



Pandora for dual phase LArTPCs

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Pandora for dual phase LArTPCs

- Pandora is a multi-algorithm approach to pattern recognition, deployed in many LArTPC experiments
 - Harnesses physics and detector information to maximise reconstruction potential
- Most of Pandora's algorithms are suitably detector agnostic and work 'out of the box' for dual-phase
- A few of Pandora's most powerful algorithms require three readout views (single phase LArTPCs) to operate
 - Unusable by a two readout view dual phase LArTPC
- The following slides overview a tailored end-to-end reconstruction workflow for dual phase LArTPCs in Pandora, including expansion and adaptation of key algorithms
 - The new Pandora workflow is a part of the ProtoDUNE-DP offline production for cosmic rays

2D reconstruction: using hit widths

- 2D clustering algorithm which assesses hit widths
- Developed by Isobel Mawby for DUNE-FD
- Improves reconstruction performance for sparse tracks
 - Such tracks are typically almost parallel to the drift axis



3

0.4

0.2

Hit width algo off

-20

-10

Particle gun muons in ProtoDUNE-DP

30

 θ_{OYZ} (°)

2D reconstruction: LEM and CK CRP gaps

- 'Blind' regions of the detector between adjacent LEMS and adjacent CRPs
- Gap positions now implemented in pandora
- Knowledge of the gaps provides guidance to Pandora's 2D clustering algorithms





- Matches compatible 2D clusters across readout views to create 3D clusters
- Pandora for single-phase/3view LArTPCs utilises all three views simultaneously to maximise 2D->3D matching potential
- Pandora for dual-phase/2-view LArTPCs now utilises
 calorimetric information for the first time to match clusters across views

- Compare all pairwise cluster combinations across views
- Find the time-overlap region for each cluster pair
- Create fractional charge profiles for each cluster in said overlap region (red and black histograms)

Fractional charge (no units)

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- For each cluster comparison, calculate:
 - Global matching score*
 - Local matching scores* for regions of the overlap (blue curves)
- A high score (global or local) indicates a good match

*Score based on correlation coefficient p-value

Di-muon particle gun in ProtoDUNE-DP



Most recent 2D->3D matching developments

- Packaged the calorimetric matching metrics into tools which decide if two 2D clusters match together to make a 3D particle
- Most tools target a particular topology

 harnesses geometry and
 calorimetry together
- Implemented tools
 - Clear tracks tool
 - Long tracks tool
 - Simple tracks tool



Clear tracks tool

- Two clusters only overlap each other
- The calorimetry matching metrics between the two clusters are suitably high

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Long tracks tool

- Cluster(s) match multiple other clusters in the opposing view
- Ambiguity broken by picking comparisons with sufficiently long regions that calorimetrically match

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Simple tracks tool

- No fancy diagram
- Ranks all remaining cluster comparisons and repeatedly picks the highest-rank comparison
- Ranking based on
 - Locally matched fraction
 - Fraction of local matching scores above threshold
 - Global matching score
 - Number of matched points
 - The raw number of local matching scores above threshold

Reconstruction performance

- All recent developments have been very focussed on ProtoDUNE-DP due to recent data taking preparation
- All performance assessments have revolved around simulated high multiplicity cosmic ray events

Pandora development iteration Fraction of cosmic rays correctly reconstructed

Starting point

46%



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Fraction of cosmic rays correctly reconstructed

Starting point

+ Hit width 2D clustering and two-view 2D->3D matching (clear tracks tool only)

57%

46%



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Pandora development iteration	Fraction of cosmic rays correctly reconstructed
Starting point	46%
+ Hit width 2D clustering and two-view 2D->3D matching (clear tracks tool only)	57%
+ Long tracks and simple tracks tools	76%



Future work

- Assess/benchmark performance to target improvements
- 2D->3D matching tool for matching across the 'otherview' detector gaps
 - See diagrams





Summary

- Pandora is a multi-algorithm approach to pattern recognition, currently deployed in multiple LArTPC experiments
- A few of Pandora's most powerful algorithms require three views to function and so are not appropriate for dual phase-style LArTPCs
- We have expanded/adapted some key areas of Pandora's workflow to harness features of dual-phase style LArTPCs
- There now exists a tailored complete end-to-end reconstruction workflow for dual-phase style LArTPCs in Pandora
 - The workflow already features in ProtoDUNE-DP's offline reconstruction chain

- 2D->3D matching takes 2D clusters (e.g. from each wire view) and matching them across views to make 3D objects
- Pandora's main 2D->3D matching algorithm requires a cluster in three distinct views to function
 - Combining positions from clusters in two of the views infers a position in the third view. A pseudo chi2 is calculated for inferred vs actual positions along the cluster
- This is problematic for any detector technology which only has two views (e.g. the CRP-based dual-phase LArTPCs)
- A solution: Use the charge depositions along the 2D clusters to provide the redundant information to over constrain the matching
- The following slides outline a new pandora algorithm which harnesses the charge depositions to help inform the 2D->3D matching in a two-view LArTPC



- 1. Compare every 2D cluster in one view with every cluster in the other view (each one of these comparisons is a **matching candidate**)
- 2. For each matching candidate, find the region along the drift coordinate that the two clusters share (the **X overlap**)







- Slide a window across the profiles.
 Calculate the correlation coefficient p-value (p) for the points in the window
 - Define L=1-p for each window



601

View 0

X overlap

601 2D->3D matching View 0 6. Fraction of windows with L, local matching score X overlap (blue line on plot), above threshold (currently 0.99) indicates a good match 0.0° Fractional charge (no units) *ocal* **DUNE Preliminar** X matching 0.06 View 0 **8.0** View 1 0.05 score 0.6 0.04 (no units 0.4 0.03 View 1 0.02 80 20 40 60 100 120160 180 200 140Position along drift axis (cm) -252



Same di-muon event, downsampling factor == 5 (factor currently used in the codebase)



Matching performance (10 GeV di-muon, complete overlap)



 Excess of incorrect matches at local. match. frac. == 1



- Excess reduced when rejecting matching candidates with the minimum number of sampling points
- 4% of the correct matches are also dropped (almost all from the 0 bin)

Example incorrect match with locally matched fraction==1



£

- A lot of EM activity from both muons
- The EM activity results in a lot of small 2D clusters
- The bad match consists of
 - 16 hit electron cluster in view 0
 - A very small segment of one of the muon clusters in view 1

Example 'correct' match where locally matched fraction==0



- EM activity is to blame, again
- This match consists of
 - Modestly sized view 0 cluster that truth matches to primary muon
 - A much larger primary muon cluster in view 1
 - Both clusters truth match to the same muon



Matching performance (10 GeV di-muon, complete overlap)

- Plot shows the average local match fraction vs the number of samples (after downsampling) in each matching candidate
 - Error bars are the standard deviation (NOT standard error on the mean)
- Matching candidates with a low number of sampling points are dominated by small EM clusters



But does the calorimetry-based metric provide any more information than checking the the cluster X geometry?

- The two pairs of distributions show the fraction of the cluster span contained in the X overlap
 - For the 10 GeV di-muon sample
- The geometry-based separation is very minimal
- Disclaimer: This sample was specifically chosen so that there was maximal ambiguity in the cluster X geometry

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But does the calorimetry-based metric provide more information than checking the the cluster X geometry?

 Compare the geometry-based metrics (RHS plots) with the calorimetry-based metric (bottom)





Correlation coefficient's pvalue

- For uncorrelated bivariate normal distribution pairs, the correlation coefficient follows a Student t-distribution with n-2 degrees of freedom
- The t-value is

$$t=r\sqrt{rac{n-2}{1-r^2}}$$

- P-value is calculated by integrating the t-distribution above the calculated t value (a one tailed test)
 - H0: r==0
 - H1: r>0
- The t-distribution supposedly approximately holds for non-gaussian variables, provided the sample sizes are large enough. I'll revisit this in a few slides

Resampled fractional charge profiles (di-muon sample)



Toy study

- Revisiting the student t-distribution assumption
- Produce 10000 fake fractional charge profiles
 - Fill 3 histograms with landau throws, smeared with a gaussian
 - Two hists. are filled with the same landau values but smeared separately
 - Third hist filled with separate landau values
 - Each bin is filled N times with distinct throws to mimic the downsampling
- Calculate correlation coefficient and pvalue
- Landau (315, 13)
- Gaus (1,0.1)
- N hist bins == 30
- N samples per bin == 5





Toy study

- Top plot shows correlation coefficient for the 10,000 universes
 - Black: correlated distributions
 - Red: uncorrelated distributions
- Bottom plot shows corresponding p-values
 - The red distribution should be flat, but it is not

P-value vs r (t-distribution)





Toy study

- Instead, calculate the p-value using permutation tests
 - Randomly shuffle the bins for one distribution in a comparison and recalculate r
 - P-value == fraction of times you measure an r that is more extreme than your original r measurement
- Top plot shows correlation coefficient (same as previous slide)
- Bottom plot shows corresponding p-value

p-value vs r (permutation test)

