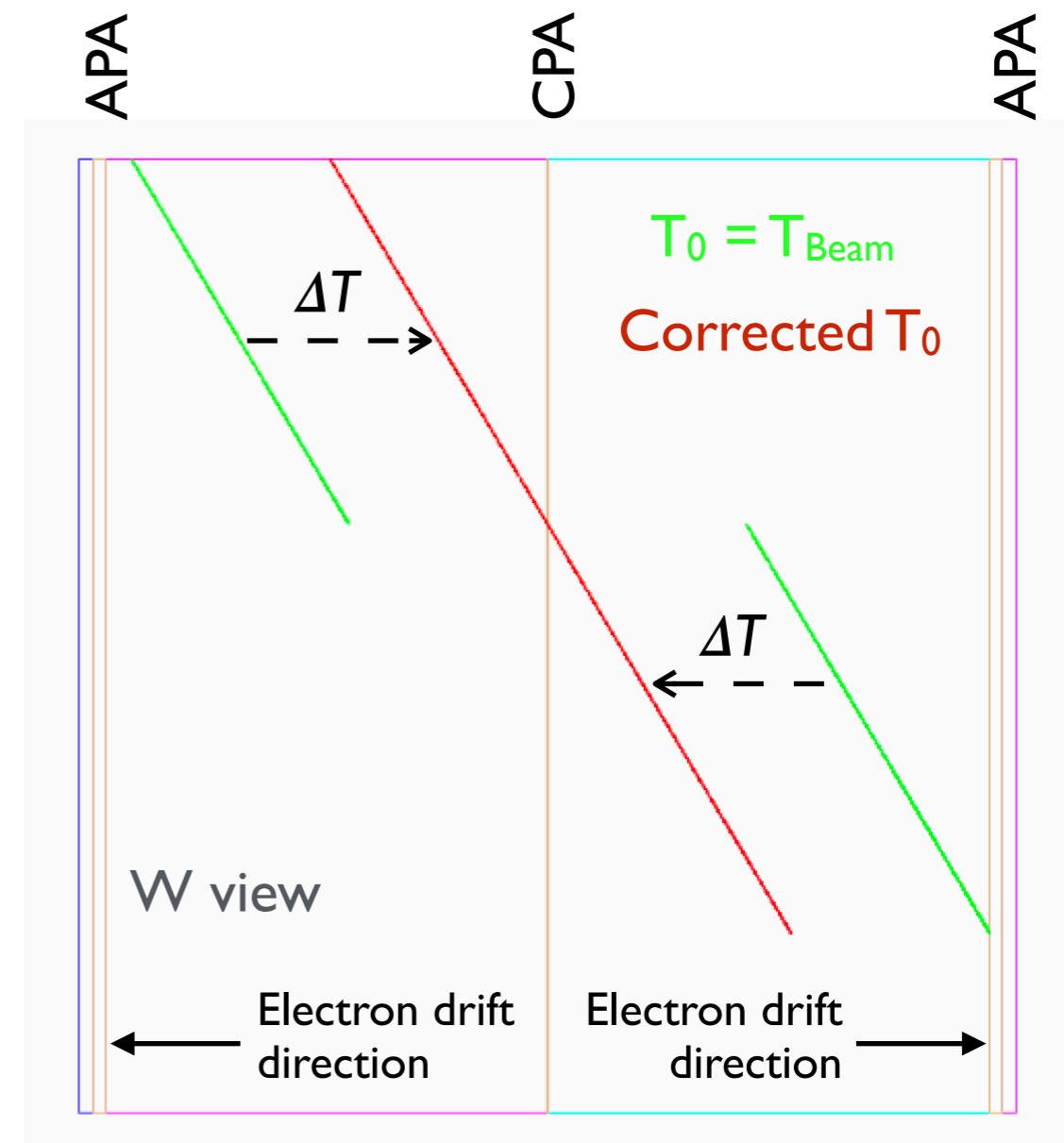
I. Reconstruct cosmic-ray muons independently for each volume of detector

Stitching and T_0 Identification

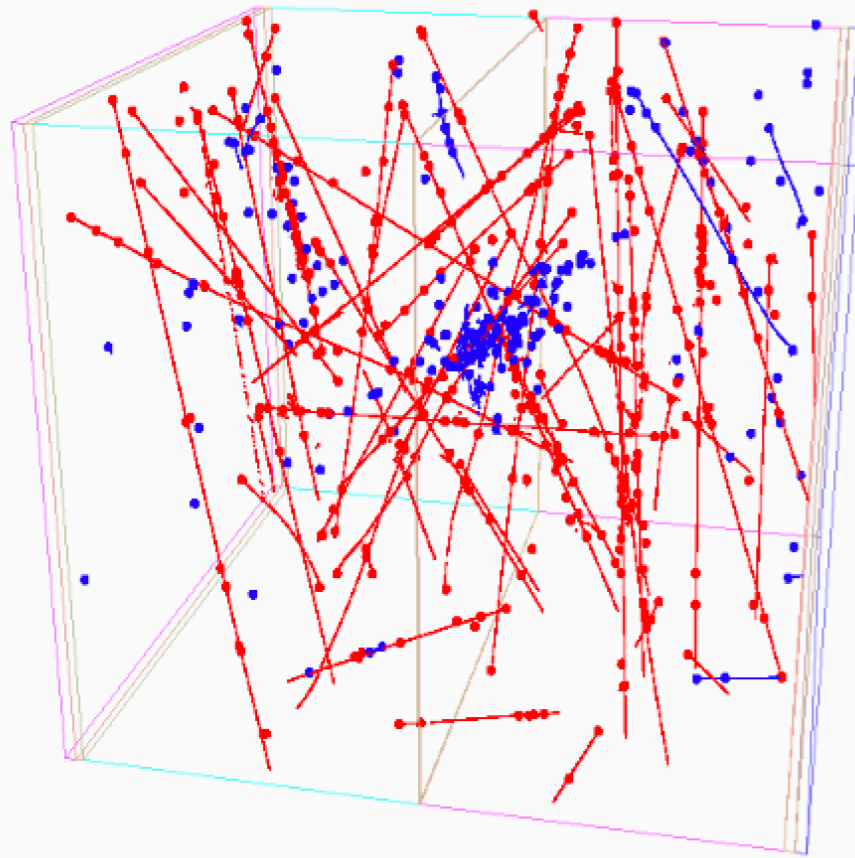
- In a LArTPC image, one coordinate derived from drift times of ionisation electrons:
 - But, only know electron arrival times, not actual drift times: need to know start time, T_0
 - For beam particles, can use time of beam spill to set T_0 , but unknown for cosmic rays
 - Place all hits assuming $T_0 = T_{\text{Beam}}$, but can identify T_0 for any cosmic rays crossing volumes



2. Stitch together any cosmic rays crossing between volumes, identifying T_0



Cosmic Ray Tagging and Slicing

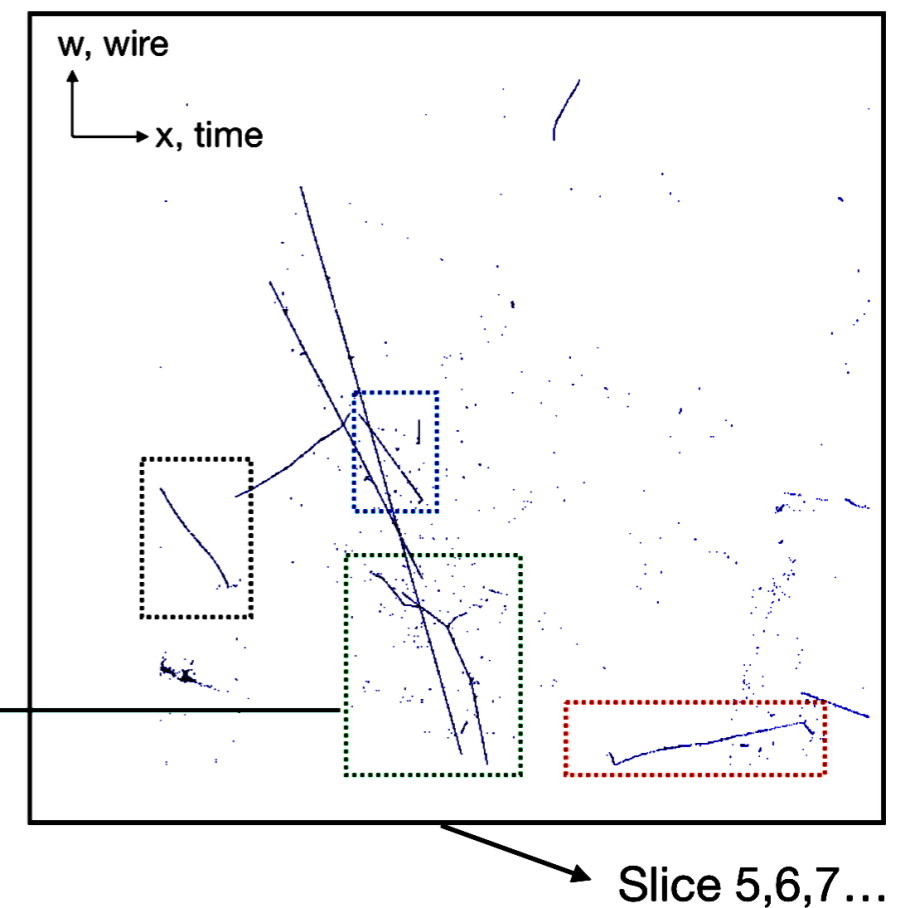
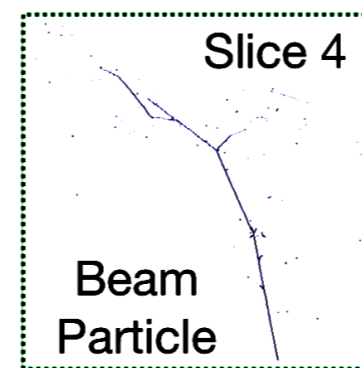
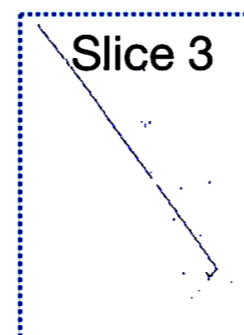
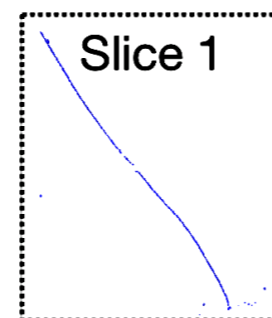


3. Identify clear cosmic rays (red) and hits to reexamine under test beam hypothesis (blue)

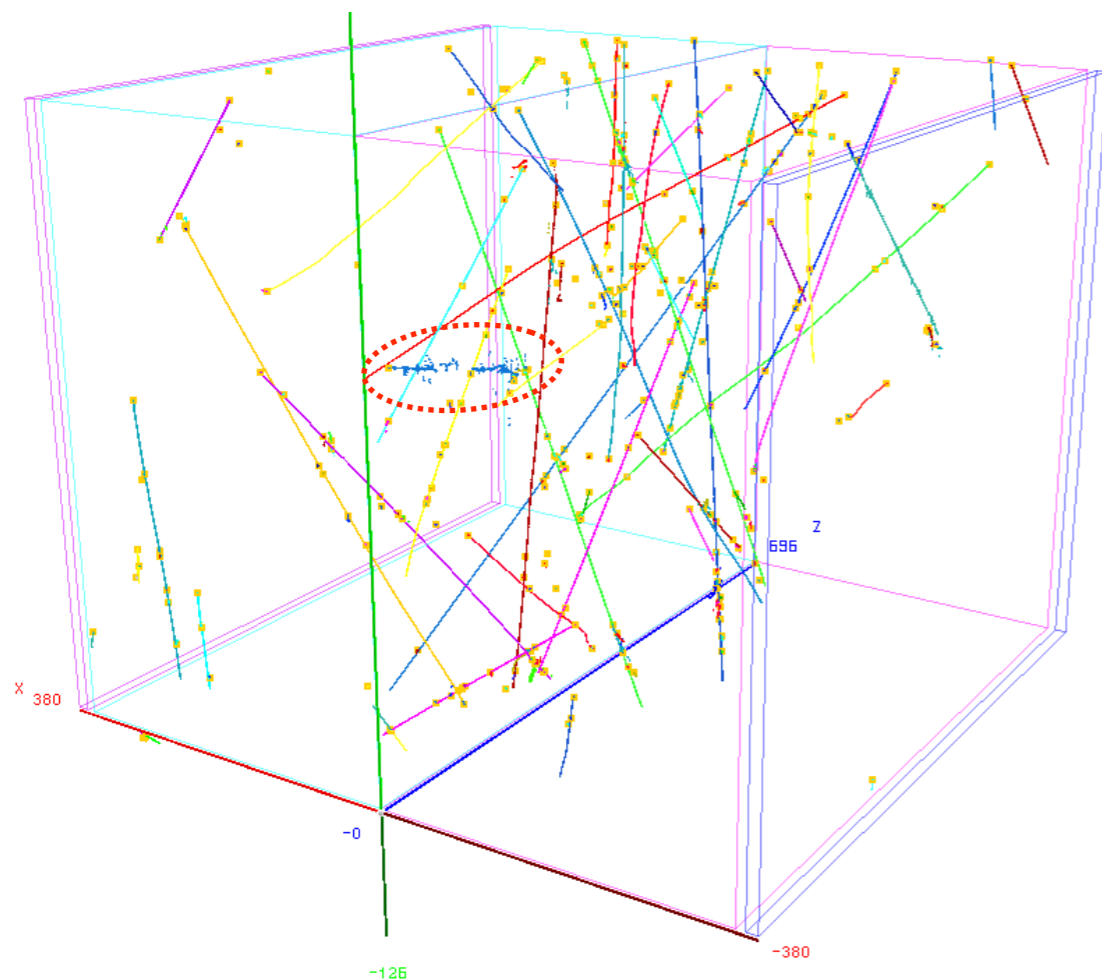
Clear cosmic rays:

- Particles appear to be “outside” of detector if $T_0 = T_{\text{Beam}}$
- Particles stitched between volumes using a $T_0 \neq T_{\text{Beam}}$
- Particles pass through the detector: “through going”

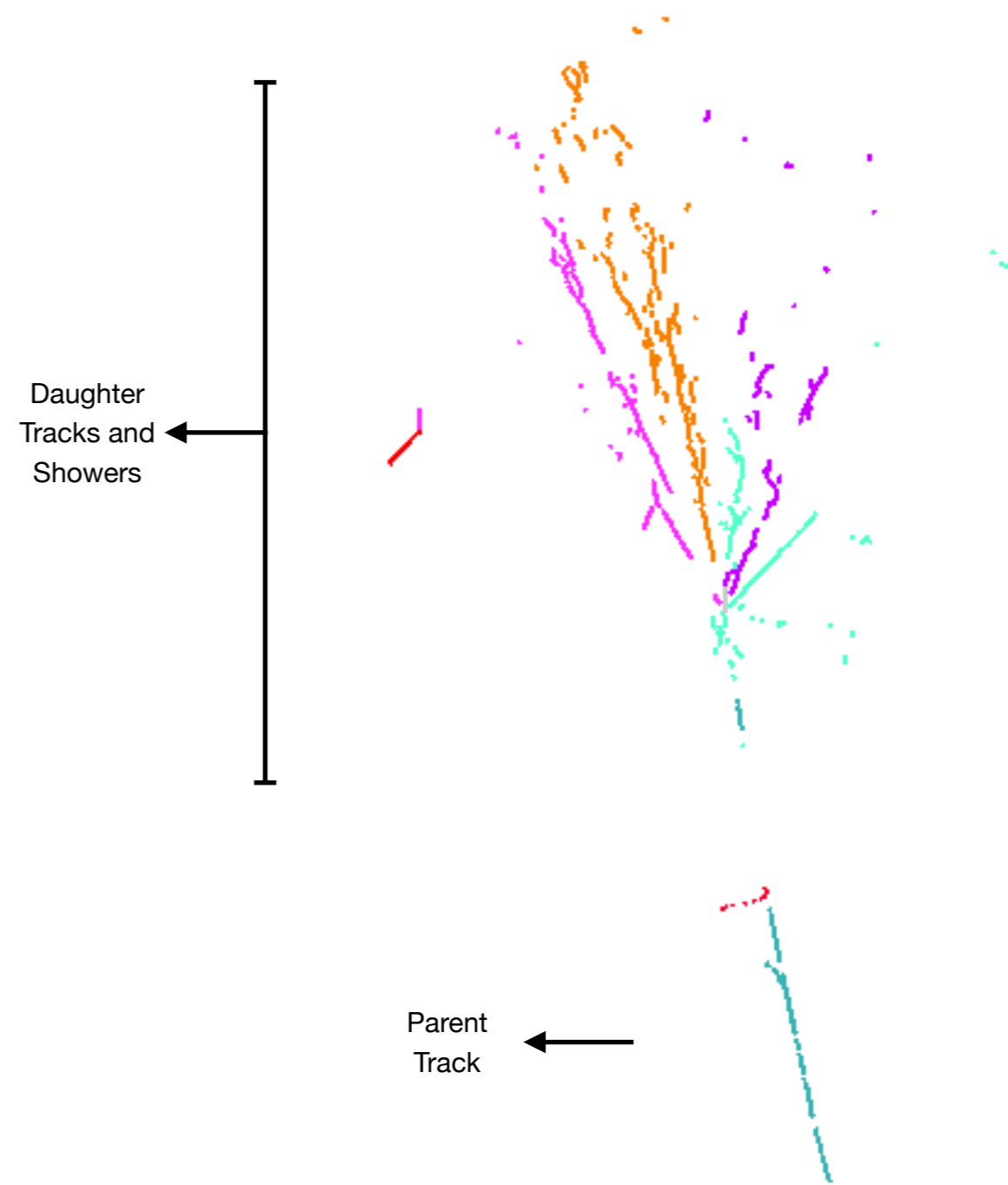
- Slice/divide blue hits from separate interactions
- Reconstruct each slice as test beam particle
- Then choose between cosmic ray or test beam outcome for each slice



Consolidated Output



*E.g. Reconstruction output: test beam particle (electron)
and: N reconstructed cosmic-ray muon hierarchies*

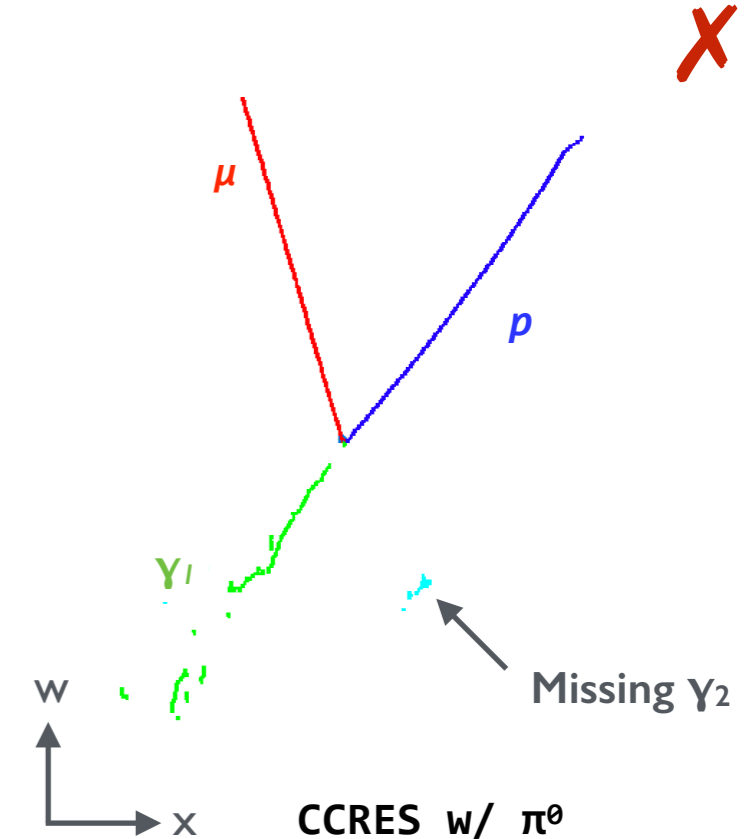
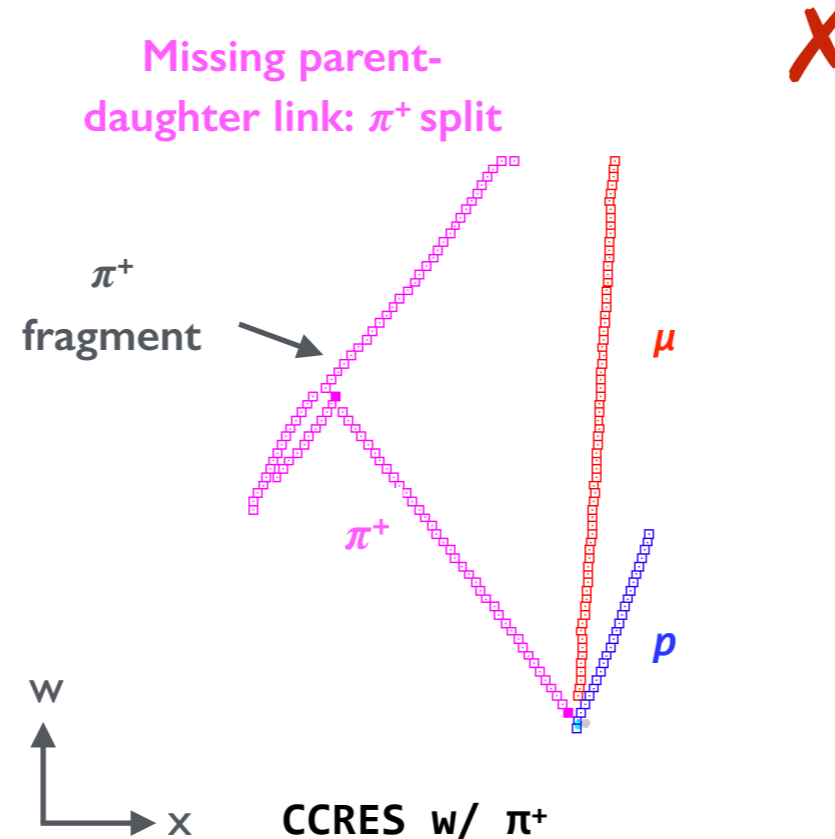
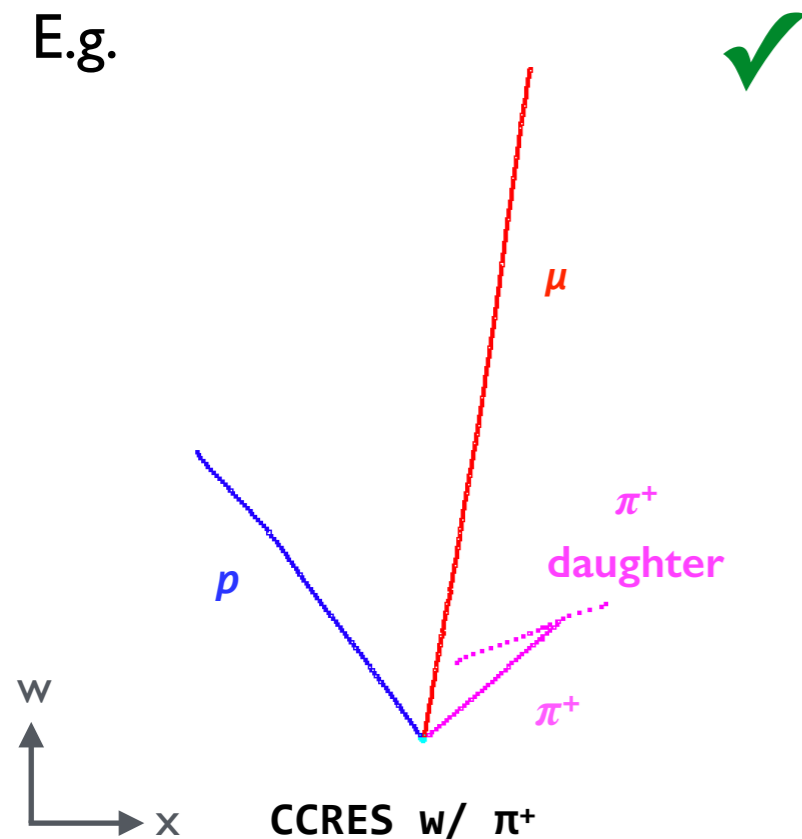


E.g. Test beam particle: charged pion

Assessing Pattern-Recognition Performance

- Assess performance for simulated events, using a selection of event topologies.
- Examine fraction of events deemed “correct” by *very strict* pattern-recognition metrics:
 - Consider exclusive final-states where all true particles pass simple quality cuts (e.g. nHits)
 - Correct means exactly one reco primary particle is matched to each true primary particle

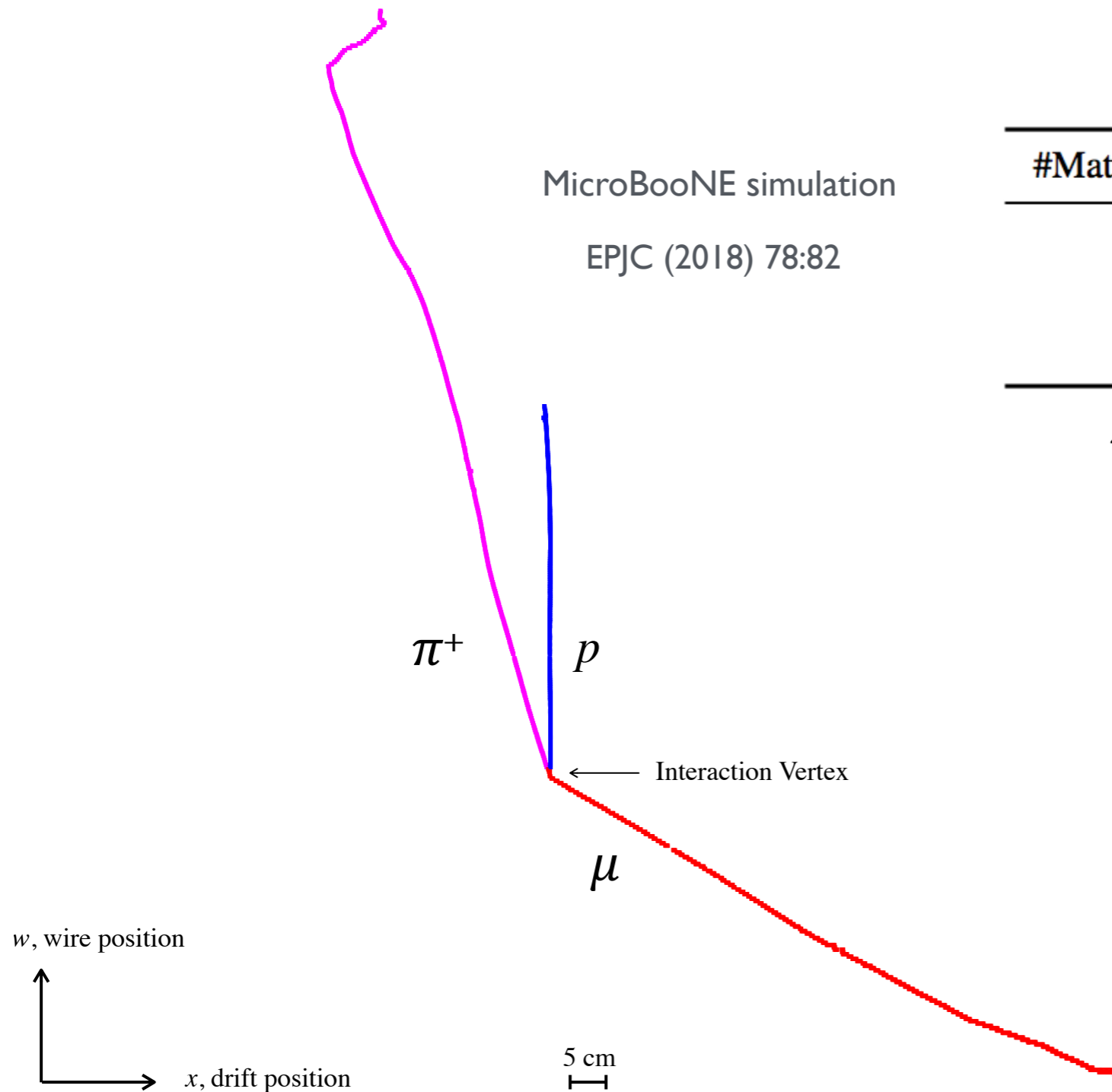
E.g.



Pandora at MicroBooNE



Three-track topology: CC ν_μ interactions with resonant charged pion production:



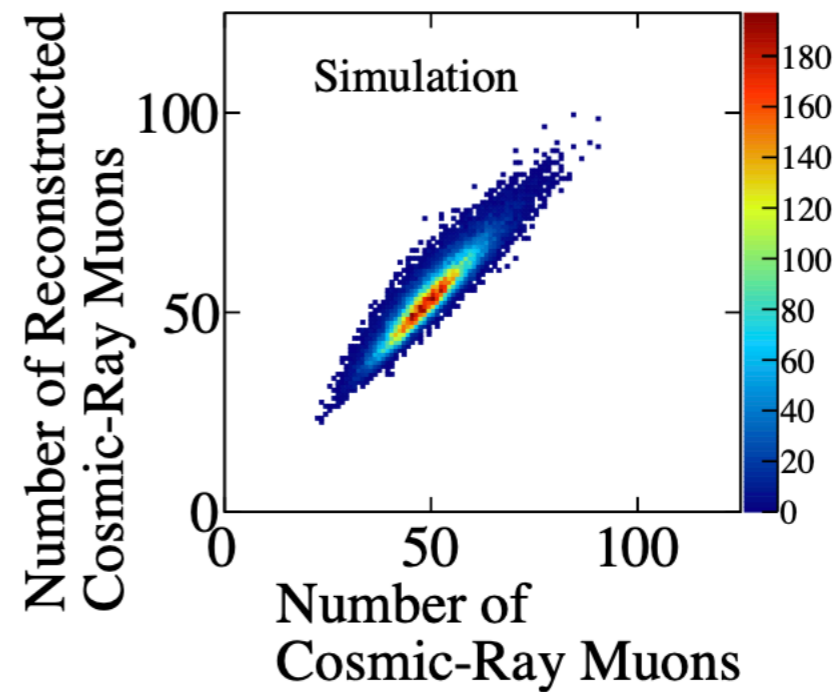
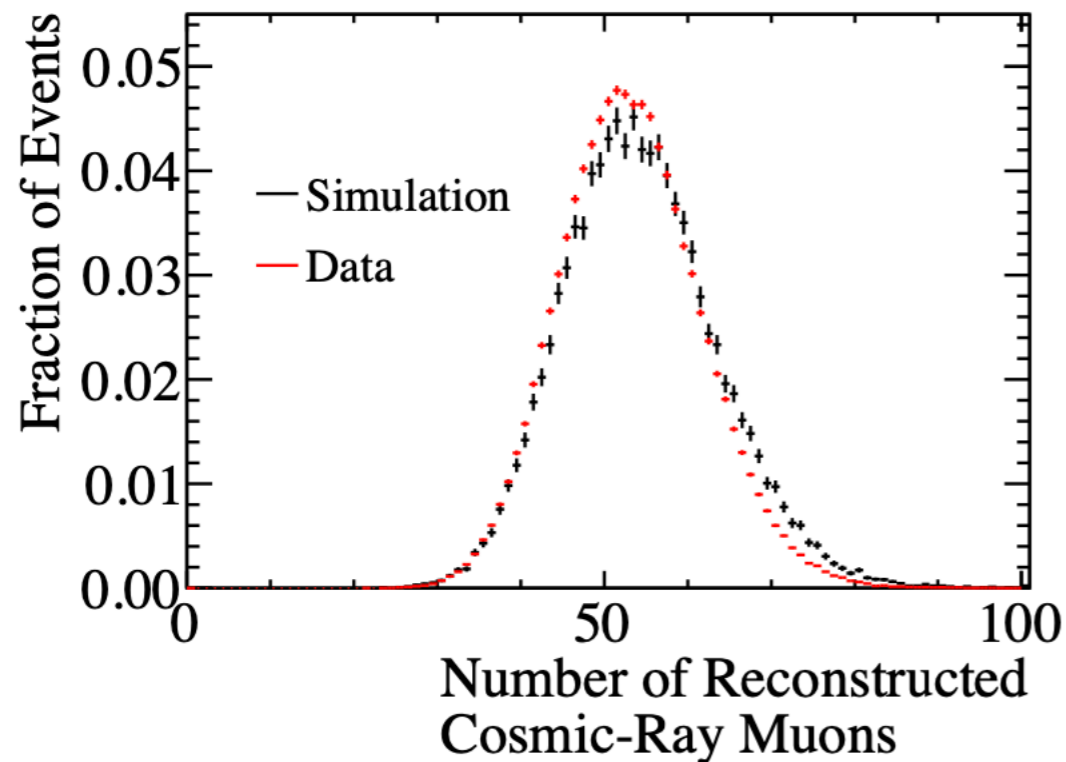
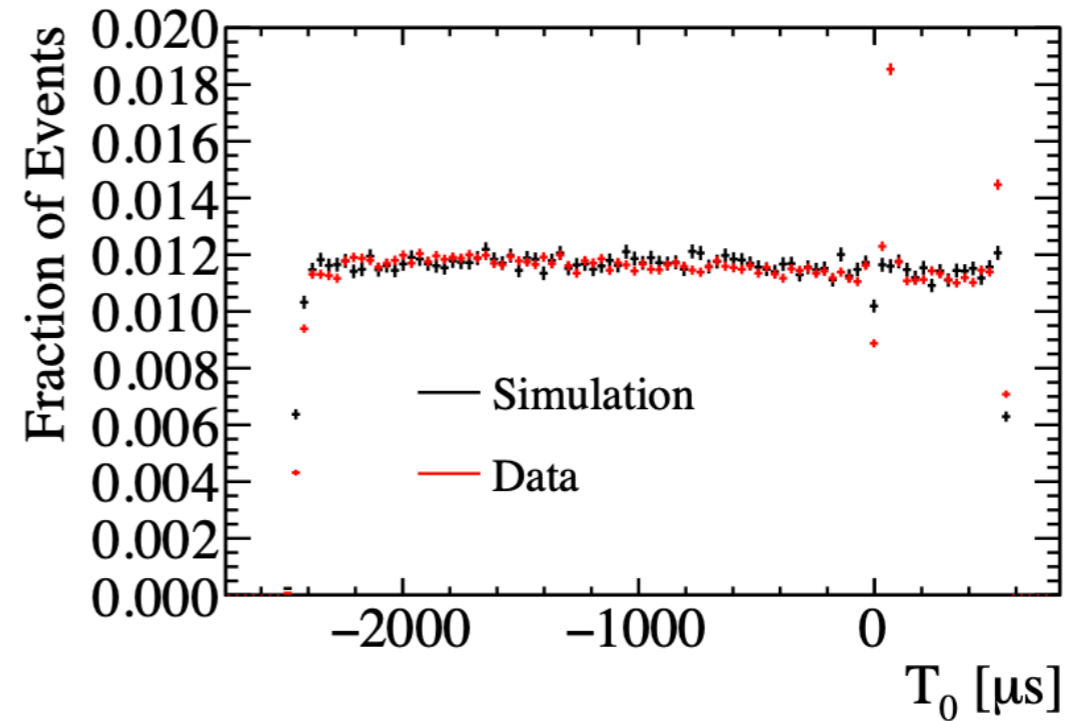
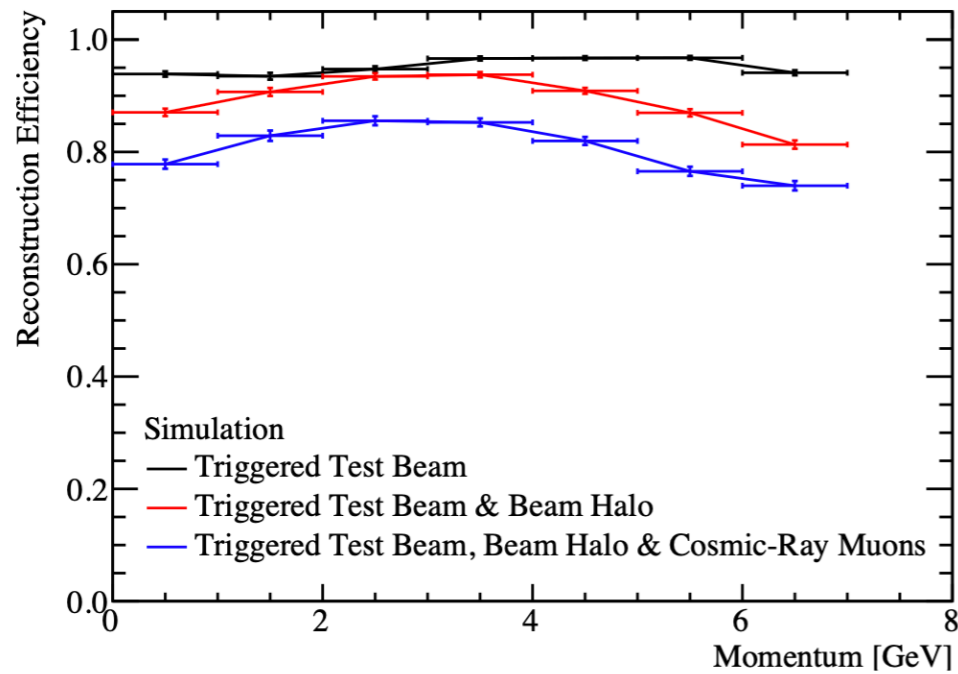
#Matched Particles	0	1	2	3+
μ	3.5%	95.1%	1.4%	0.0%
p	9.0%	86.8%	4.0%	0.3%
π^+	6.9%	80.9%	11.4%	0.8%

47,754 events, 70.5% have exactly one reco particle matched to each target.

- Performance for μ and p similar to that reported for quasi-elastic events.
- π^+ interactions can lead to hierarchy of visible particles. If reconstructed separately (without parent-daughter links), π^+ is reportedly split.

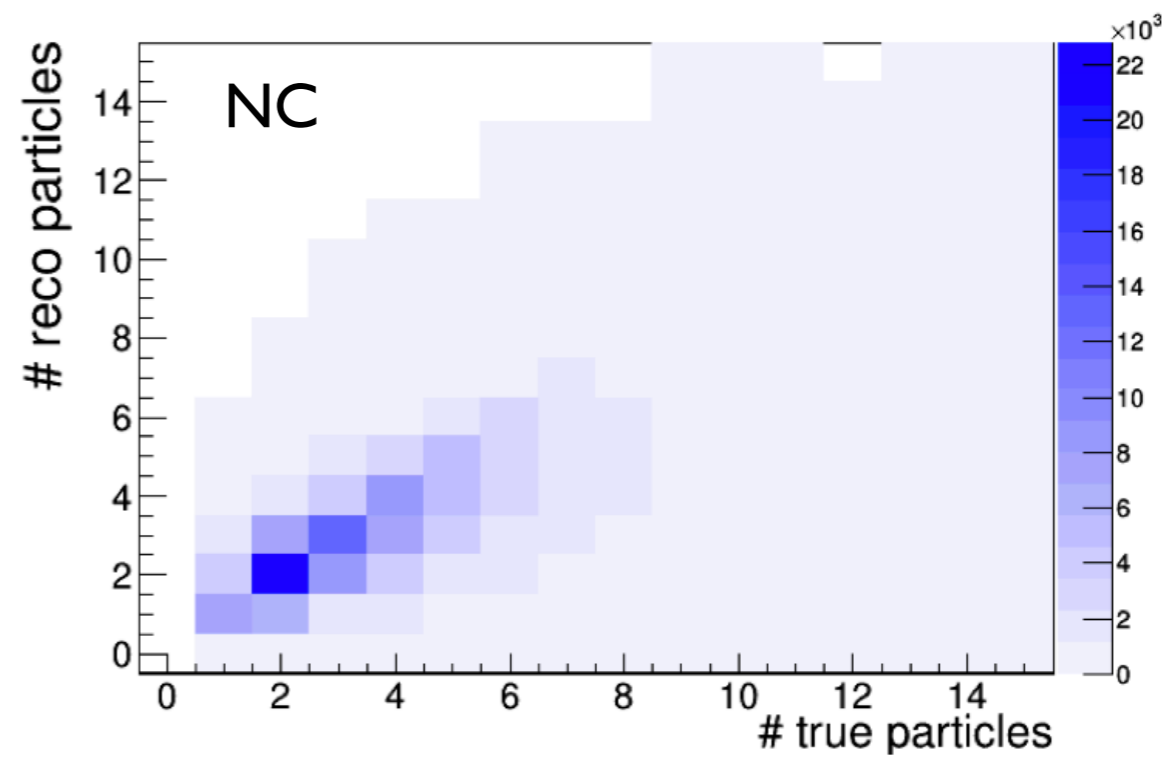
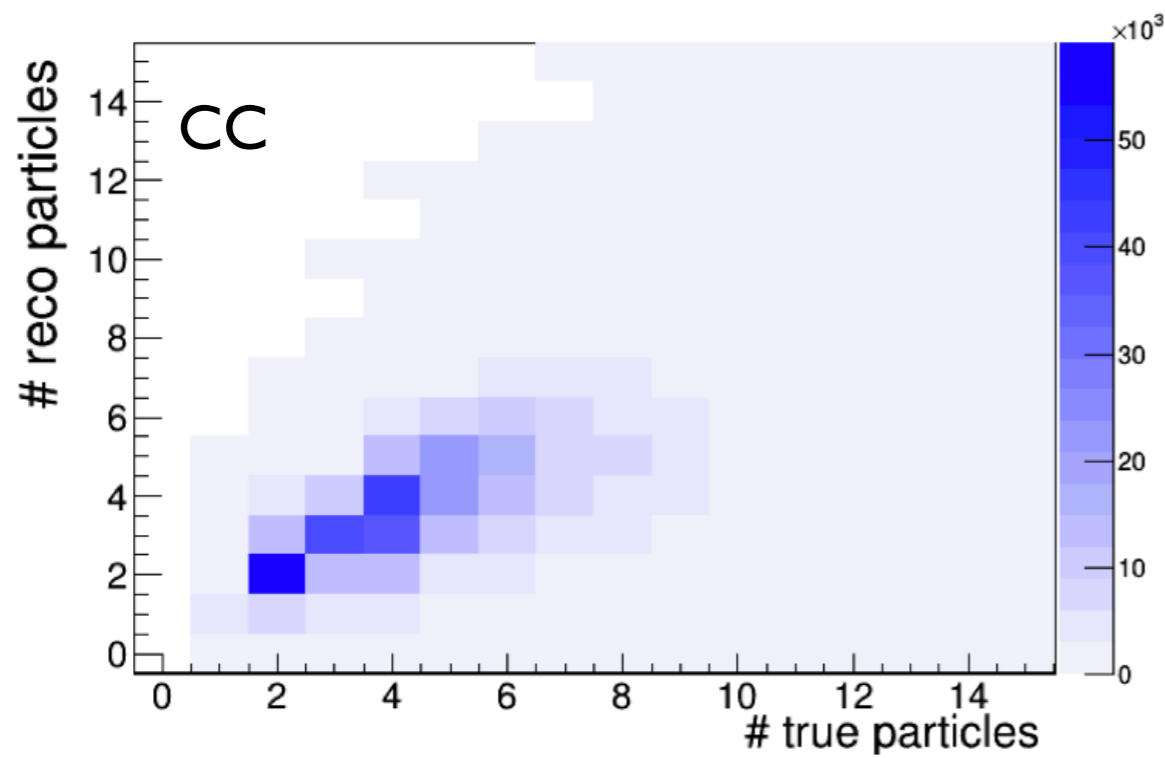
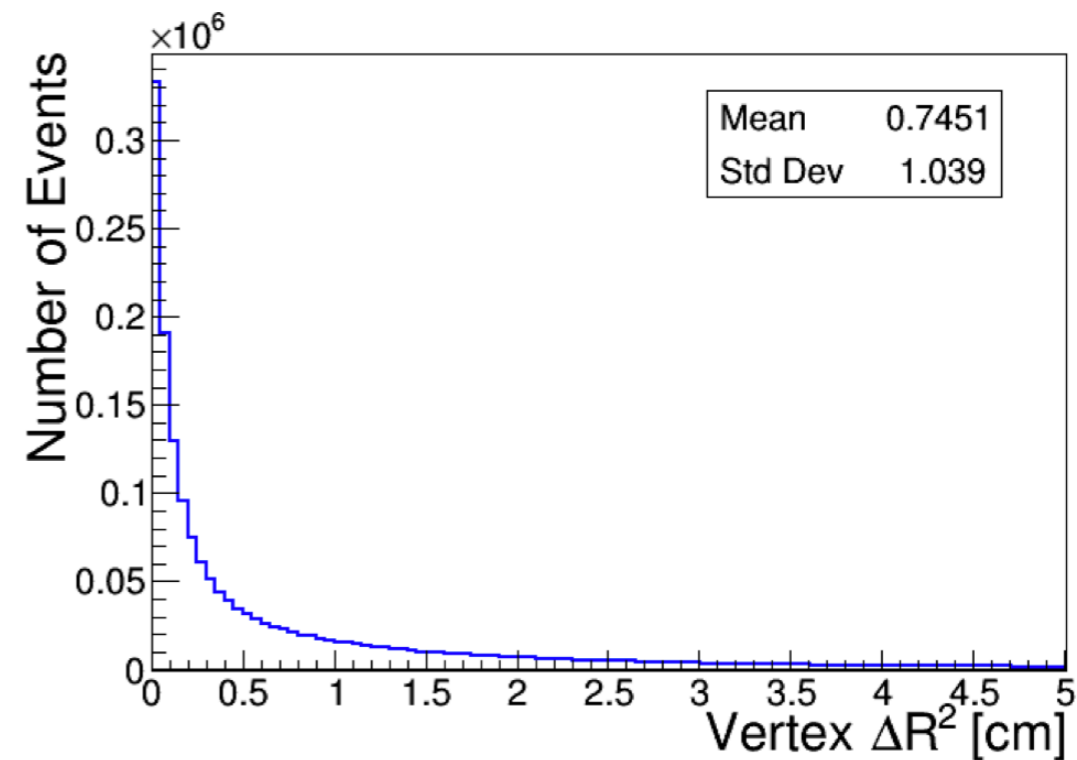
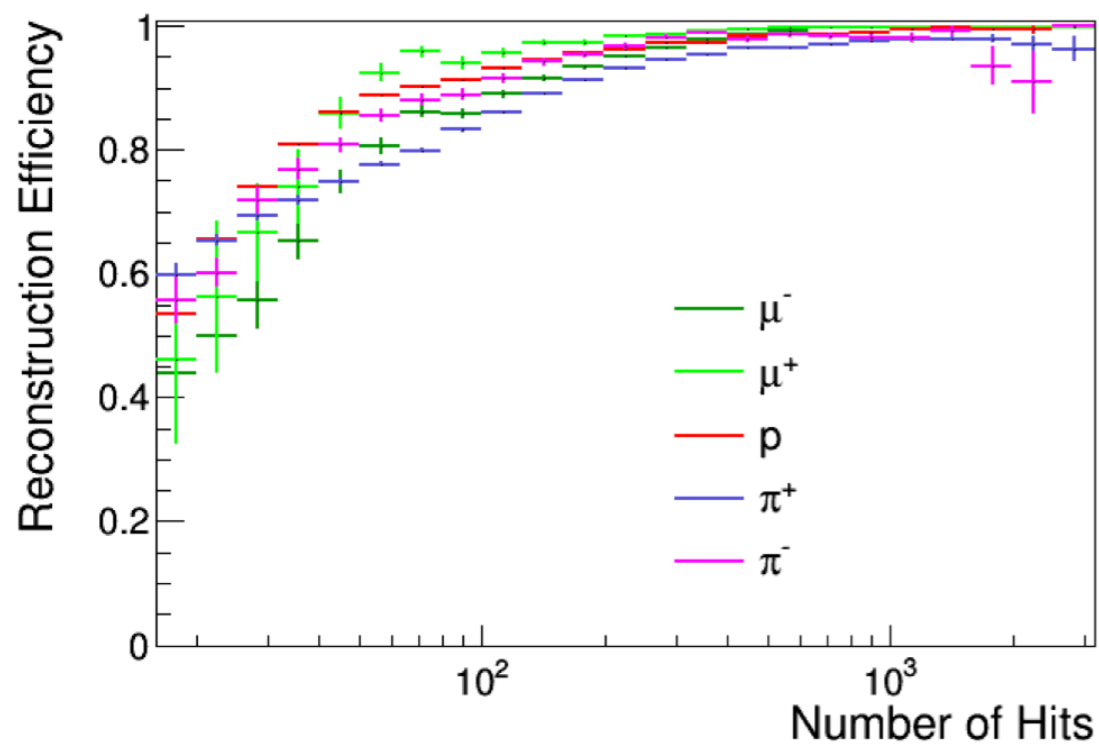
1.1 GeV ν_μ

Pandora for ProtoDUNE-SP



<https://arxiv.org/pdf/2002.03005.pdf>

Pandora for DUNE Far Detector

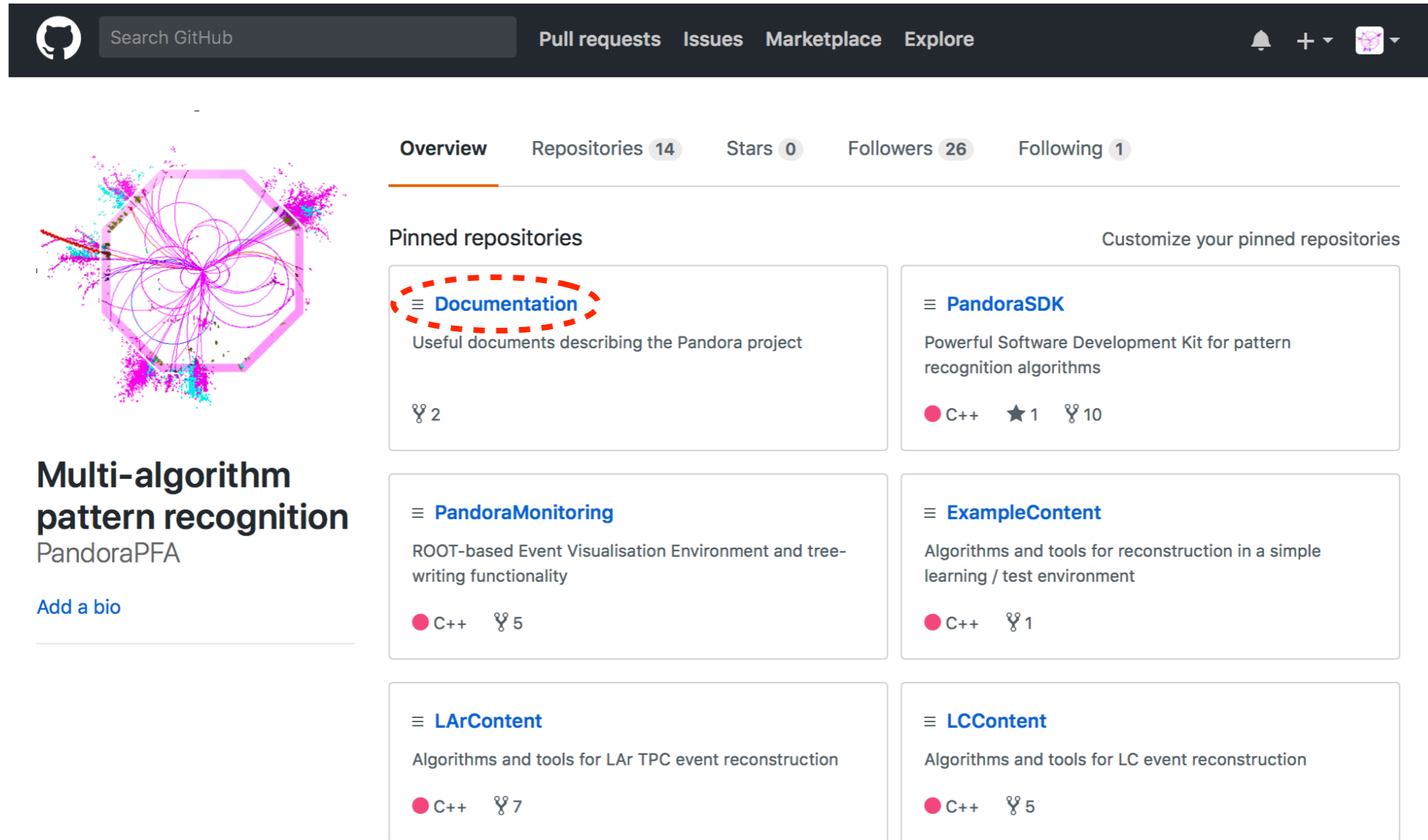


<https://arxiv.org/pdf/2002.03005.pdf>

Summary

- The use of Liquid Argon technology is one of the cornerstones of the current and future neutrino programmes.
- High-performance reconstruction techniques are required in order to fully exploit the imaging capabilities offered by LArTPCs:
 - Pandora multi-algorithm approach uses large numbers of decoupled algorithms to gradually build up a picture of events.
 - Output is a carefully-arranged hierarchy of reconstructed particles, each corresponding to a distinct track or shower.

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Multi-algorithm pattern recognition
PandoraPFA
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● C++ ★ 1 🍴 10
- PandoraMonitoring**
ROOT-based Event Visualisation Environment and tree-writing functionality
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- ExampleContent**
Algorithms and tools for reconstruction in a simple learning / test environment
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- LArContent**
Algorithms and tools for LAr TPC event reconstruction
● C++ 🍴 7
- LCContent**
Algorithms and tools for LC event reconstruction
● C++ 🍴 5



<https://github.com/PandoraPFA>



<https://pandorapfa.slack.com>





Additional slides

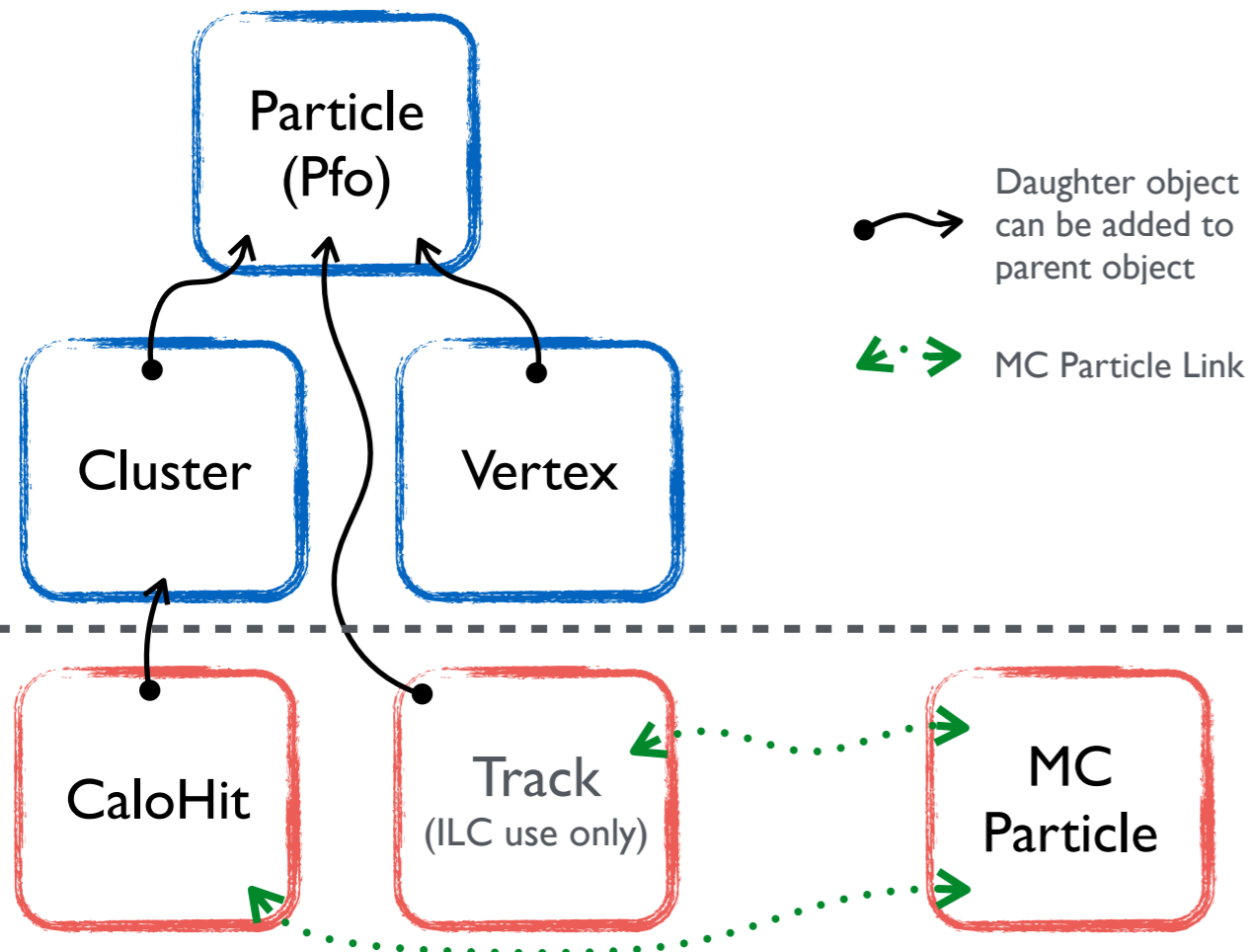
Pandora - Multi-Algorithm Approach

- Algorithms contain high-level logic and concentrate on the important bits: physics/patrec ideas.
 - Pandora software framework provides functions to access objects, make new objects, modify existing objects, etc.

Algorithm 1 Cluster creation pseudocode. The logic determining when to create new Clusters and when to extend existing Clusters will vary between algorithms.

```
1: procedure CLUSTER CREATION
2:   Create temporary Cluster list
3:   Get current CaloHit list
4:   for all CaloHits do
5:     if CaloHit available then
6:       for all newly-created Clusters do
7:         Find best host Cluster
8:         if Suitable host Cluster found then
9:           Add CaloHit to host Cluster
10:        else
11:          Add CaloHit to a new Cluster
12:        Save new Clusters in a named list
```

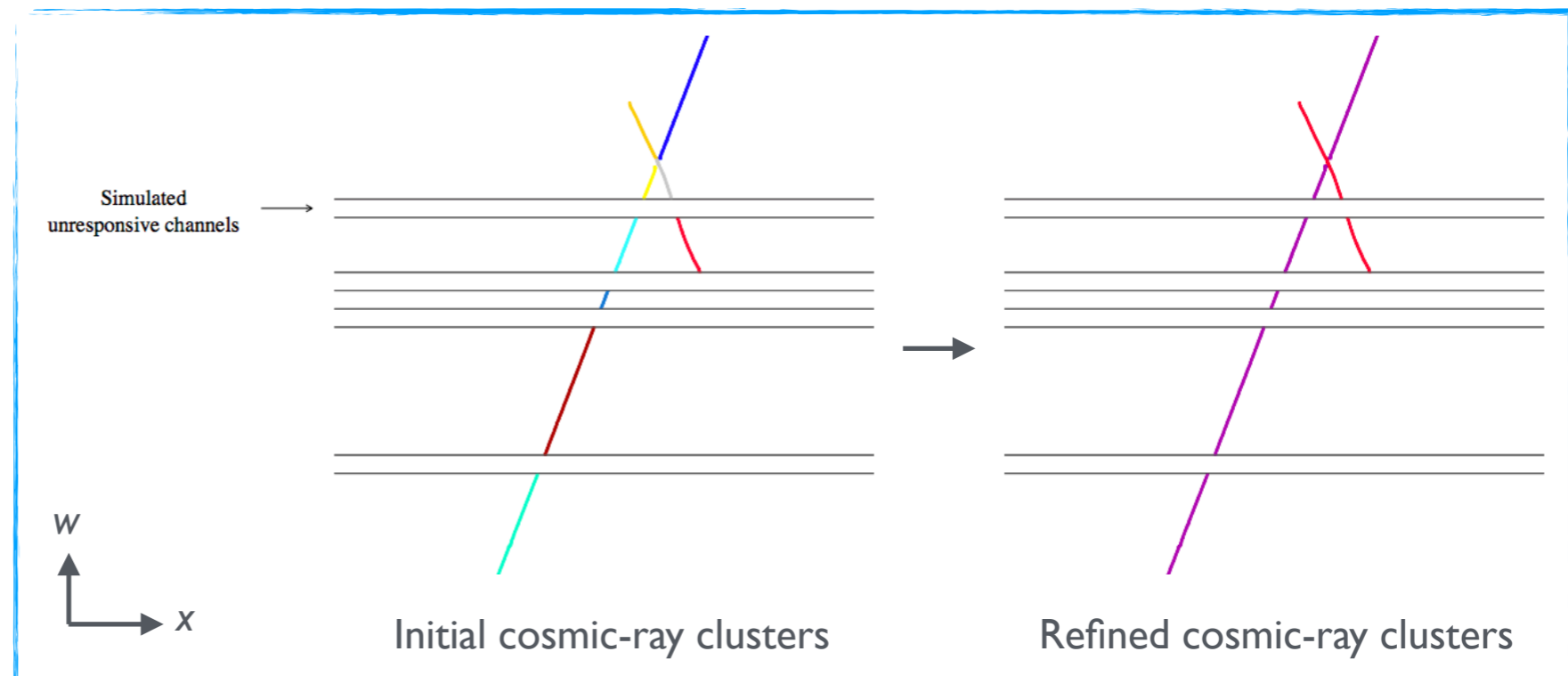
Example alg structure



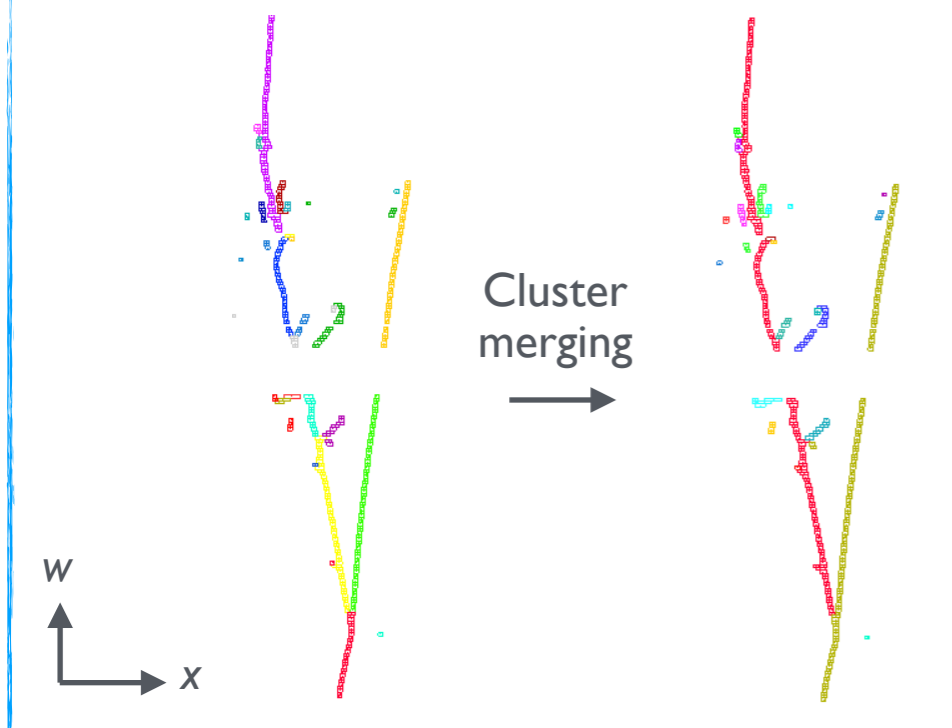
Event Data Model

Pandora - "Traditional" Approaches

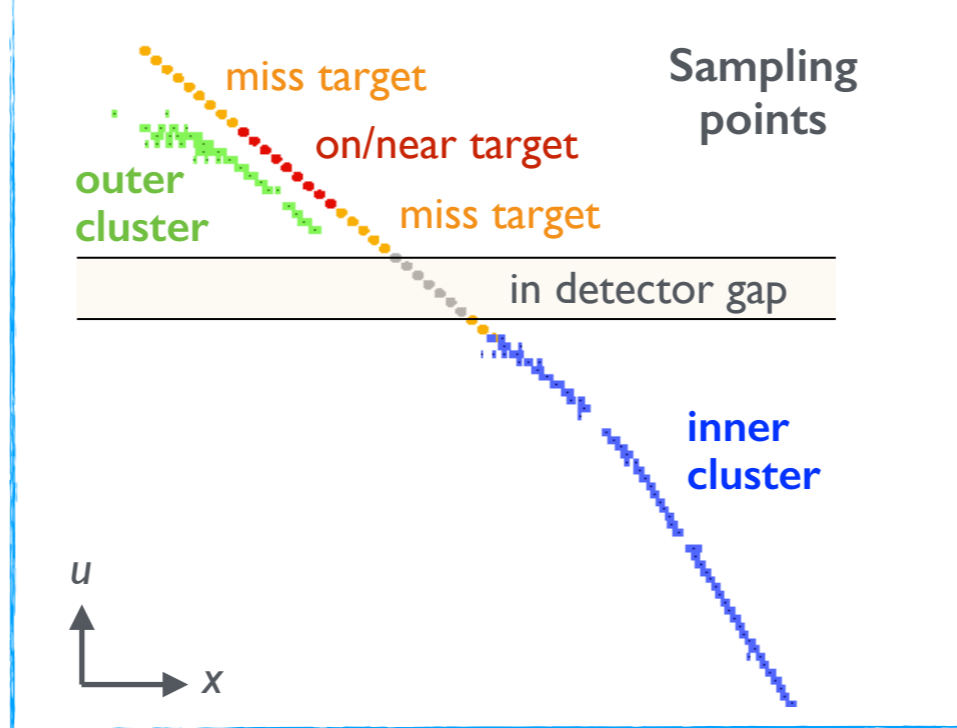
- For each wire plane, create a list of 2D clusters that represent continuous, unambiguous lines of hits:
 - Separate clusters for each structure, with clusters starting/stopping at any branch or ambiguity.



Longitudinal Association Algorithm



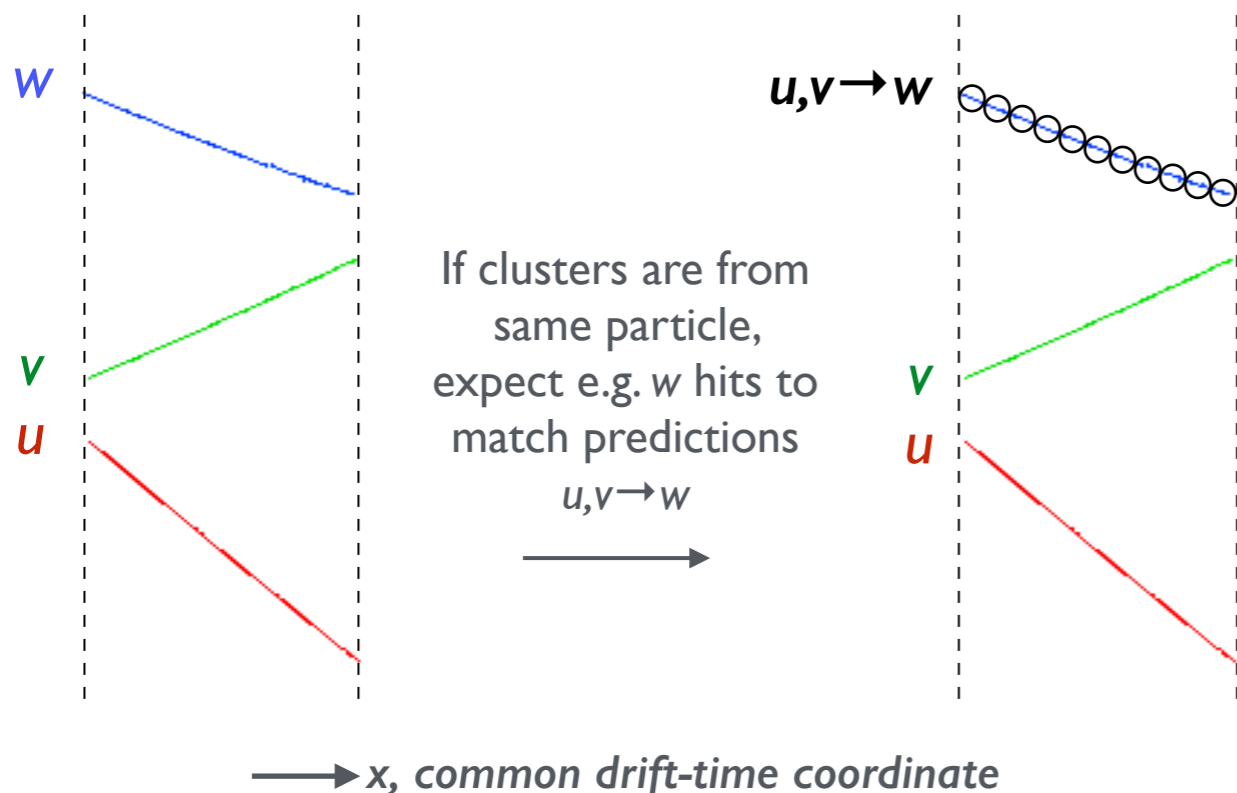
Cross Gaps Association Algorithm



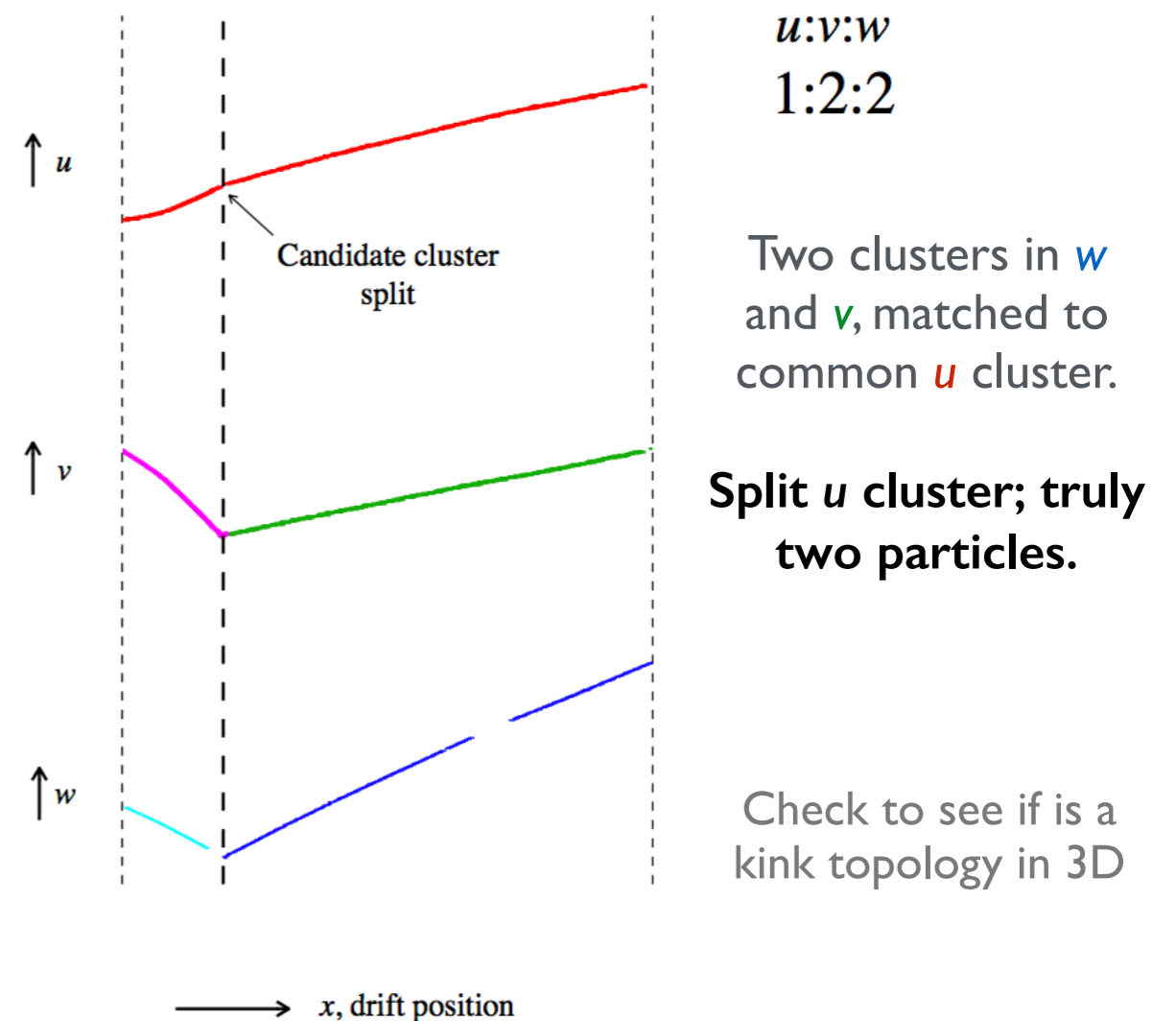
Initial clusters are refined by a series of **cluster-merging** and **cluster-splitting** algorithms that use **topological info.**

Pandora - “Detector-Physics” Approaches

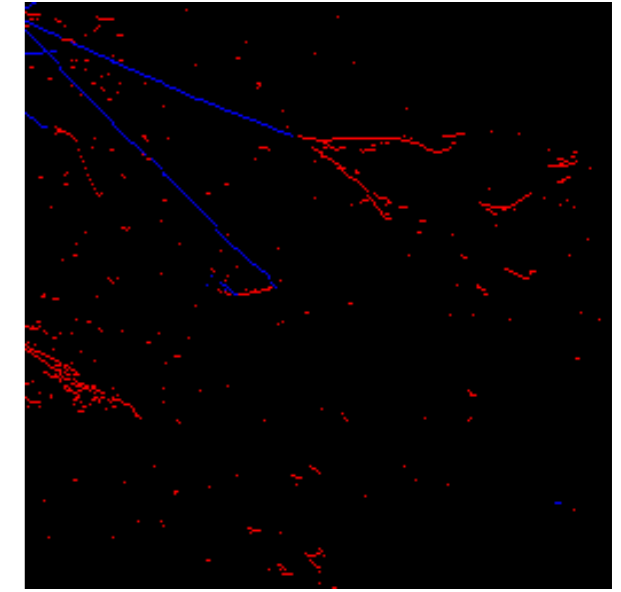
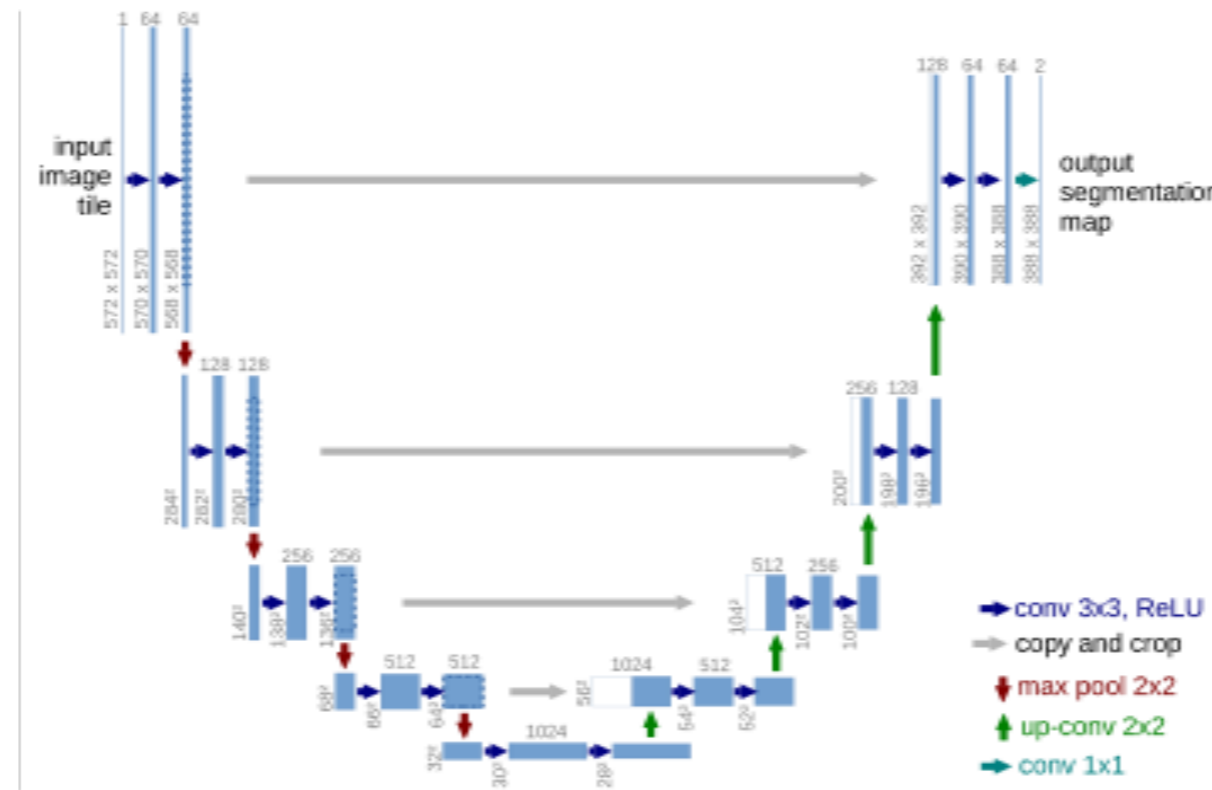
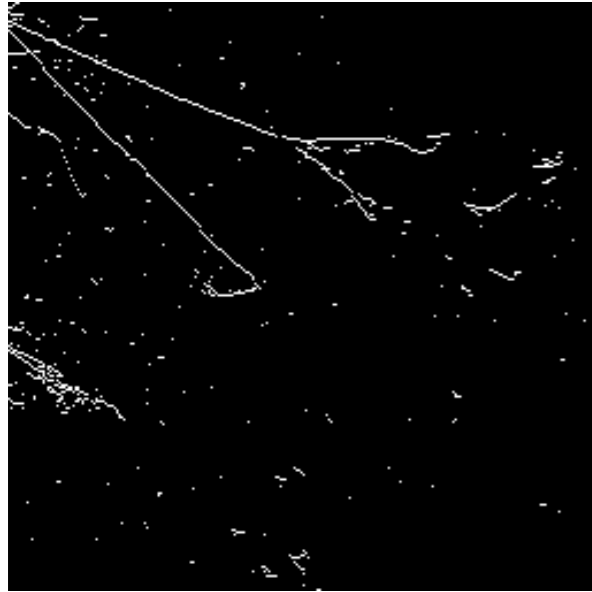
- Our original input was 3x2D images of charged particles in the detector.
- Should now have reconstructed three separate 2D clusters for each particle:
 - Compare 2D clusters from u, v, w planes to find the clusters representing same particle.
 - Exploit common drift-time coordinate and our understanding of wire plane geometry.



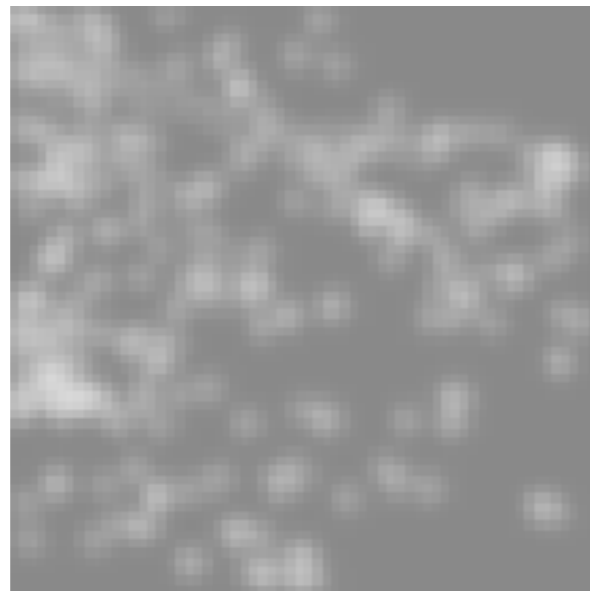
- Approach really comes to life when the 2D clustering “disagrees” between wire planes:
 - Automated detection of 2D PatRec issues, with treatment for specific cases, e.g.:



Pandora - "Deep-Learning" Approaches



<https://arxiv.org/abs/1505.04597>



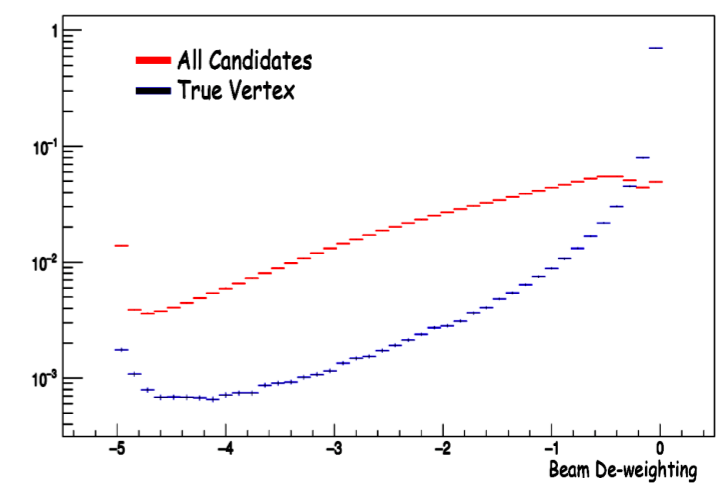
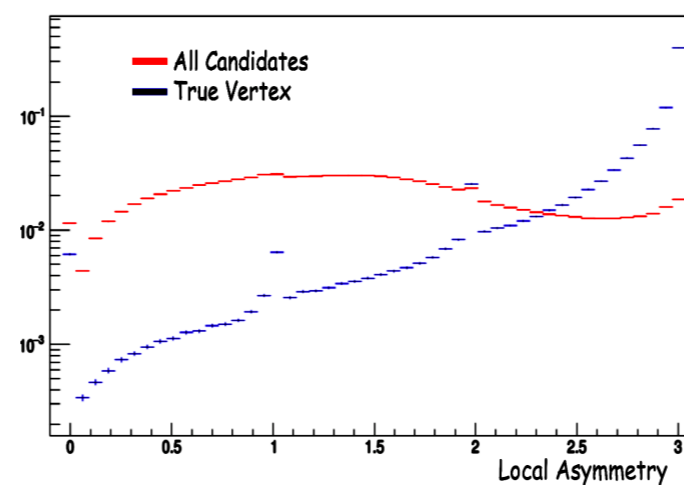
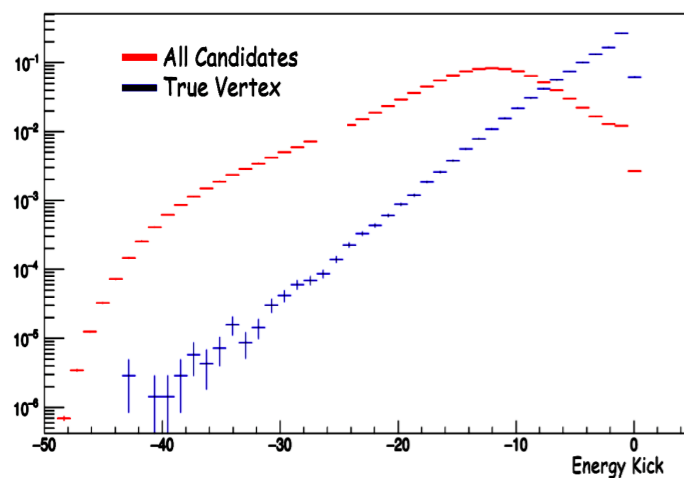
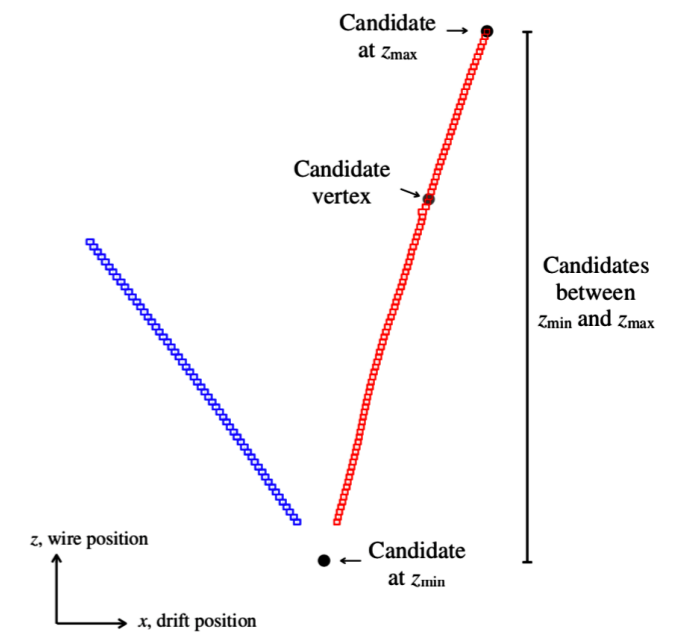
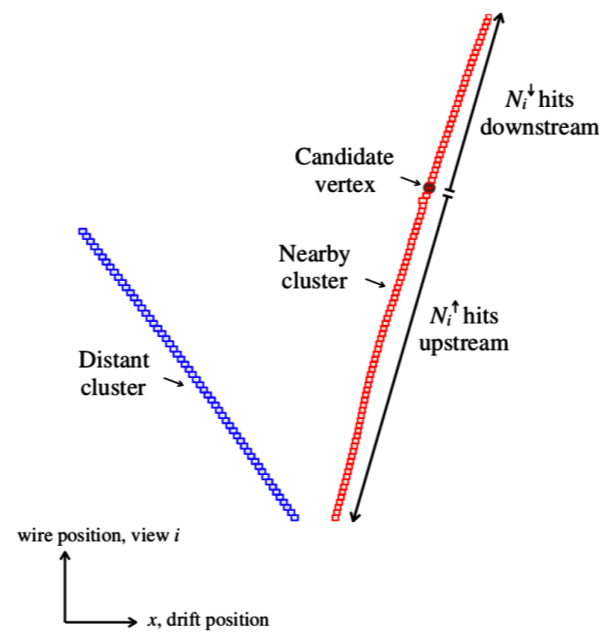
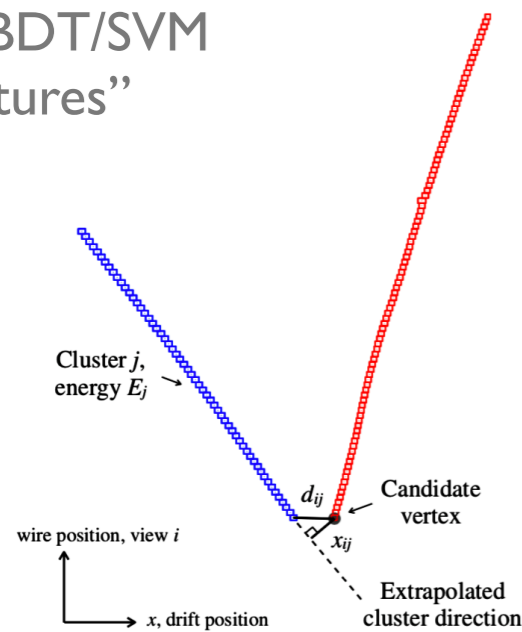
- Convolution filters identify features
- Track/Shower probability map constructed

Vertex Reconstruction - 3D

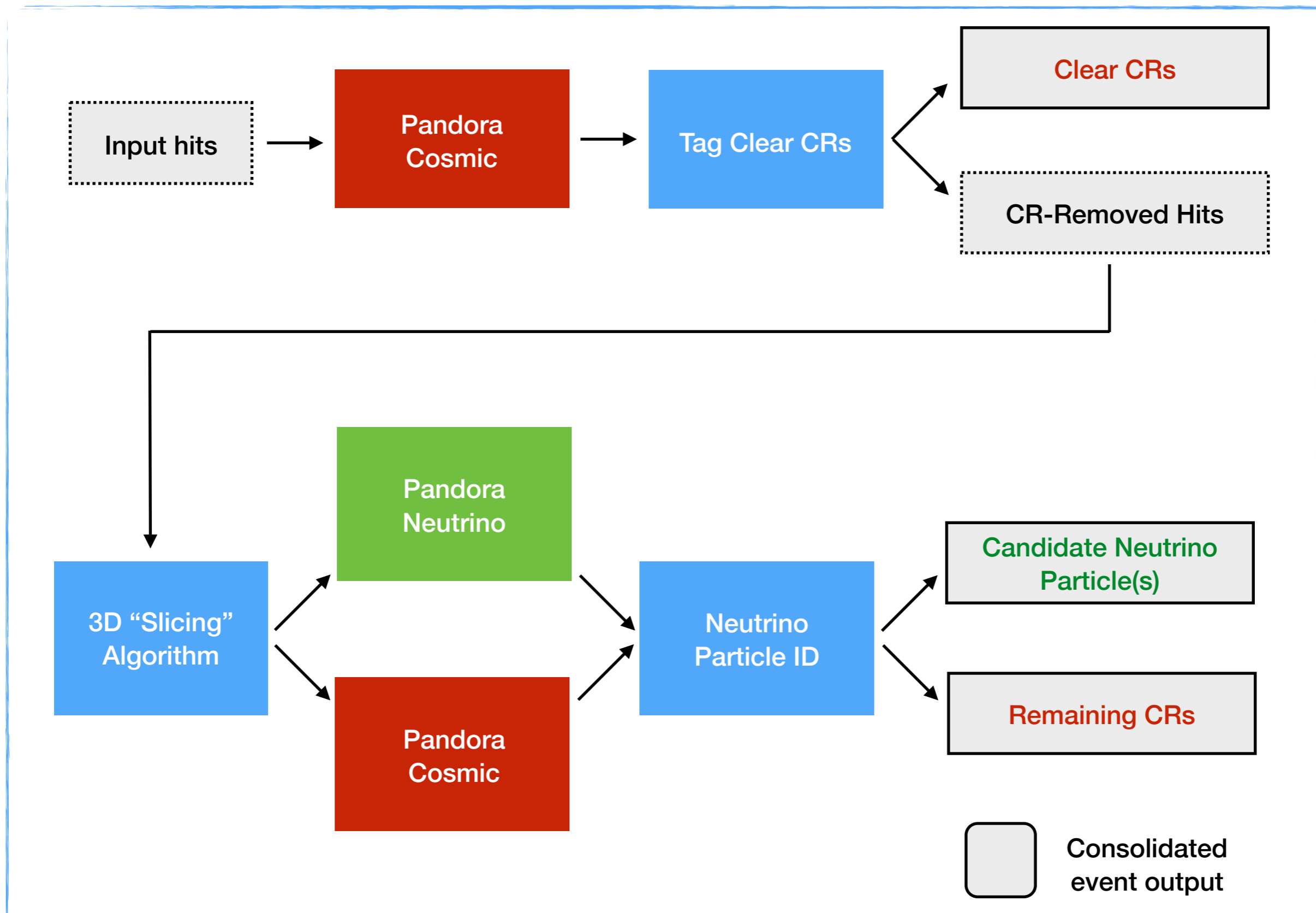
Interaction vertex is an important feature point in our LArTPC images:

Continued development, ever-more sophisticated approaches to finding 3D vertex position
Boosted Decision Trees (BDTs) or Support Vector Machines (SVMs) to select best candidate
Exploit Convolutional Neural Networks (CNNs)

E.g BDT/SVM
“features”



Pandora - Consolidated Reconstruction



Pandora - Performance Metrics

1. Determine target MCParticle associated to each hit

- Use MCParticle hierarchy to determine primary “targets” for reco
- Associate hits to target MCParticle making largest E contribution

Target MCParticles must satisfy quality cuts

Reco/MCParticles matches must satisfy quality cuts.

2. Match reco particles to target MCParticles

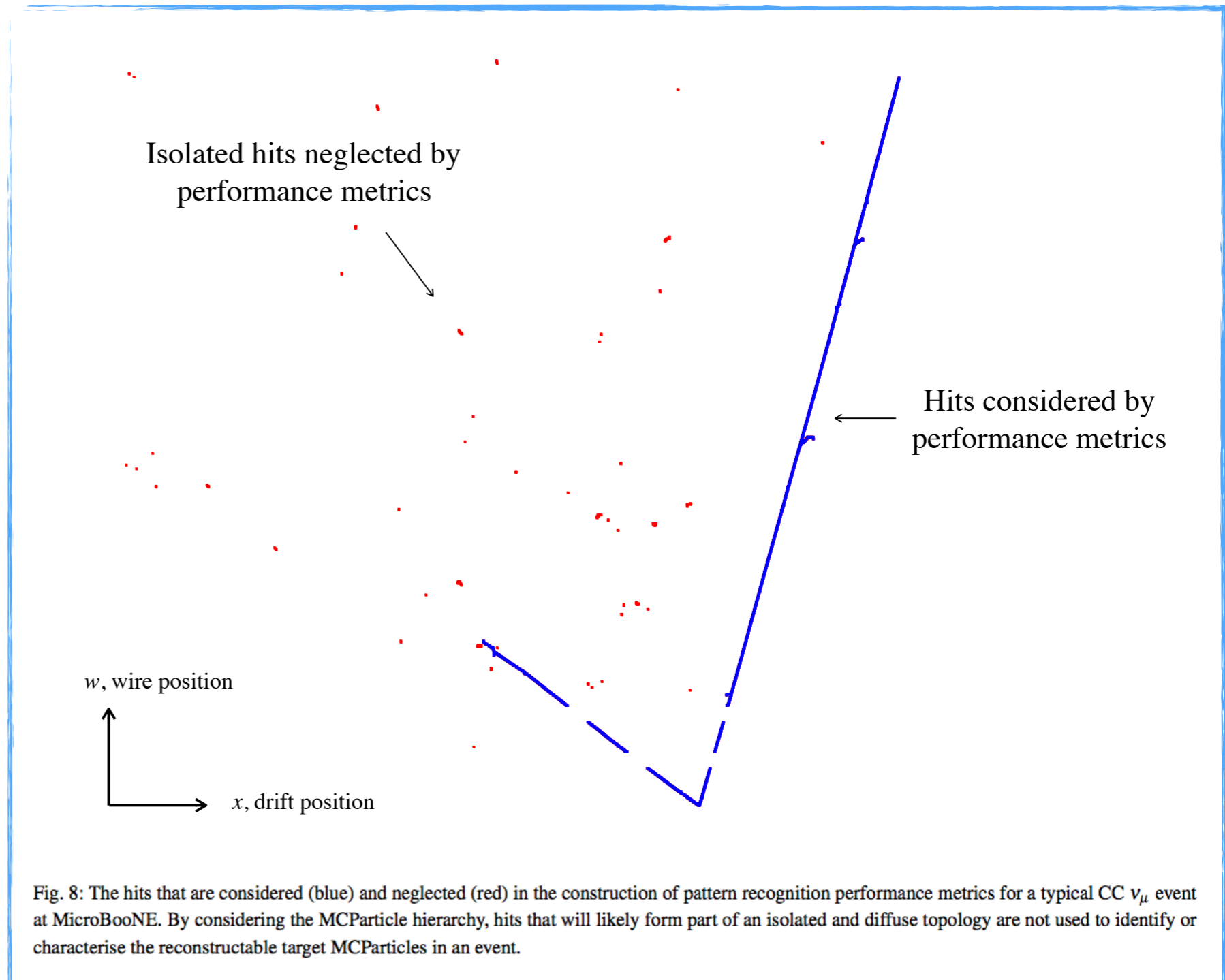
- For each combination of reco particle and target MCParticle, find the number of shared hits; fold all daughter particles, in both reco and MCParticle hierarchies, back into parent primaries
- Interpret raw/comprehensive matching information to clarify pattern recognition performance:
 - i. Find strongest (most shared hits) match between any reco particle and target MCParticle
 - ii. Repeat step i, using reco and MCParticles at most once, until no further matches possible
 - iii. Assign any remaining reco particles to target MCParticle with which they share most hits

3. Define performance metrics

- **Efficiency:** Fraction of target MCParticles with at least one matched reco particle
- **Completeness:** Fraction of MCParticle true hits shared with the reco particle
- **Purity:** Fraction of hits in reco particle shared with the target MCParticle
- **Match exactly one reco particle to each target MCParticle \Rightarrow Event is “correct”**

Pandora - Performance Metrics

- In practice, some MCParticles not reconstructable. Targets must satisfy quality cuts:
- ≥ 15 hits in total, at least five hits in at least two views.
- Target must deposit $>90\%$ E in these hits.
- Plus, ignore all hits which are downstream of far-travelling neutron in MC hierarchy.



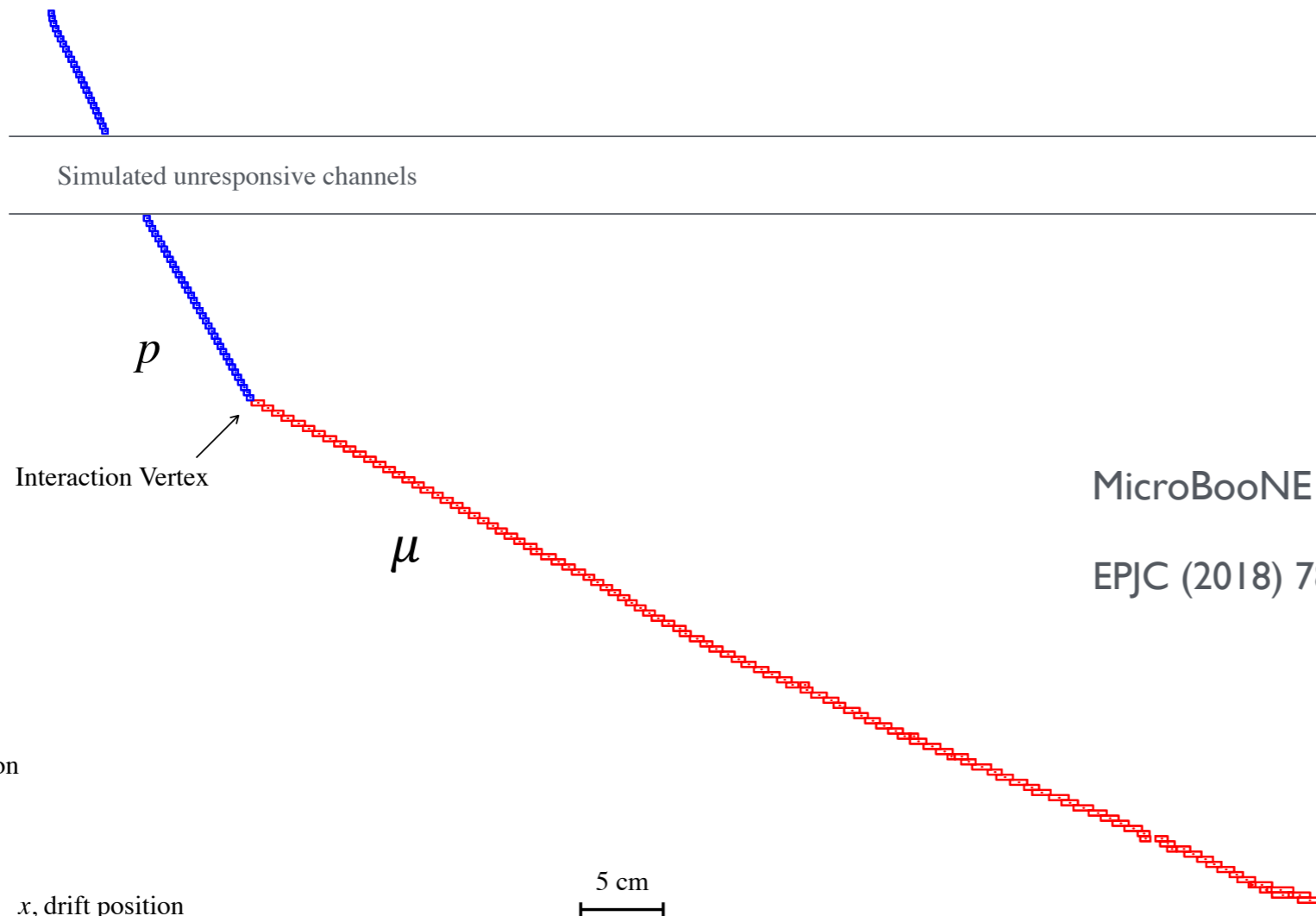
MicroBooNE - CC QE: $\nu_\mu + \text{Ar} \rightarrow \mu^- + p$

Clean topology: ν_μ CC QE interactions with exactly one reconstructable muon and one reconstructable proton in visible final state:

No cosmic rays here

#Matched Particles	0	1	2	3+
μ	1.3%	95.8%	2.9%	0.1%
p	8.9%	87.3%	3.6%	0.2%

53,168 events, 86.0% have exactly one reco particle matched to each target.

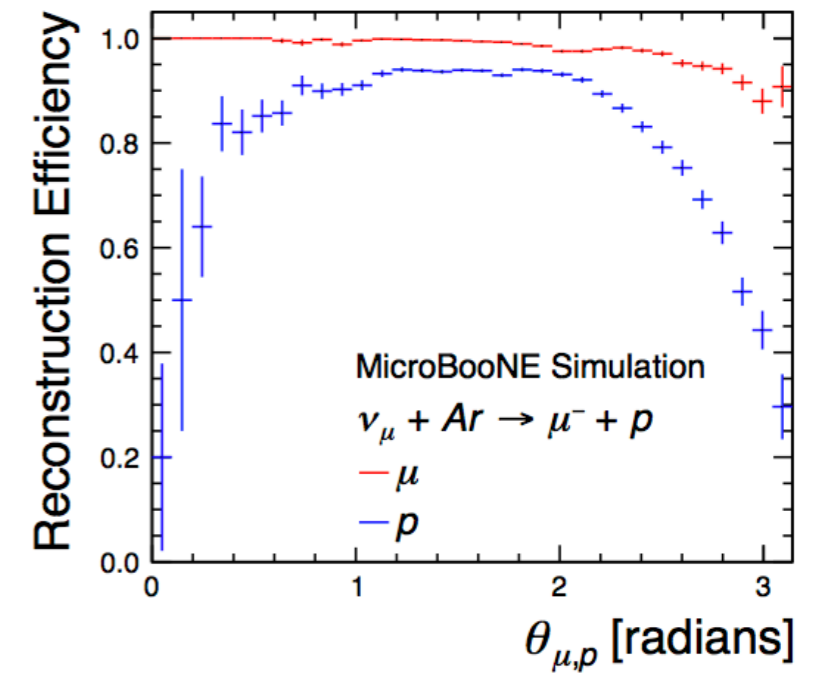
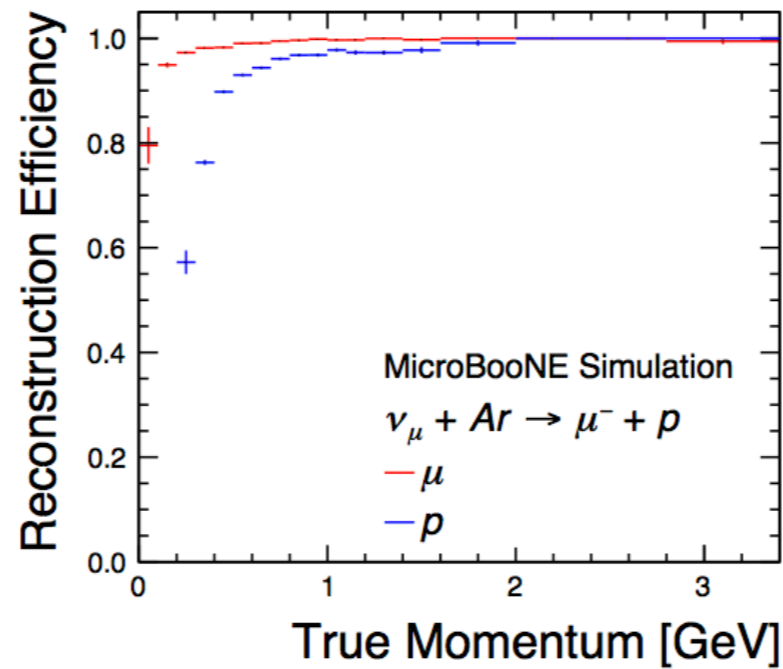
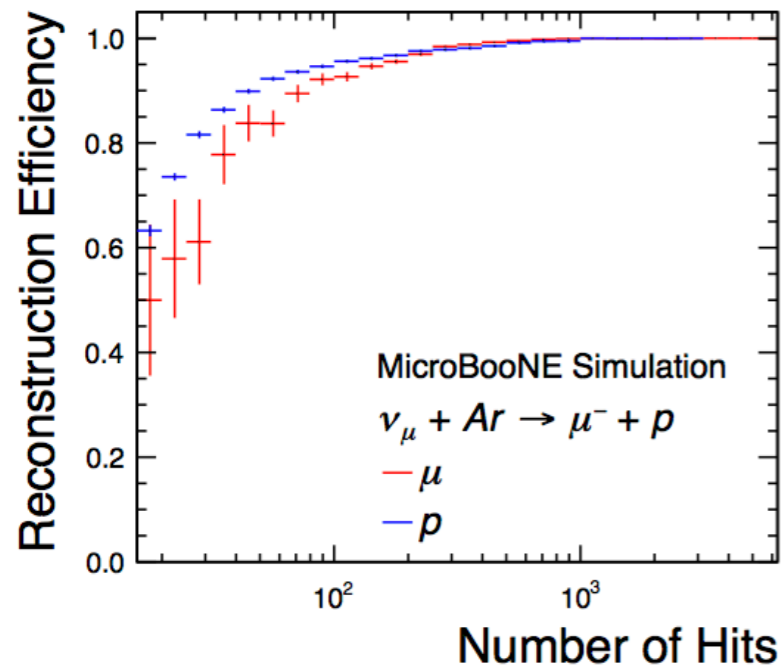


MicroBooNE simulation

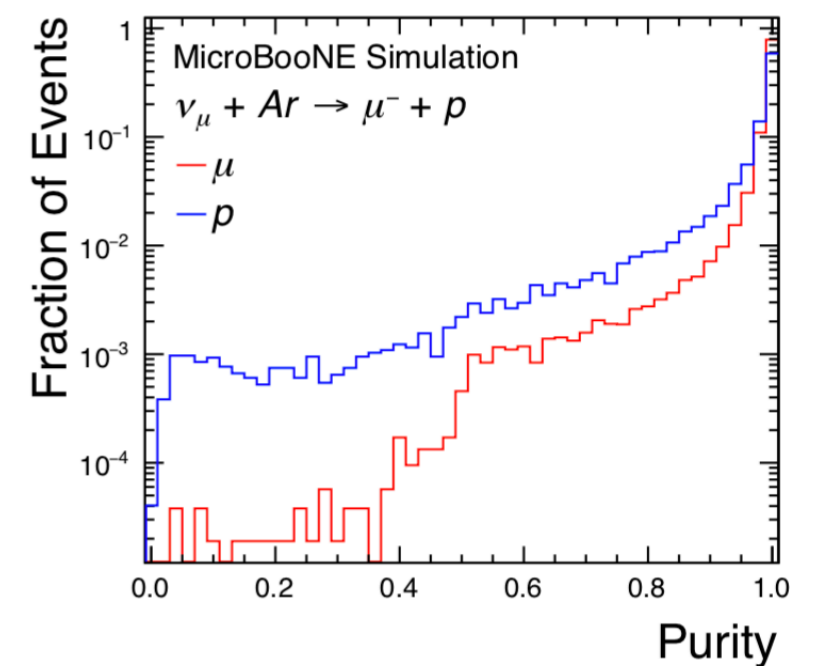
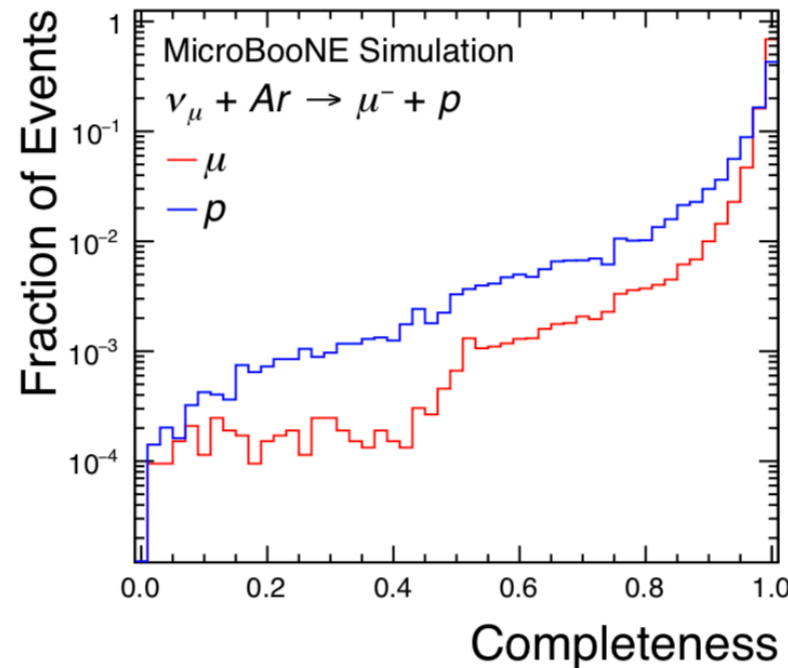
EPJC (2018) 78:82

520 MeV ν_μ

MicroBooNE - CC QE: $\nu_\mu + Ar \rightarrow \mu^- + p$



- The most common failure mechanism is merging muon and proton into a single reconstructed particle.
- Single particle is matched to target with which it shares most hits, which will preferentially be the muon.



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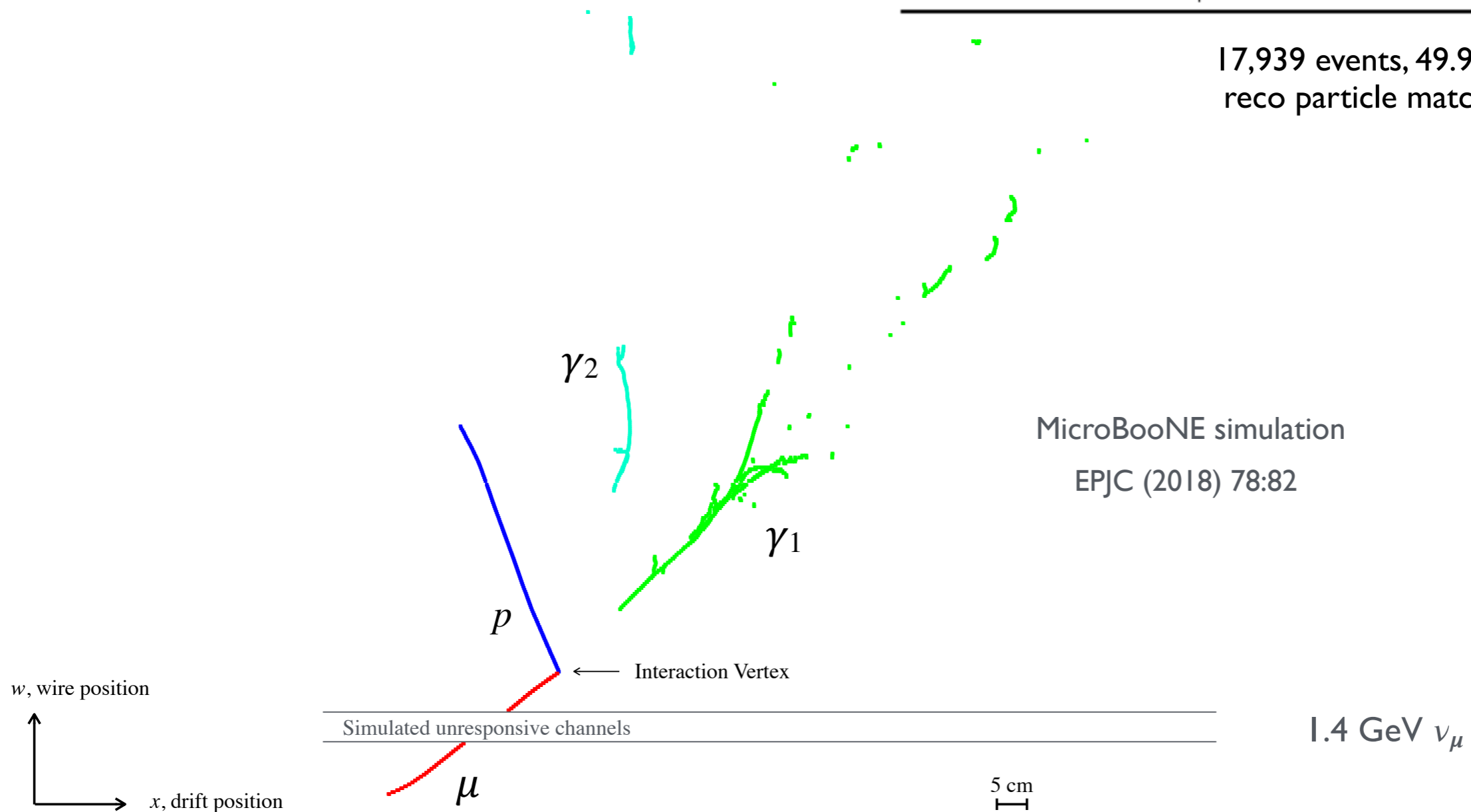
MicroBooNE - CC RES: $\nu_\mu + \text{Ar} \rightarrow \mu^- + p + \pi^0$

Two-photon topology: CC ν_μ interactions with resonant neutral pion production:

#Matched Particles	0	1	2	3+
μ	3.7%	94.8%	1.5%	0.0%
p	9.9%	85.5%	4.3%	0.3%
γ_1	6.8%	88.0%	4.8%	0.4%
γ_2	29.9%	66.4%	3.6%	0.2%

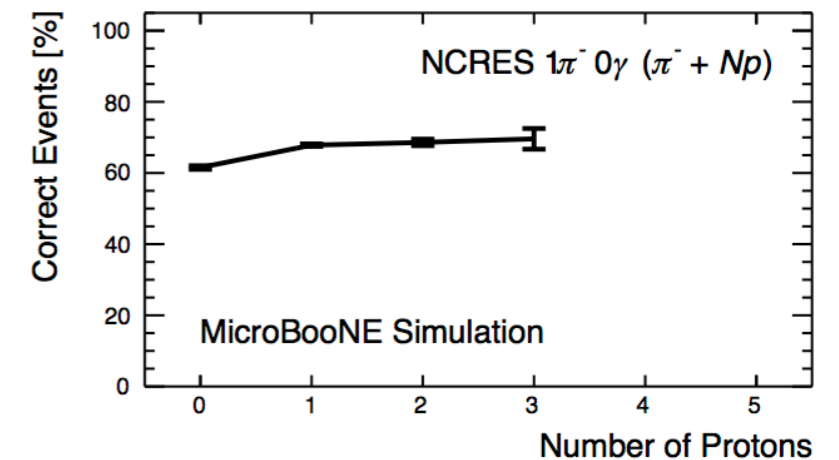
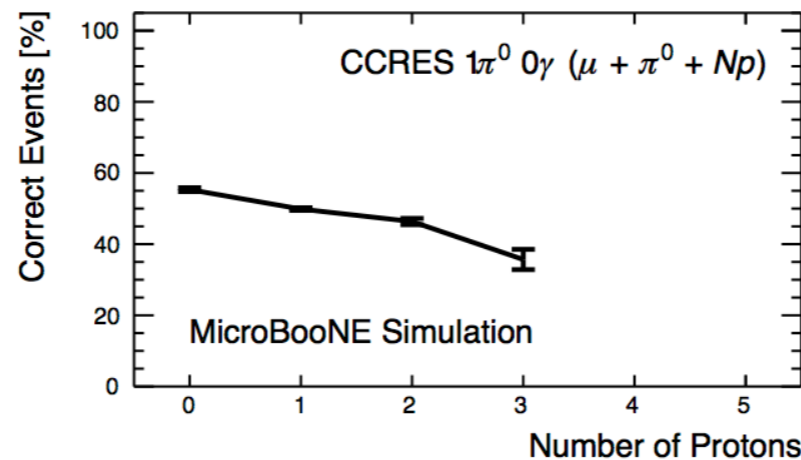
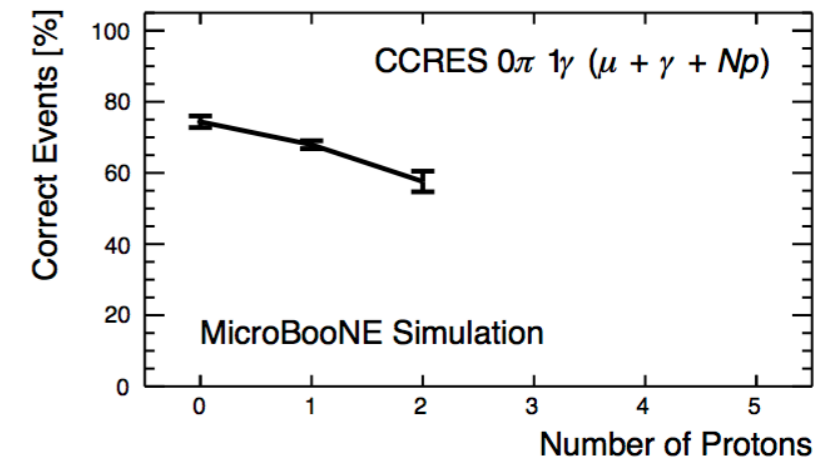
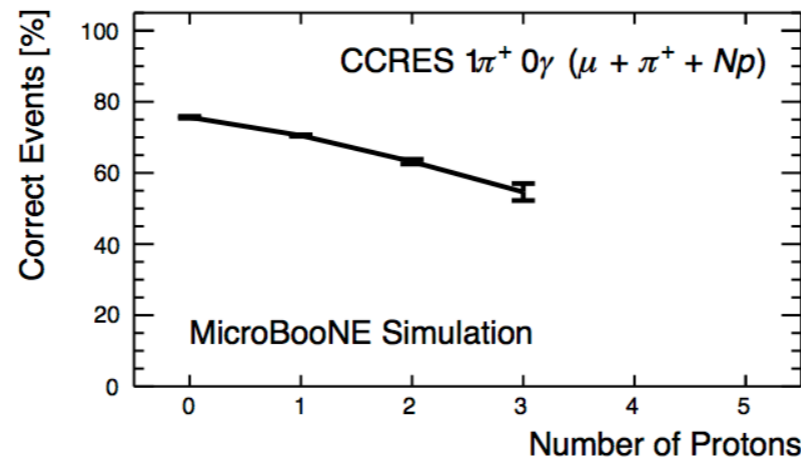
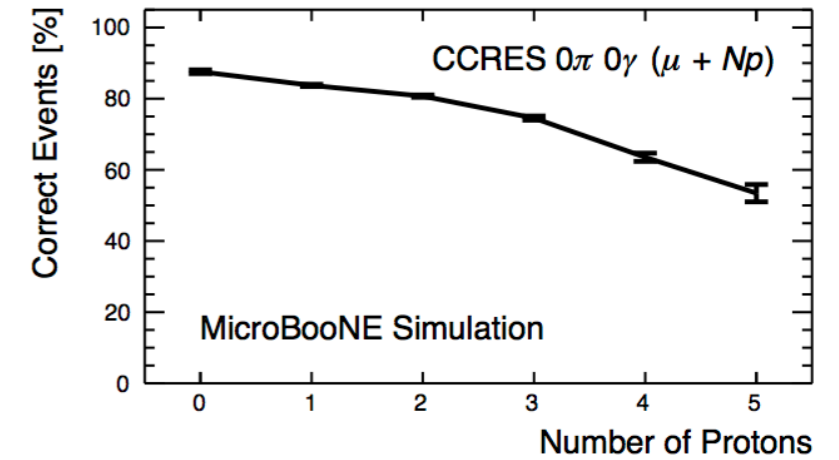
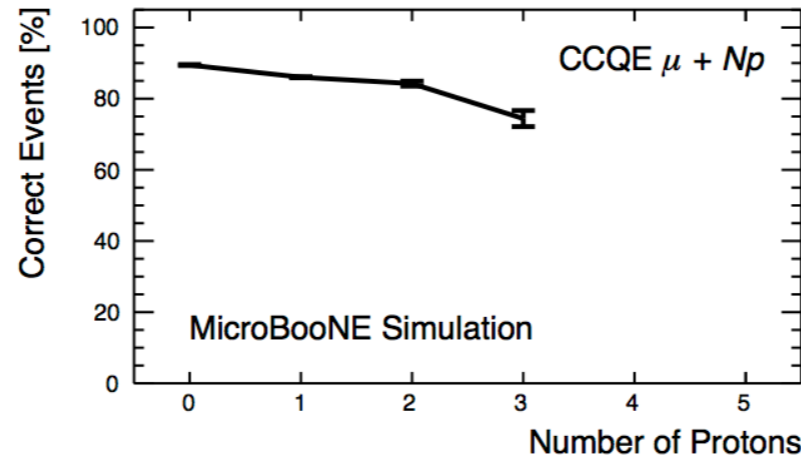
17,939 events, 49.9% have exactly one reco particle matched to each target.

#hits γ_1 > #hits γ_2



MicroBooNE - Selection of Exclusive Final States

- Assess larger selection of exclusive final states using correct event fraction.
- Recall aim: a general purpose reconstruction for diverse event topologies.



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