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LArTPC physics and reconstruction

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Excellence in Detector and Instrumentation Technologies/EDIT School 2023

Neutrino Oscillation



Neutrino oscillation experiments have provided the first evidence for physics beyond the Standard Model of particle physics



Neutrino Oscillation Experiments

- > 50 years
- > 30 experiments
- Phase space over tens of orders of magnitude

Courtesy: Hitoshi Murayama http://hitoshi.berkeley.edu/neutrino/



Three-v Paradigm













K2K JZR

Deep Underground Neutrino Experiment (DUNE)



- Search for new CP violation and determine the mass ordering through precision measurement of (anti)v_µ→(anti)v_e oscillation₃
 - Also search for proton decay and detection of supernova neutrinos
- Four 10 kT LArTPC detectors (each 20 x 20 x 70 m³)







Experimental Anomalies

There are a series of experimental anomalies hinting towards eV scale sterile neutrino(s)

- Reactor anomaly (missing anti-v_e?)
- Gallium anomaly/BEST (missing v_e?)
- Neutrino-4 (anti-v_e oscillation?)
- LSND and MiniBooNE (anti-v_e & v_e appearance?)



Phys. Rept. 427, 257 (2006)

 $N_v = 2.9840 \pm 0.0082$

If there are additional neutrinos beyond three, they just don't participate in weak interactions (i.e., "sterile")

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Reactor Antineutrino "Anomaly"





MiniBooNE Anomaly



- MiniBooNE (2002-2019) observed low-energy excess (LEE) with 4.8σ (systematics limited) significance
- If LEE is interpreted as v_e appearance in the primarily v_µ beam, would suggest 4th (sterile) neutrino

MiniBooNE: A Cherenkov Detector



An excellent e/ γ separation can be achieved with the Liquid Argon Time Projection Chamber (LArTPC) technology \rightarrow MicroBooNE is built to understand the nature of MiniBooNE LEE (e? γ ? or what?)



Principle of Single-Phase Liquid Argon Time Projection Chamber (LArTPC)

- ~mm scale position resolution with multiple 1D wire readouts
- Particle identification (PID) with energy depositions and topologies





Drift velocity 1.6 km/s \rightarrow several ms drift time

Why high res?



Event topology to separate EM showers (e/γ) from tracks (proton, muon)
Separation of e and γ : Gap Identification

1 1

Why Wire?



5.920 m

□Modular design for transportation and underground installation

□Three (1D) wire planes with ~ 2560 readout channels. In comparison, would require half million channels for a 2D pixel readout with similar resolution (~ 5 mm pitch)

Considerations of cost and power consumption of electronics inside LAr

Challenge in Automated Event Reconstruction





- How to convert the excellent resolution and calorimetry in these pictures to rigorous physics analyses?
 - Massive amount of information with tiny signal to background ratio → a big challenge for automated event reconstruction



JINST 12 P08003 (2017) JINST 13 P07006 (2018) Brouking Ven P07007 (2018) National NSTate P01036 (2020)

JINST 17 P01037 (2022)

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- 🔲 🗋 e- 21 MeV

4 MIP

Phys. Rev. Applied 15 064071 (2021) arXiv:2012.07928

10 15

5

20

Residual range [cm]

25

P05032 (2018) Phys. Rev P06043 (2021) ar

<u>JINST 13 P05032 (2018)</u> <u>JINST 16 P06043 (2021)</u>

MicroBooNE Data

Search for Low-Energy Excess in v_eCC

Comprehensive search for (examination of) the MiniBooNE lowenergy excess in v_eCC with multiple final-state topologies with different reconstruction paradigms

Channels	Reconstruction	Purity	Efficiency	Selected Events	References	d (no eL		
CCQE 1e1p	Deep Learning	75%	6.6%	25	PRD 105 , 112003	edicte		
1e0p0π	Pandora	43%	9%	34	PRD 105 , 112004	ed / Pı		
1eNp0π	Pandora	80%	15%	64	PRD 105 , 112004	bserv		
Inclusive 1eX	Wire-Cell	82%	46%	606	PRD 105 , 112005	vents (
						ш		



0.5 0.0 1e1p CCQE 1eNp0π 1e0p0π 1eX [200 MeV,500 MeV] [150 MeV,650 MeV] [150 MeV,650 MeV] [0 MeV,600 MeV]

2.5

2.0

1.5

1.0

Phys. Rev. Lett. 128, 241801 (2022)

No excess of low-energy v_e candidates!



MicroBooNE Observed

Total, no eLEE (x = 0.0)

Total, w/ eLEE (x = 1.0)

Non- v_e background

Intrinsic v_e

Wire-Cell Low Level Signal Processing

□ Noise filtering

- Coherent noise channel-bychannel correlated
- Signal processing
 - Reverse the "field response" and "electronic response"



Online display: http://lar.bnl.gov/magnify/



Impulse response

Room acoustic response





Image credit: www.prosoundweb.com

LArTPC field + electronic response





h(t)

2D Impulse response

Camera optical response



Hubble telescope impulse response (Image credit: http://web.mit.edu)

V Plane

LArTPC 2D response: y: t * drift velocity x: wire number * pitch

h(x,y)





2-D Deconvolution

2D measurement formation

$$M(t',x') = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R(t,t',x,x') \cdot S(t,x) dt dx + N(t',x')$$





Further improvement of LArTPC Signal Processing



Cold Electronics

Placing the preamplifier inside LAr significantly reduced the electronics noise

- 5-6 times compared to past warm electronics \rightarrow 60:1 MIP peak-to-noise ratio in the collection
- Significantly improve the performance of induction wire plane → An enabling technology

DNN ROI finding with multi-plane information

JINST 16 P01036 (2021)

Multi-plane information in Signal Processing

Brookhaven

National Laboratory

DNN ROI finding with multi-plane information

ProtoDUNE simulation ROI finding on V plane (2nd induction) Ref. 1.4 --- DNN w/o MP DNN w/ MP 1.2 Pixel Efficiency 8.0 Bixel Efficiency DNN With 3-plane information 0.4 0.2 0.0 75, 75 85, 85 87,75 87,85 87,87 80, 80 82,82 $\theta_{xz}(V), \theta_{xz}(U)$ 🗕 Ref. 1.4 DNN w/o MP 1.2 DNN w/ MP 1.0 Pixel Purity 9.0 0.4 0.2 0.0 80, 80 75, 75 82,82 85, 85 87, 75 87, 85 87,87 $\theta_{xz}(V), \theta_{xz}(U)$ Brooknaven National Laboratory JINST 16 P01036 (2021)

tested on ProtoDUNE data

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Wire-cell Tomographic Reconstruction

CAT Scan

- Detector (x-ray generator/receiver) moves across the object (body)
- Axial projections (~180) by detector rotation
- Cross section can be reconstructed at each position along detector movement

LArTPC

- Objects (ionizing electrons) move across detectors (wire planes)
- Axial projections (~3) by wire orientation
- Cross section can be reconstructed at each time slice along electron drift

Wire-Cell 3D Imaging Principle

Fig.1:Basic principle of tomography: superposition free tomographic cross sections S1 and S2 compared with the projected image P

https://en.wikipedia.org/wiki/Tomography

"Three-dimensional Imaging for Large LArTPCs", JINST 13, P05032 (2018)

 \sim

de-ghosting

- Solving alone cannot eliminate all ghosts
- In MicroBooNE, the situation is worse when 2-view blobs are allowed
 - 10% dead channels \rightarrow 3view only is not acceptable

Active detector if three live wires are required prior to tiling

Active detector if two live wires are required to tile

MicroBooNE

- 2view tiling is needed → more ghosts
- https://arxiv.org/abs/2011.01375

3view Vertice

2view Vertica

- □ de-ghosting: larger, connected blobs tends to be true
 - future AI/ML opportunity

3D Pattern Recognition

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Deep Learning based Neutrino Interaction Vertex Finding

Regressional segmentation with a sparse U-Net

- U-Net: efficiently use geometry info which is critical
 - compared to graph networks
- Regressional loss on distance based "confidence map" to use a region of points instead of only one
 - otherwise, data is highly imbalanced (Z. Cao etc, arXiv:1812.08008)
- Sparse: boosted computing efficiency with our sparse 3D data
 - Submanifold Sparse Convolutional Networks (B. Graham etc, arXiv:1706.01307)

Regressional segmentation

Initially we used Cross Entropy loss

- effectively only use the vertex information for one space point
- doesn't care about the distance between the prediction and the target.
 - while our main metric is this distance.
- \rightarrow encode the distance information for a region of points
- predicting the full "confidence map" instead of only one point

• current mapping:
$$\operatorname{Conf}_{\operatorname{truth}} = \exp\left(-\frac{\|\vec{x} - \vec{v}_{\operatorname{truth}}\|^2}{2\sigma^2}\right)$$

OpenPose: https://arxiv.org/pdf/1812.08008.pdf

Network structure and data format

Used SparseConvNet to realized 3D sparse conv. DNN https://github.com/facebookresearch/SparseConvNet

This work: https://github.com/HaiwangYu/uboone-dl-vtx

coordinates			features		label
Х	У	Z	q	•••	conf.
int	int	int	float		float
int	int	int	float		float
int	int	int	float		float

<u>SparseConvNet</u>

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Deep Learning based Neutrino Interaction Vertex Finding

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 v_e CC vertex identification efficiency

Distance between reco. and true vertex (cm)

Summary

□ LArTPC technology provides unique opportunities for new physics

- Successful of LArTPC based experiment requires efforts made to both hardware and software
 - Software can not solve everything
 - AI/ML fits many tasks in LArTPC reconstruction
 - $_{\odot}$ Human understanding of LArTPC is also important

backup

Demo

- □ Prerequisites: Linux OS (bare metal or VM)
- https://github.com/HaiwangYu/wct-demo
- □ Tools used in the demo:
 - git, Linux commands, singularity container, bash script, editor
 - wire-cell-toolkit (C++, jsonnet)
 - o gen/src/DepoTransform.cxx
 - o sigproc/src/OmnibusSigProc.cxx
 - ROOT

Construct Linear Equations

y	-	= Ax	
((0, 0, 0, 1, 1, 1)	(H1
ui		000111	H2
u2		111000	НЗ
v1	=	$0\ 0\ 1\ 0\ 0\ 1$	H4
v2		010010	НБ
(v3)		100100/	

y: measured charge signal on each wirex: the (unknown) true charge deposition in each possible cell

A: bi-adjacency matrix connecting wires and cells (determined solely by wire geometry)

Use two planes as an illustration

Solve underdetermined linear problem: regularization $\chi^2 = (y - Ax)^T \cdot V^{-1} \cdot (y - Ax)$ $\equiv ||y' - A'x||_2^2,$

- Previous example has 6 unknowns, 5 equations: under-determined system
- Adding constraints: find the sparsest solution (applies to most physics events): L0regularization

minimize $||x||_0$, subject to: y = Ax

(L0-norm: number non-zero elements)

NP-hard! https://web.stanford.edu/~yyye/lpmin_v14.pdf

$$x = (A^T V^{-1} A)^{-1} A^T V^{-1} \cdot y$$

non-invertible, 2 zero-eigenvalues out of 6.

Procedure

- □ Remove unknowns until equations can be solved, then find the best solution with the minimum χ^2
- □ a combinatorial problem
 - $\circ~$ 2 out of 6: 15 combinations
 - 10 out of 40: 0.8 billion combinations

Compressed Sensing (L1-regularization)

Breakthrough: mathematical proof that L0 problem can be well approximated by the L1 problem (Compressed Sensing, Candes, Romberg, and Tao, 2005.)

Emmanuel Candes. (Photo courtesy of Emmanuel Candes.)

Justin Romberg. (Photo courtesy of Justin Romberg.)

Terence Tao. (Photo courtesy of Reed Hutchinson/UCLA.)

https://arxiv.org/abs/math/0503066

minimize $||x||_1$, subject to: y = Ax

(L1-norm: sum of absolute values of the elements)

Or, equivalently, minimize

$$\chi^2 = (y - Ax)^T \cdot V^{-1} \cdot (y - Ax) + \lambda ||x||_1$$

Performance in Wire-Cell

Typically, ~tens of seconds to reconstruct the whole 3D image (originally a few hours)

JINST 13, P05032 (2018)

An example MicroBooNE event after 3D reconstruction

Finding Neutrino Interaction

- □ 3D Cluster based on proximity (kd-tree)
 - associate the light flash to the corresponding TPC cluster based on light pattern

BEST, Baksan Experiment on Sterile Transitions, nue source
 <u>C. Zhang, Neutrino Mass Hierarchy</u>