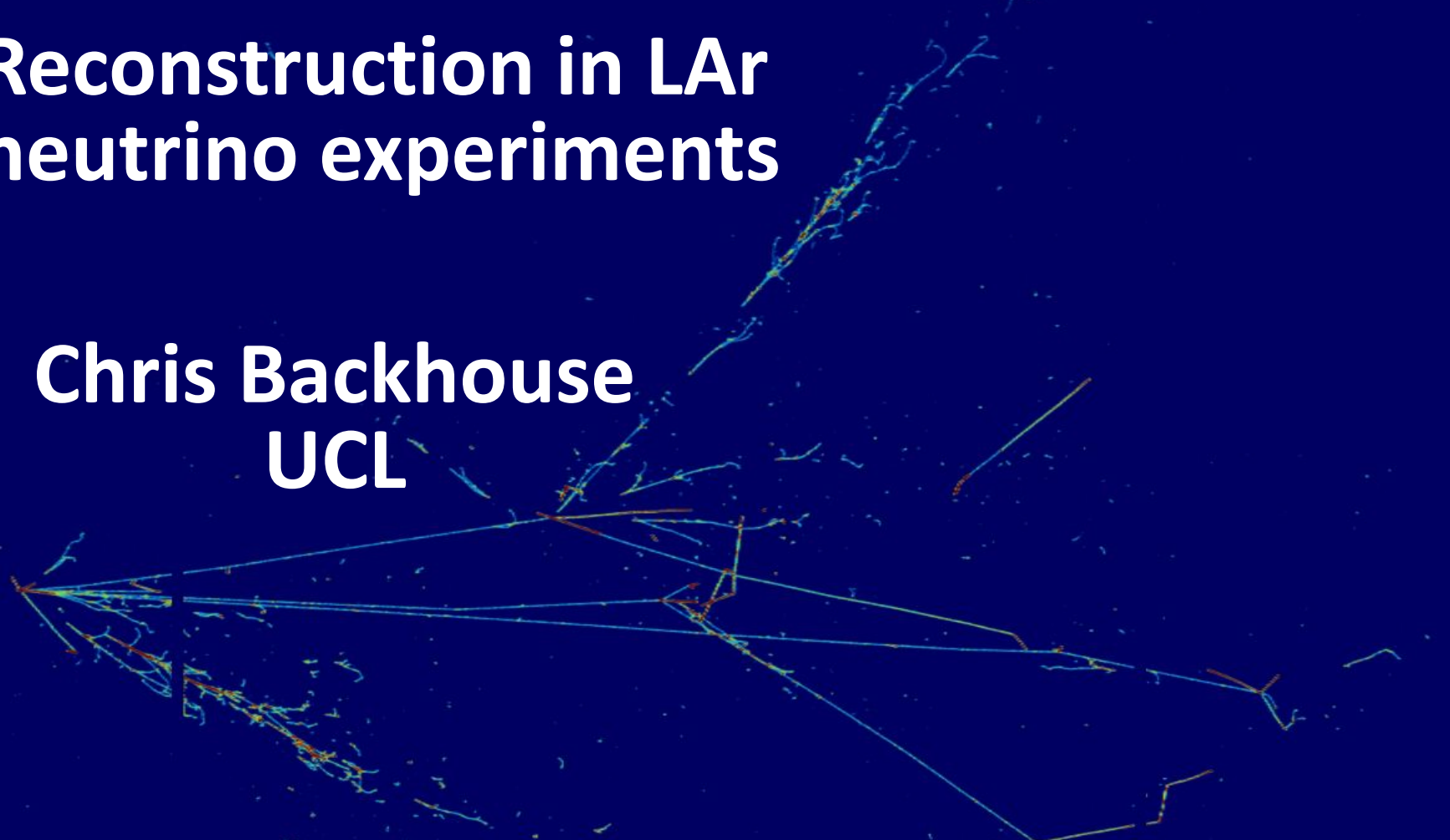


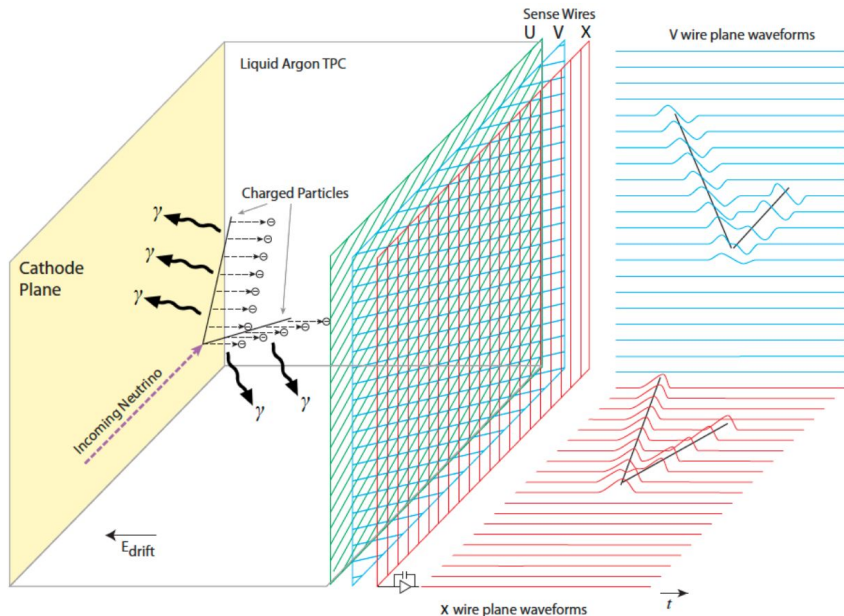
Reconstruction in LAr neutrino experiments

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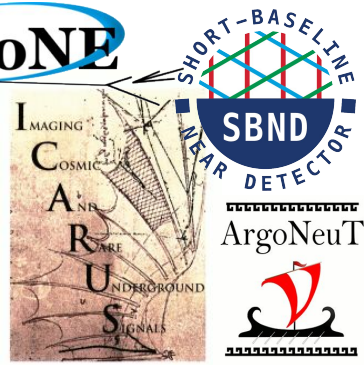
Introduction



- Thanks to Andy Chappell (Warwick) for many of these slides 2
- Need to reconstruct the flavour and energy of the incident neutrino
- Ideally reconstruct detailed final state
- Other event sources
 - Cosmic rays, SN neutrinos, nucleon decay, atmospheric neutrinos, calibration sources...

- I am presenting with a DUNE-oriented bias, but there is a whole family of LArTPC neutrino experiments:
 - DUNE (HD, VD, ND), ArgoNeuT, MicroBooNE, SBND, Icarus...
- Lots of shared infrastructure and cross-pollination
 - nutools, larsoft, wirecell, pandora...

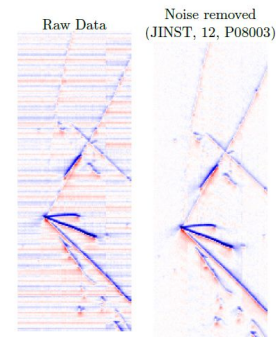
μBooNE



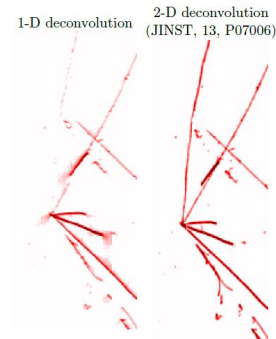
DUNE

Reconstruction stages

- “DataPrep” - remove pedestals, correlated noise, etc
 - Eminently parallelizable

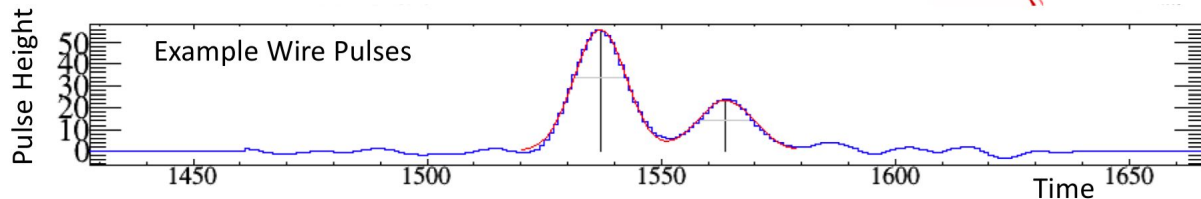


- Deconvolution - map bipolar induction signals to unipolar
 - Lots of Fourier transforms



- Hit finding
 - Also needs to happen online for trigger

- **Pattern recognition**



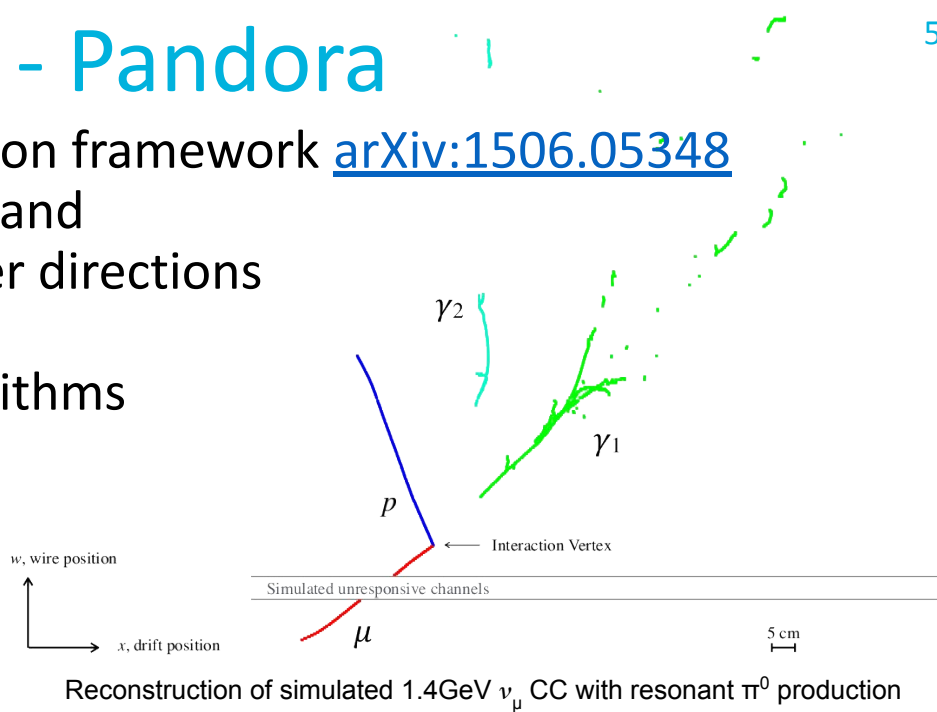
- Event ID
 - Traditional MVAs based on reconstructed particles, or full event image CNN (CVN)

Documentation

- **MicroBooNE have done a great job documenting their analysis. Some highlights:**
 - “Wire-Cell 3D Pattern Recognition Techniques for Neutrino Event Reconstruction in Large LAr TPCs”, [arXiv:2110.13961](https://arxiv.org/abs/2110.13961)
 - “Cosmic Ray Background Rejection with Wire-Cell LAr TPC Event Reconstruction in the MicroBooNE Detector”, [arXiv:2101.05076](https://arxiv.org/abs/2101.05076)
 - “Semantic Segmentation with a Sparse Convolutional Neural Network for Event Reconstruction in MicroBooNE”, [arXiv:2012.08513](https://arxiv.org/abs/2012.08513)
 - “Neutrino Event Selection in the MicroBooNE LArTPC using Wire-Cell 3D Imaging, Clustering, and Charge-Light Matching”, [arXiv:2011.01375](https://arxiv.org/abs/2011.01375)
 - “A Convolutional Neural Network for Multiple Particle Identification in the MicroBooNE LArTPC”, [arXiv:2010.08653](https://arxiv.org/abs/2010.08653)
 - “Vertex-Finding and Reconstruction of Contained Two-track Neutrino Events in the MicroBooNE Detector”, [arXiv:2002.09375](https://arxiv.org/abs/2002.09375)
 - “A Deep Neural Network for Pixel-Level Electromagnetic Particle Identification in the MicroBooNE LArTPC”, [arXiv:1808.07269](https://arxiv.org/abs/1808.07269)
 - “The Pandora Multi-Algorithm Approach to Automated Pattern Recognition of Cosmic Rays and Neutrinos in MicroBooNE”, [arXiv:1708.03135](https://arxiv.org/abs/1708.03135)
 - “Convolutional Neural Networks Applied to Neutrino Events in a LArTPC”, [arXiv:1611.05531](https://arxiv.org/abs/1611.05531)
- **Other experiments**
 - “Cosmic Background Removal with Deep Neural Networks in SBND”, [arXiv:2012.01301](https://arxiv.org/abs/2012.01301)
 - “Neutrino interaction classification with a convolutional neural network in the DUNE far detector”, [arXiv:2006.15052](https://arxiv.org/abs/2006.15052)

Pattern recognition - Pandora

- Pandora is a widely-used reconstruction framework [arXiv:1506.05348](https://arxiv.org/abs/1506.05348)
- Goal is to cluster hits by true particle and reconstruct track trajectories / shower directions
- Multi-algorithm approach
- Mostly “classical” so far, but ML algorithms can slot in where appropriate
- Focus on two today:
 - Track/shower CNN
 - Vertex CNN



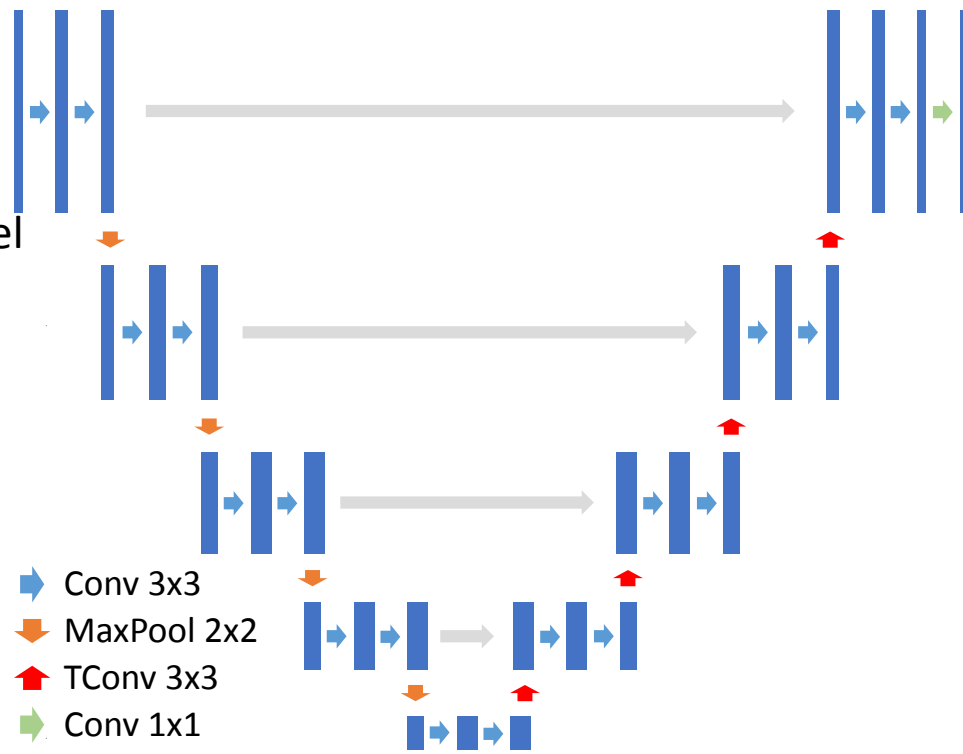
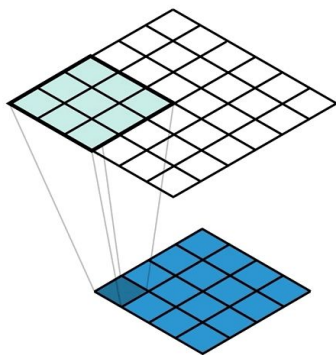
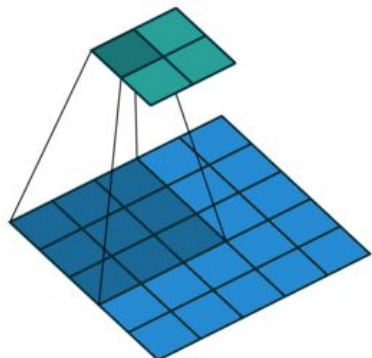
- Other event reconstruction approaches exist
- e.g. the three strands of the recent MicroBooNE result used
 - Semantic segmentation
 - Pandora
 - WireCell <https://lar.bnl.gov/wire-cell/>

Track / shower separation

- Different algorithms are appropriate for track-like and shower-like parts
- Classify hits up-front using a network
- U-Net architecture developed for biomedical image segmentation in 2015

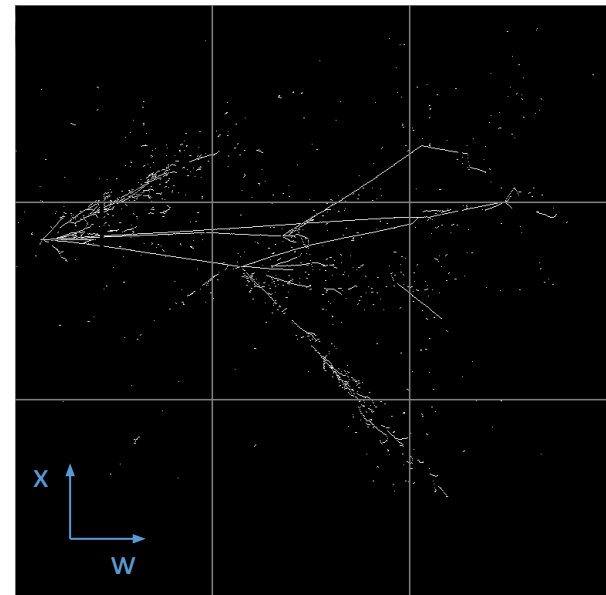
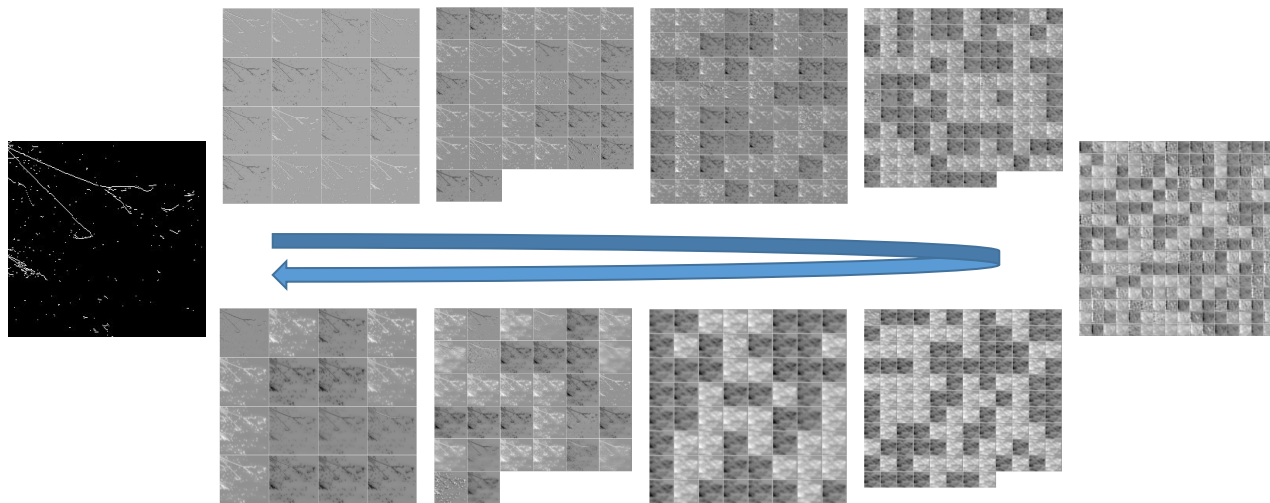
[arXiv:1505.04597](https://arxiv.org/abs/1505.04597)

- Downsampling convolutions
- Upsampling transpose convolutions
- Skip connections
- Assign track/shower prob. to each pixel

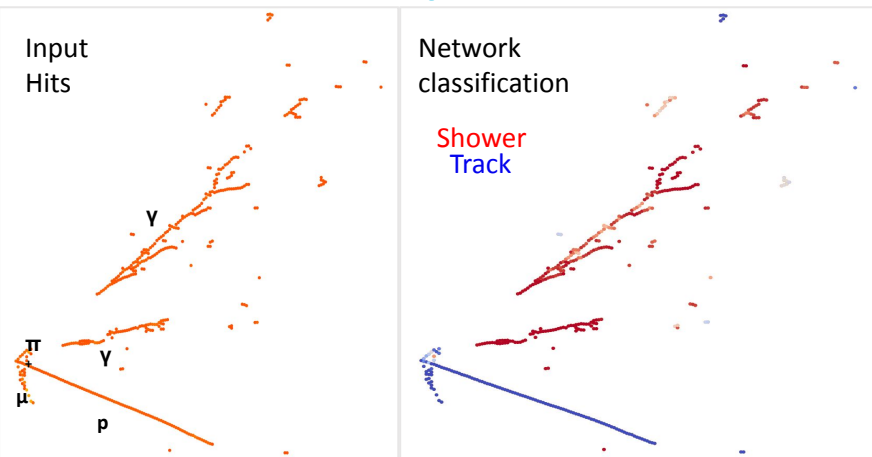


Track / shower separation

- Train on 100k simulated neutrino interactions
- Break image into 256^2 pixel tiles
- Allows a degree of sparsification (skip empty tiles)



Track / shower separation

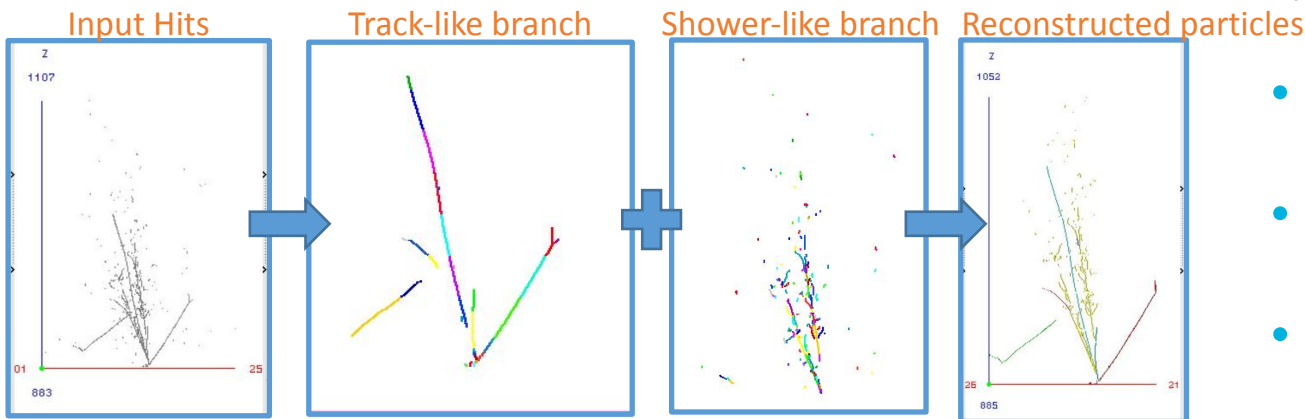


True\Net	Shower	Track
Shower	95.0%	3.9%
Track	5.0%	96.0%
Σ	100%	100%

Confusion matrix for W view

- Pandora algorithms provide initial clusters
- Classify clusters based on majority vote of hit-level classification
- Algorithms can target separated track and shower topologies

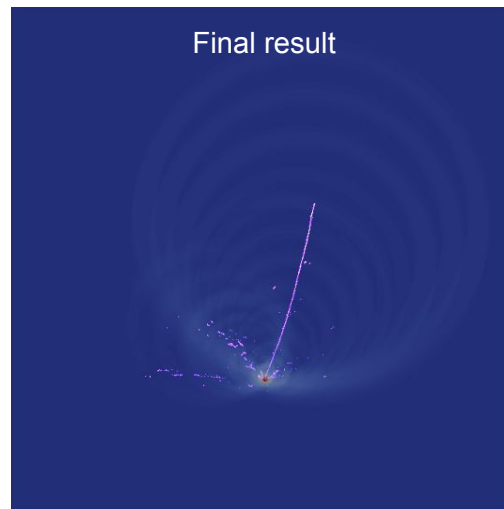
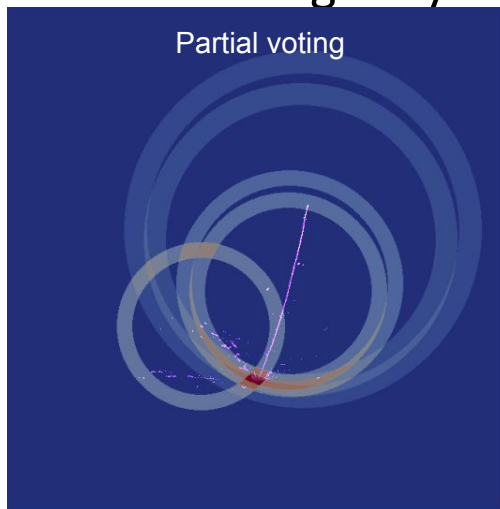
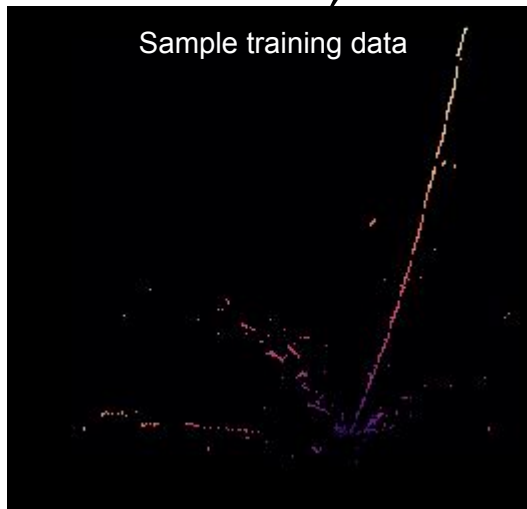
Example ~ 2 GeV CC DIS interaction



- Executed on CPU in production contexts
- Some experimentation with GPU-as-a-service
- Wrapper around NVidia's Triton API

Vertex finding

- Location of neutrino interaction vertex is an important reconstruction hint
 - Reconstruct showers directed away from the vertex
 - Be less willing to merge reco objects close to the vertex
- Regression networks exist for vertex position, neutrino energy, lepton angle
[arXiv:2012.06181](https://arxiv.org/abs/2012.06181)
- Pandora experimenting with differently-designed DL vertexing algorithm
- Rather than regressing single vertex position, network predicts distance of each hit to vertex, result based on a Hough-style vote



Conclusion

- Low level data processing ripe for parallelization / acceleration
- Full-event classification and regression networks trivially so
- My impression: neutrino reconstruction less “monolithic” than colliders, no single algorithm that should be obsessively optimized
- Reconstruction is often using hybrid of traditional approaches with DL elements
 - Highlighted a couple of examples here
 - Interesting challenge to make efficient use of acceleration for small subsets of the workflow
- Other whole-event approaches exist, both for classification/characterization and full reconstruction
 - Graph networks
 - 3D sparse convolutions
 -