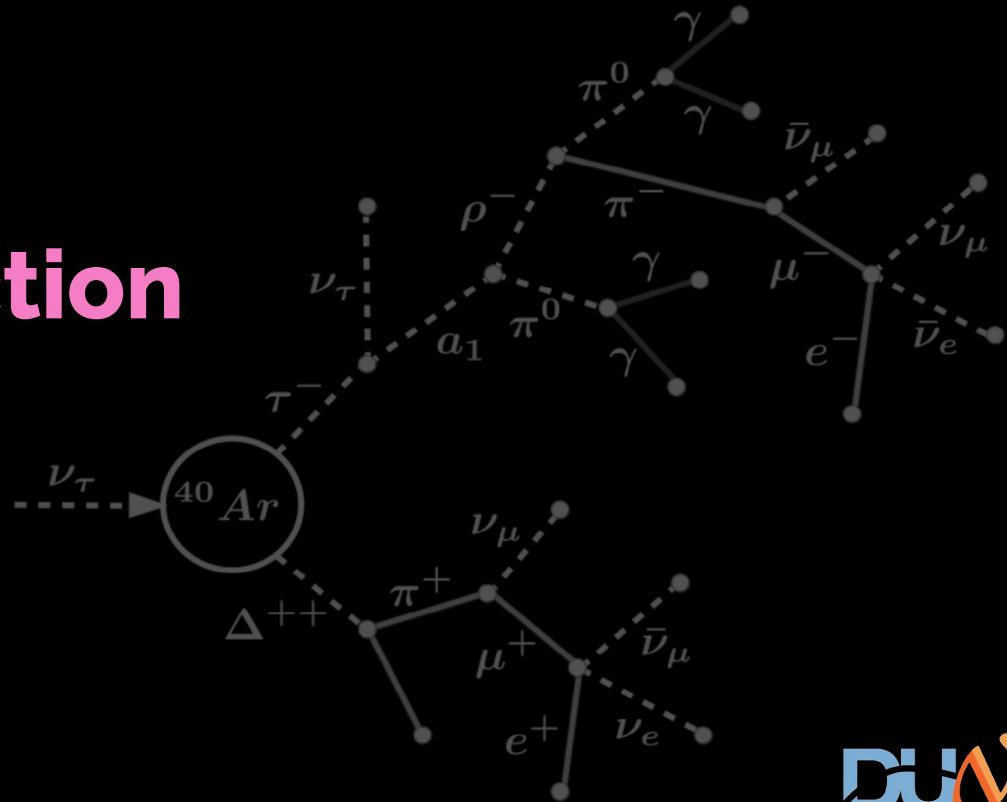


3D-Reconstruction of ν_τ in LArTPC Detectors



Barbara Yaeggy

On behalf of the DUNE Collaboration

Outline

1. The DUNE Experiment
2. ν_τ at DUNE
3. Graph Neural Networks
4. NuGraph
5. ν_τ Reconstruction via NuGraph

The Deep Underground Neutrino Experiment (DUNE)

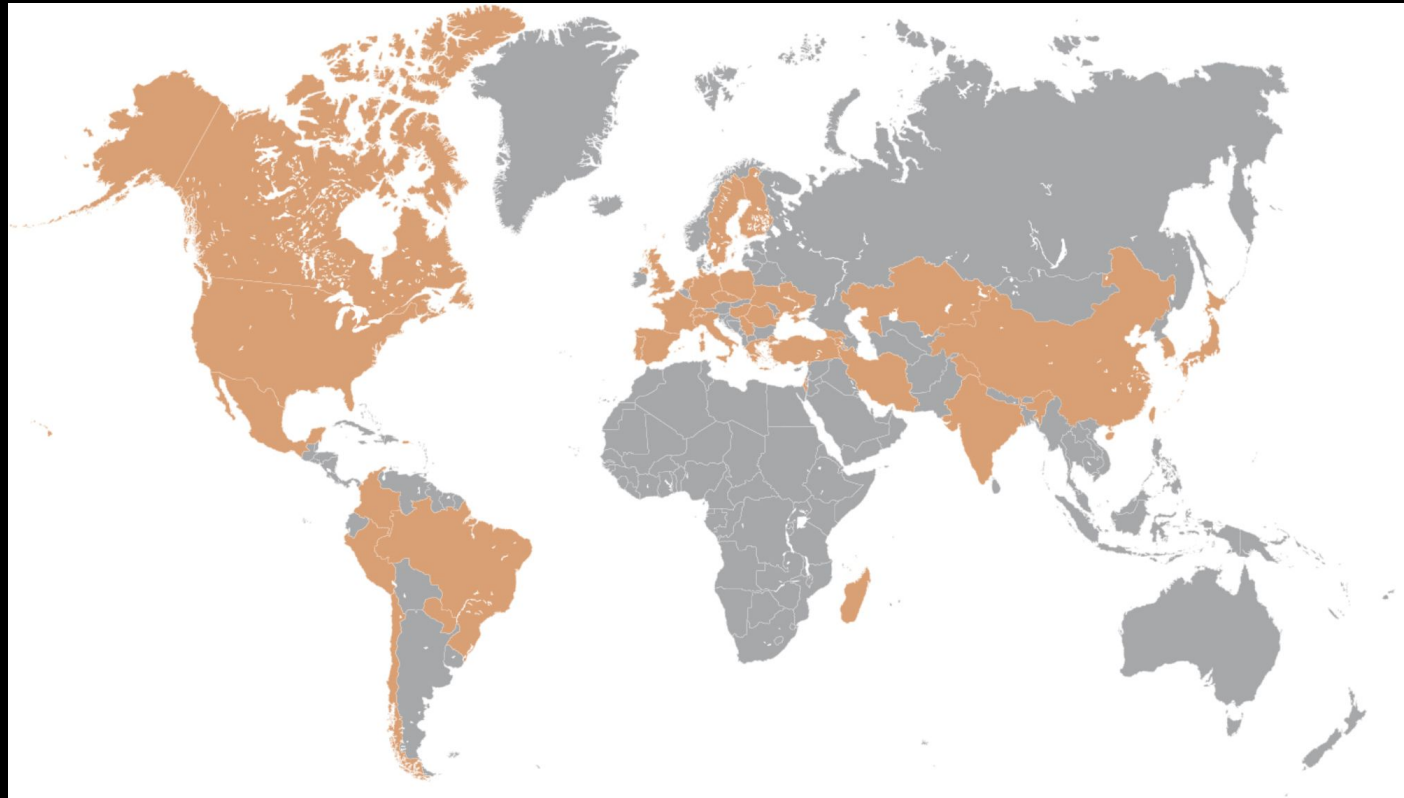
Currently under construction

~1450 collaborators

200 institutions

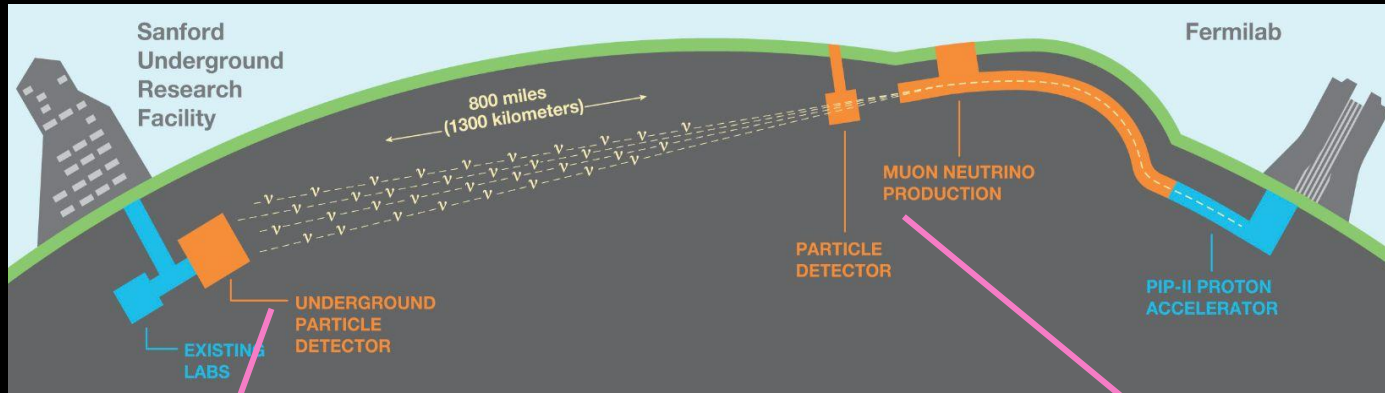
35 countries

DUNE will be able to constrain the three-massive-neutrinos paradigm by providing complementary measurements to those from the ν_e - appearance and the ν_μ - disappearance channels.



Detailed overview about DUNE in Martin Tzanov talk (morning session)

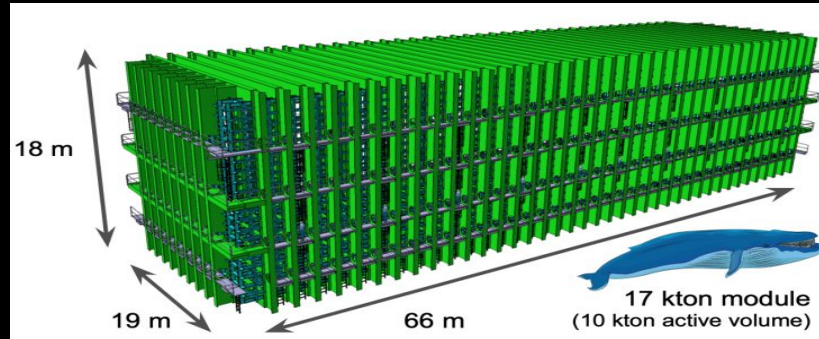
The Deep Underground Neutrino Experiment (DUNE)



Far Detector (FD)

- 1300 Km baseline, 1.5 Km deep
- 80% excavation done
- Detector commissioning expected in 2029
- Liquid argon time projection chamber (LArTPC) technology → high resolution neutrino interaction imaging
- 4x17 kton LArTPC modules.

Wideband neutrino beam, (~ 100 MeV – 10 GeV)



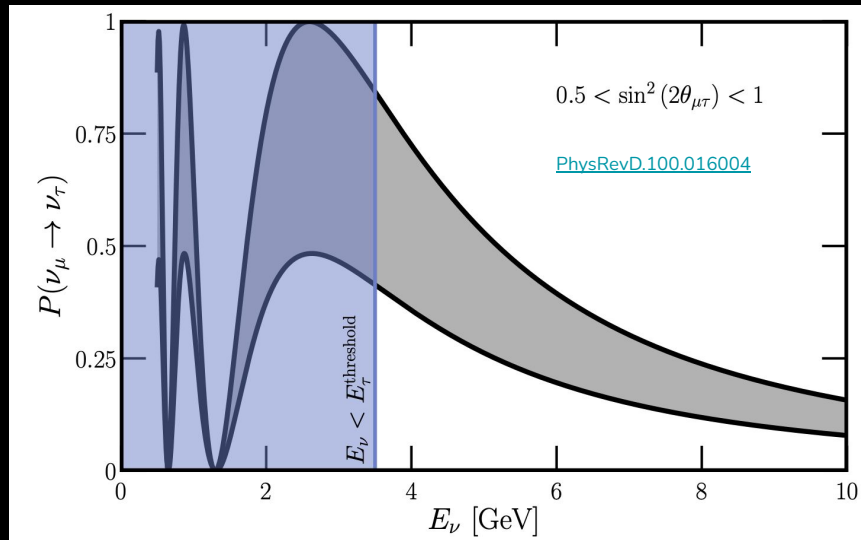
FD facility ~ 8 football fields

Near Detector (ND)

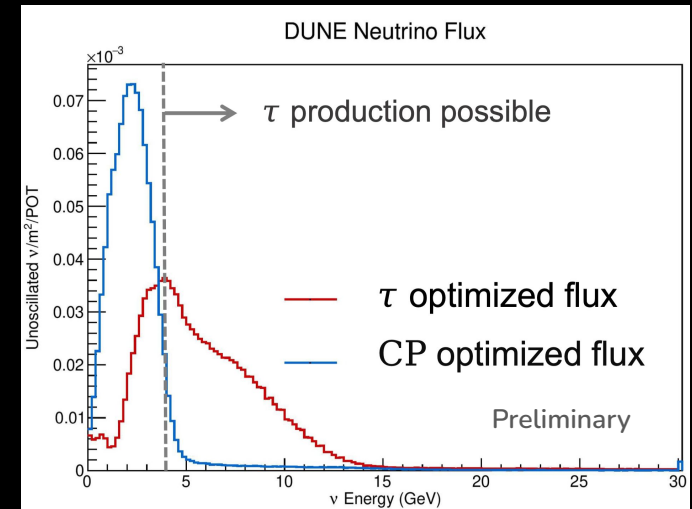
- 62 m deep
- One LArTPC with pixelated readout
- Temporary Muon Spectrometer
- On-axis, magnet and calorimeter for flux monitoring

DUNE Beam Flux

- DUNE's flux peaks between 2-3 GeV being the max. oscillation at 2.5 GeV, which is not ideal for ν_τ
- The far detector (FD) is at a fixed distance from the neutrino production point, the first oscillation maximum is below the ν_τ -CC kinematic thresholds, creating ambiguities between Δm_{31}^2 and $\sin^2\theta_{23}$
- Still, we will have quite a few oscillated tau neutrinos in the high energy tail.



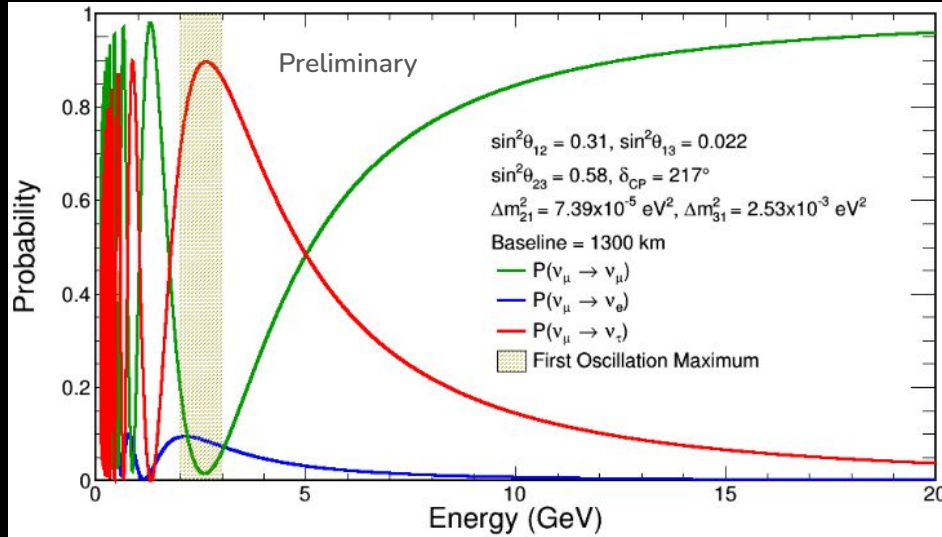
- CP-optimized beam (3 horns configuration)
 - Low energy
 - Default starting configuration
- Tau-optimized beam (2 horns configuration)
 - High energy beam
 - Possible configuration after CP programs has completed



Expected counts/year:

- ~ 30 $\bar{\nu}_\tau$ in CP-optimized anti-neutrino mode
- ~ 130 ν_τ in CP-optimized neutrino mode
- ~ 800 ν_τ in Tau-optimized neutrino mode

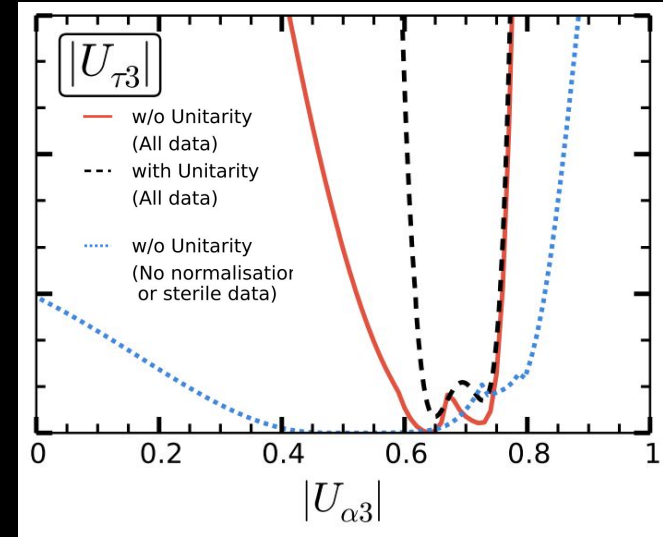
Why Tau Neutrinos? We have the chance to check if our assumption of unitarity (three flavor paradigm) its right or not



At oscillation maximum (atmospherics) the majority of ν_μ oscillate to ν_τ

Almost all the knowledge of ν_τ sector comes from the assumptions:

- Lepton universality for cross-sections
- PMNS unitarity for oscillations



[S. Parke and M. Ross-Lonergan, PRD 93, 1103009 \(2016\)](#)

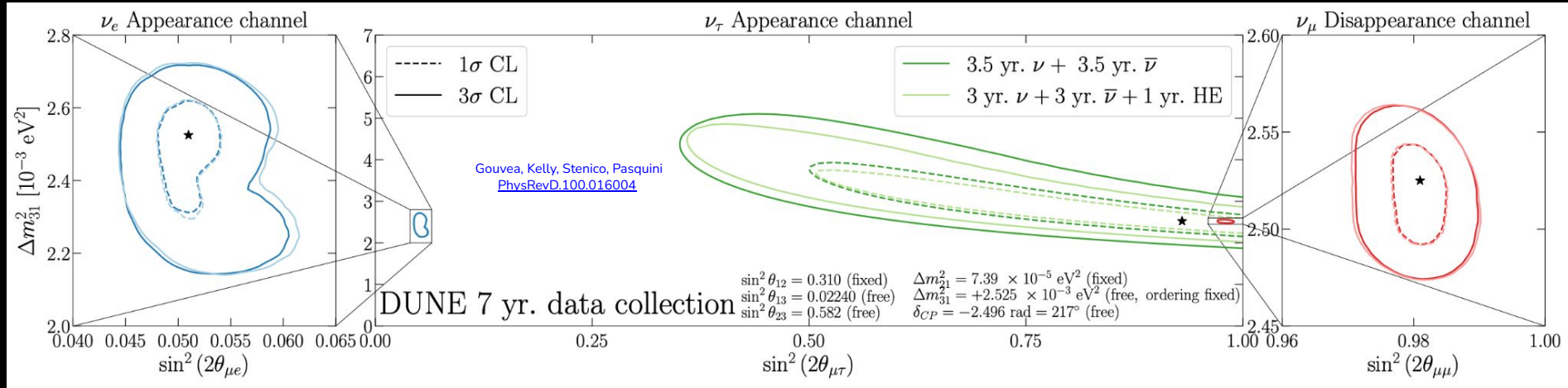
DUNE will be the only experiment able to directly observe tau neutrinos appearing

- High statistics samples both from beam and atmospheric

Why Tau Neutrinos? Model-independent Non-Unitarity

Within the three-neutrino picture, **effective mixing angles** are related, but we can determine the measurement capability of each of the three channels at DUNE. **Consistency check: do the mixing angles sum properly?**

$$\sin^2(2\theta_{\mu e}) + \sin^2(2\theta_{\mu\tau}) = \sin^2(2\theta_{\mu\mu})$$



- **DUNE data alone are expected to constrain the normalization of the 3rd PMNS column to ~ 5%**

Gouvea, Kelly, Stenico, Pasquini [PhysRevD.100.016004](https://arxiv.org/abs/1001.016004)

- **All other neutrino data constrain normalization to ~ 7.5 %**

[S. Parke and M. Ross-Lonergan, PRD 93, 1103009 \(2016\)](https://arxiv.org/abs/1103.0009)

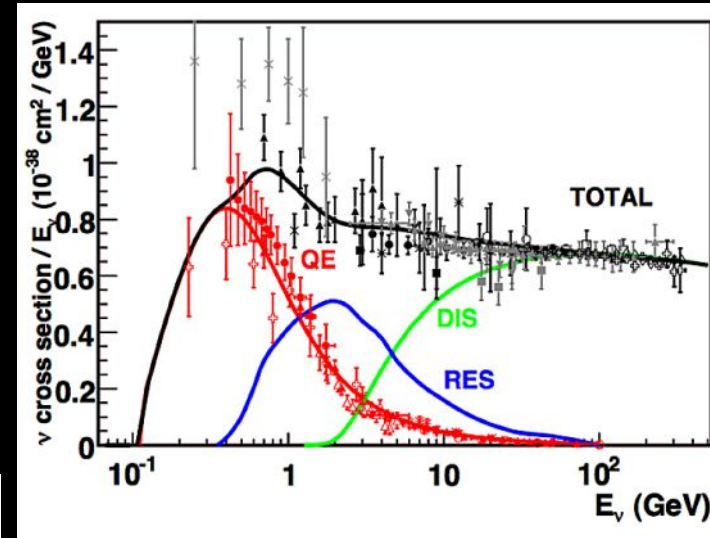
Why Tau Neutrinos? ν_τ (CC) interactions (heavy lepton) give access to cross section physics not accessible otherwise, adding two structure functions (F_4 and F_5) to the **cross section**. [Nucl. Phys. B 84, 467\(1975\)](#)

$$\frac{d^2\sigma^{\nu(\bar{\nu})}}{dxdy} = \frac{G_F^2 M E_\nu}{\pi(1 + Q^2/M_W^2)^2} \left((y^2 x + \frac{m_\tau^2 y}{2E_\nu M}) F_1 + \left[(1 - \frac{m_\tau^2}{4E_\nu^2}) - (1 + \frac{Mx}{2E_\nu}) \right] F_2 \right. \\ \left. \pm \left[xy(1 - \frac{y}{2}) - \frac{m_\tau^2 y}{4E_\nu M} \right] F_3 + \frac{m_\tau^2(m_\tau^2 + Q^2)}{4E_\nu^2 M^2 x} F_4 - \frac{m_\tau^2}{E_\nu M} F_5 \right),$$

[Physical Sciences, Vol 8, issue 1.](#)

- **Neutrino interactions** (cross section) are the major contributor of systematic uncertainties in oscillation measurements (T2k, NOvA).
- E_ν & ν -nucleus interactions relies on **reconstruction techniques** either based on **kinematics** (T2K/HK) or **calorimetric methods** (DUNE/NOvA/SBN) and both requires reliable predictions from **interaction models**.
- Extraction of oscillation parameter is biased by the **interaction model**.

$$P(\nu_\alpha \rightarrow \nu_\beta) = \sin^2 2\theta_{ij} \sin^2 \left(\frac{\Delta m_{ij}^2 L}{4 E_\nu} \right)$$

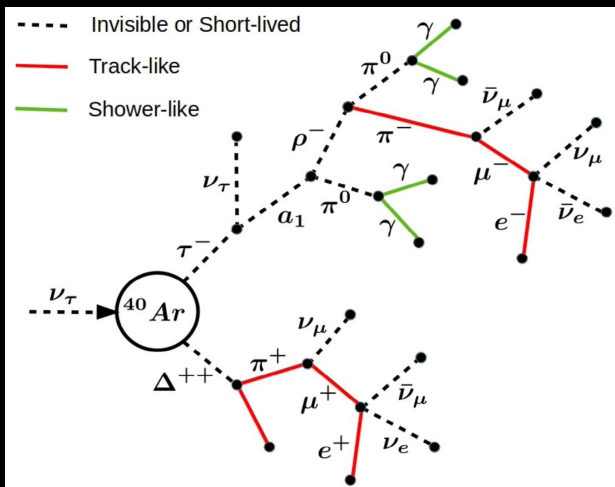
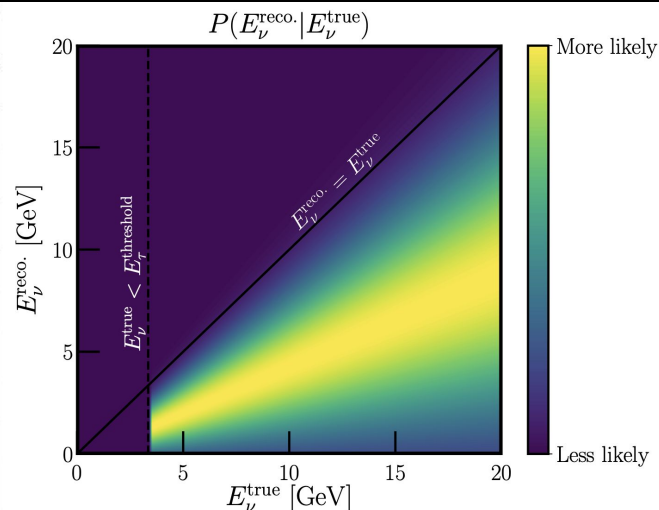


Kinematical changes in Q^2 and E_l due to the presence of m_τ

Challenges

- Tau leptons have many decay modes
- CC-(ν_e, ν_μ) or NC events have same particle content
 - Angular correlations due to missing neutrino(s) from τ decay is the key signature
- Hadronic modes can be complicated
- Difficult to separate hadronic systems
- from τ decay and nucleus

Decay mode	Branching ratio
Leptonic	35.2%
$e^- \bar{\nu}_e \nu_\tau$	17.8%
$\mu^- \bar{\nu}_\mu \nu_\tau$	17.4%
Hadronic	64.8%
$\pi^- \pi^0 \nu_\tau$	25.5%
$\pi^- \nu_\tau$	10.8%
$\pi^- \pi^0 \pi^0 \nu_\tau$	9.3%
$\pi^- \pi^- \pi^+ \nu_\tau$	9.0%
$\pi^- \pi^- \pi^+ \pi^0 \nu_\tau$	4.5%
other	5.7%



Tau decay length $\sim 87 \mu\text{m}$
 ^{40}Ar nuclear radius, $\sim 3.4 \text{ fm}$

Tau decay products aren't subject to the ^{40}Ar nuclear potential

Tau lifetime $(2.903 \pm 0.005) \times 10^{-13} \text{ s}$
 Mass: $1.7 \text{ GeV}/c^2$

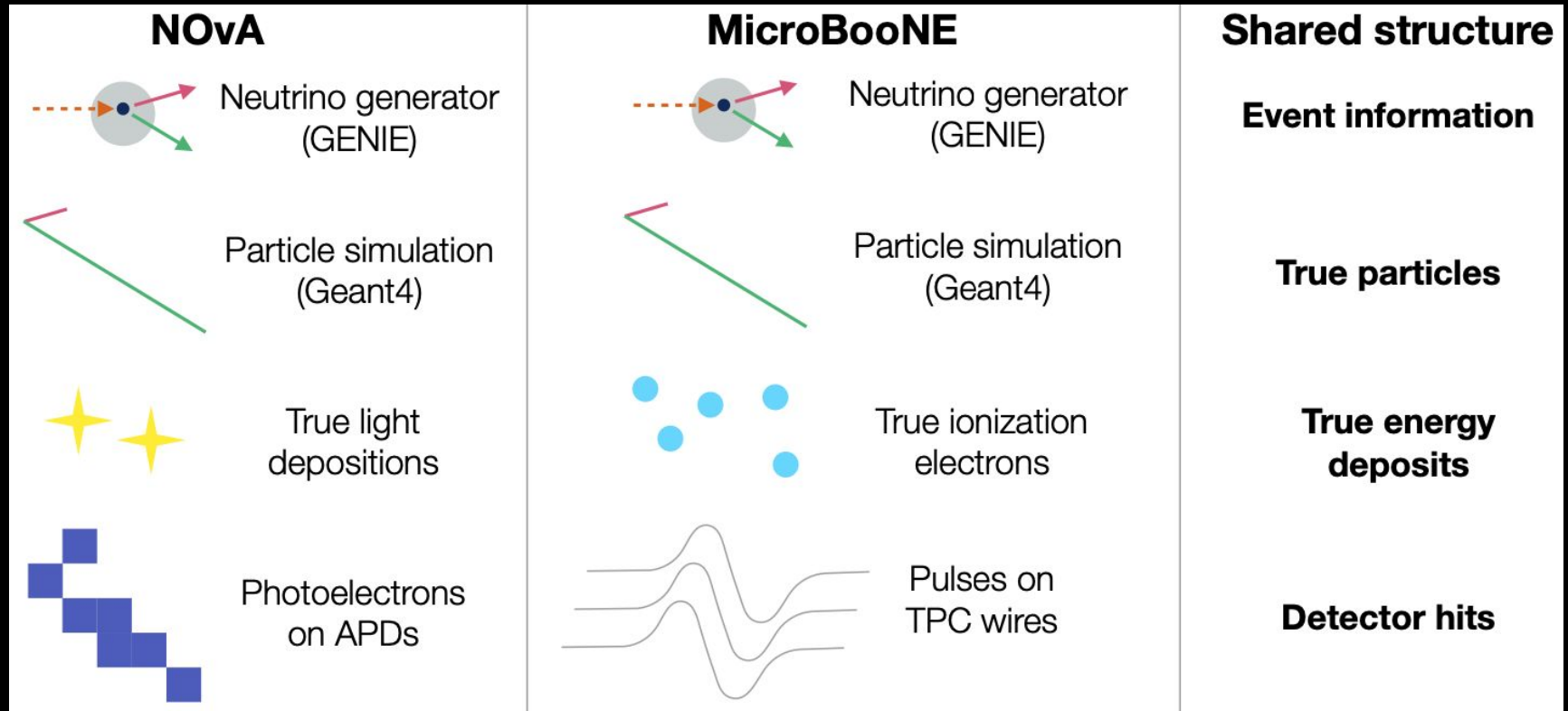
Tau doesn't lead to observables displaced vertices

DUNE granularity is limited by a wire spacing of 3mm

Observation of Tau tracks is unlikely!

Graph Neural Networks

The idea of using a GNN started with finding a solution to a common paradigm at a high level

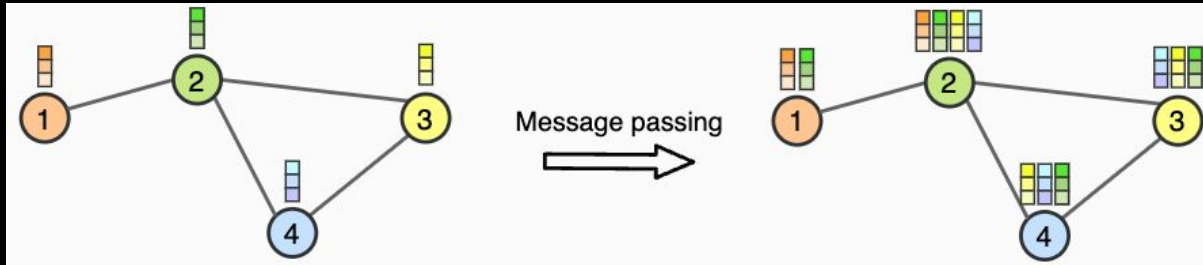
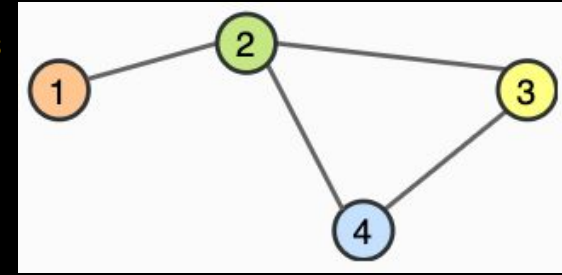


What is a graph?

1) A graph is defined as a tuple of set of **nodes/vertices** and a set of **edges/links**

2) **Each edge is a pair of two vertices** & represents a connection between them

Graphs are excellent in dealing with complex problems with relationships and interactions. They are used in pattern recognition, social networks analysis, recommendation systems, and semantic analysis



- **Each node creates a feature vector** that represents the message to be send to all its neighbors.
- **The node receives one message per adjacent node**

Graphs are an ideal structure for understanding physics data

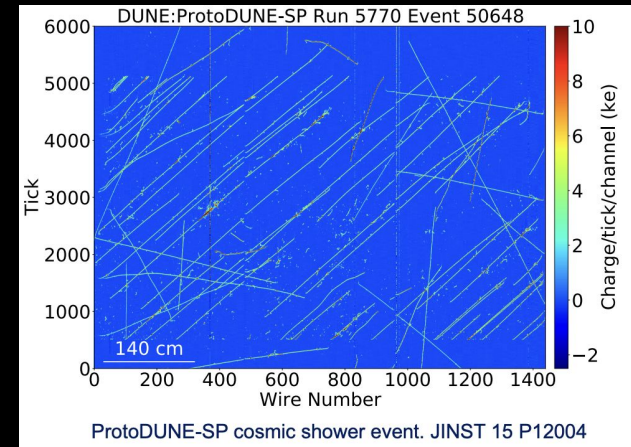
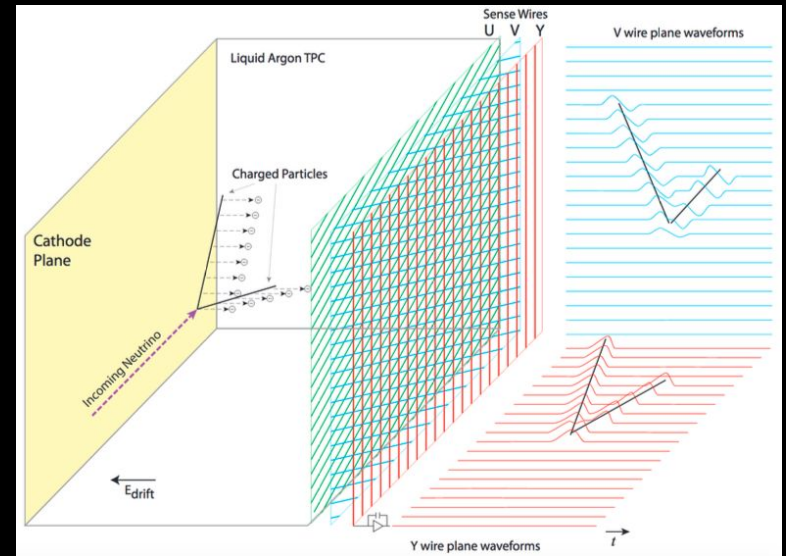
- Naturally sparse
- Hits have a causal structure that can easily be modeled by edges
- Accommodates relationships beyond nearest neighbor

Liquid Argon Time Projection Chambers (LArTPC)

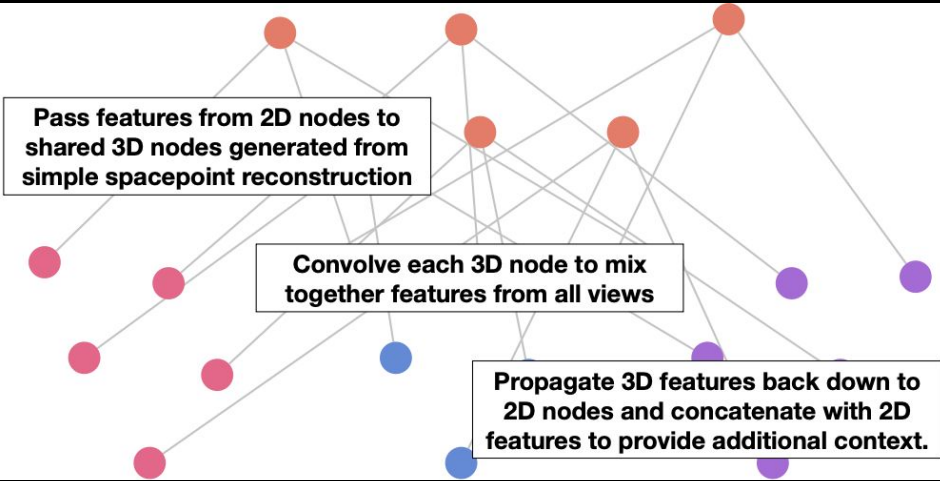
- LArTPC are currently heavily used in neutrino physics
 - Today: ICARUS, MicroBooNE, SBND, ProtoDune
 - Future: DUNE (70 kT far detector deep underground)
- Charged particles ionize liquid argon as they travel
- Ionization electrons drift due high potential between cathode and anode planes
- Closely spaced wires ($\sim 3\text{mm}$) at anode provide high-resolution image of neutrino interaction
- Multiple wire planes provide 3D information

We would like to:

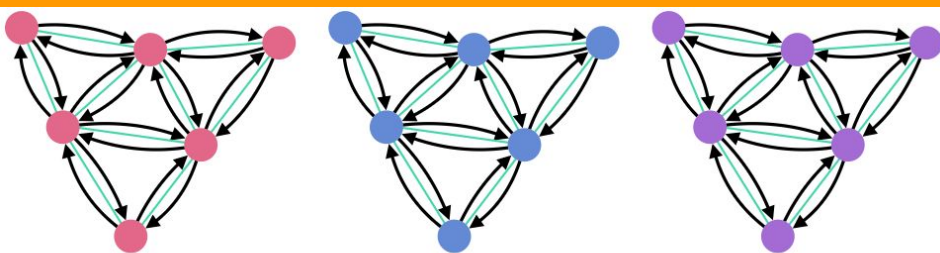
- Cluster hits into objects
- Classify objects according to the particle that created it
- Assemble the objects into an event
- Determine type and kinematic properties of the event



NuGraph: originally designed for identifying jets in hadron interactions ([Exa.TrkX is a collaboration - U.Cincinnati](#)). **Primary goal is to classify each detector hit according to particle type.**



3D graph nodes/convolutions: perform message-passing independently in each detector view.



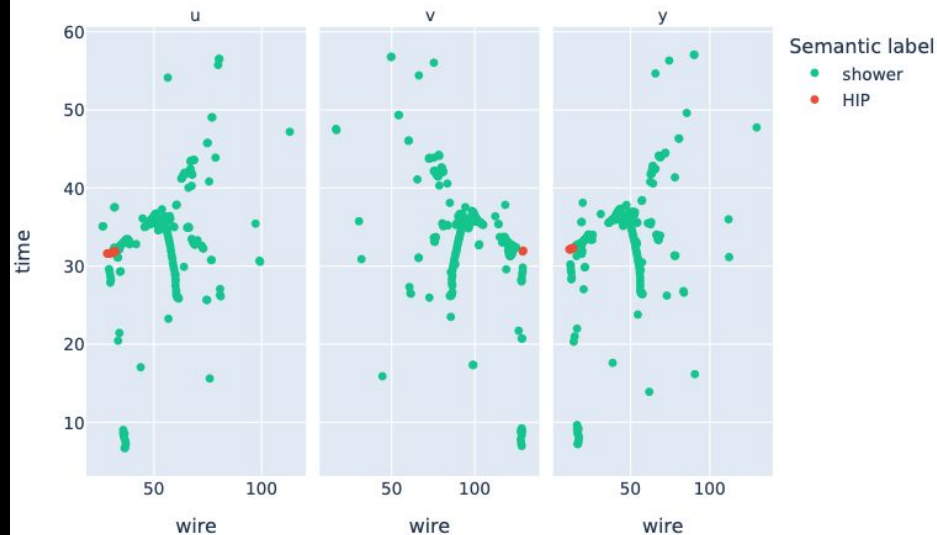
- All of this is repeated through many iterations, allowing key information to propagate through the graph and use these key features for training (energy depositions, 3D position, etc)
- There are two phases for the network: a planer encoder and then a nexus block
- This allows messages to be passed through the nexus block and context to be shared via the planer encoder

Originally tested with MicroBoone open data set with an excellent performance at semantic segmentation:

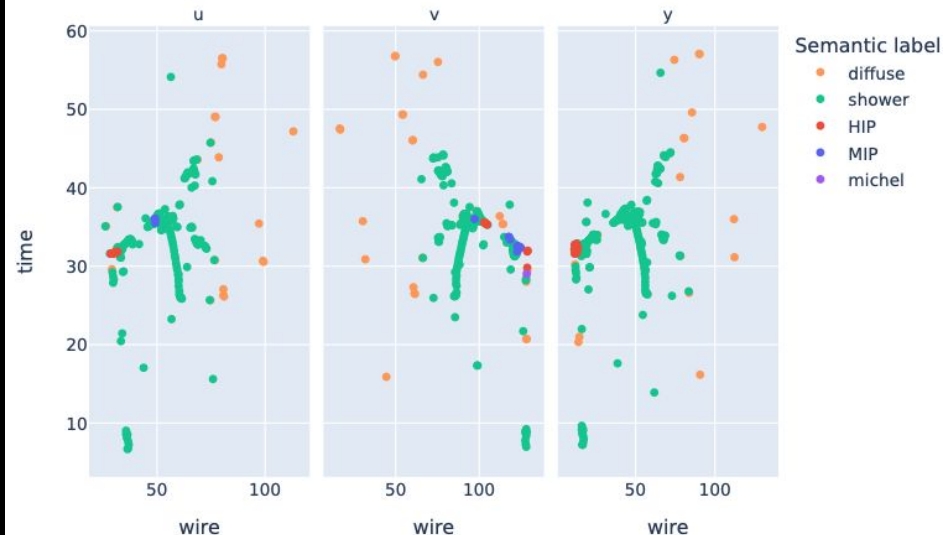
Overall efficiency and purity ~ 95%
Consistency between planes ~98%

NuGraph Output: each of the particle category has a separate set of embedded features which are convolved independently

True semantic labels



Predicted semantic labels



We have five semantic types:

Shower

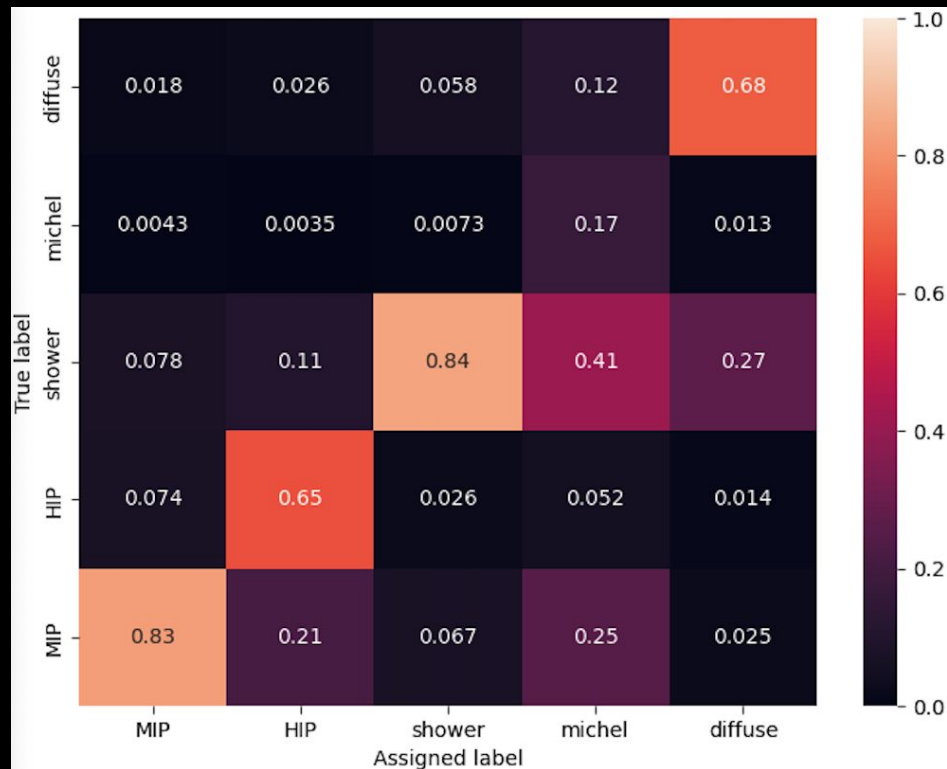
HIP: highly ionizing particle

MIP: minimum ionizing particle

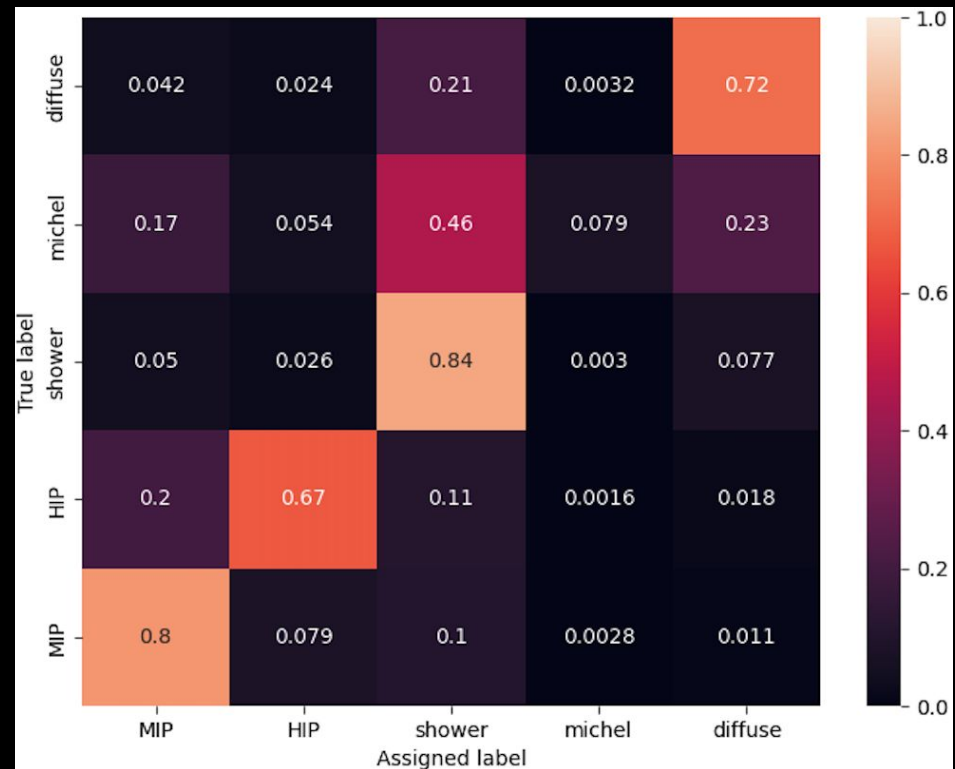
Michele

Diffuse: any small EM activity (compton scatters, etc) + anything that creates small blips (neutrons)

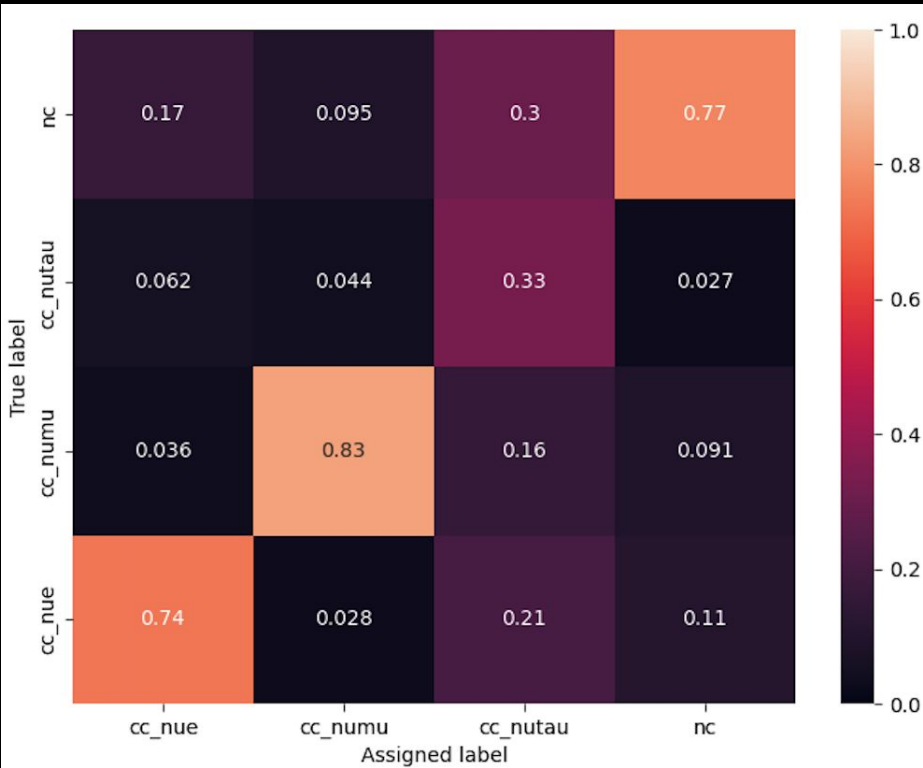
Confusion matrix - **predicted semantic label to show purity (precision/prediction)**



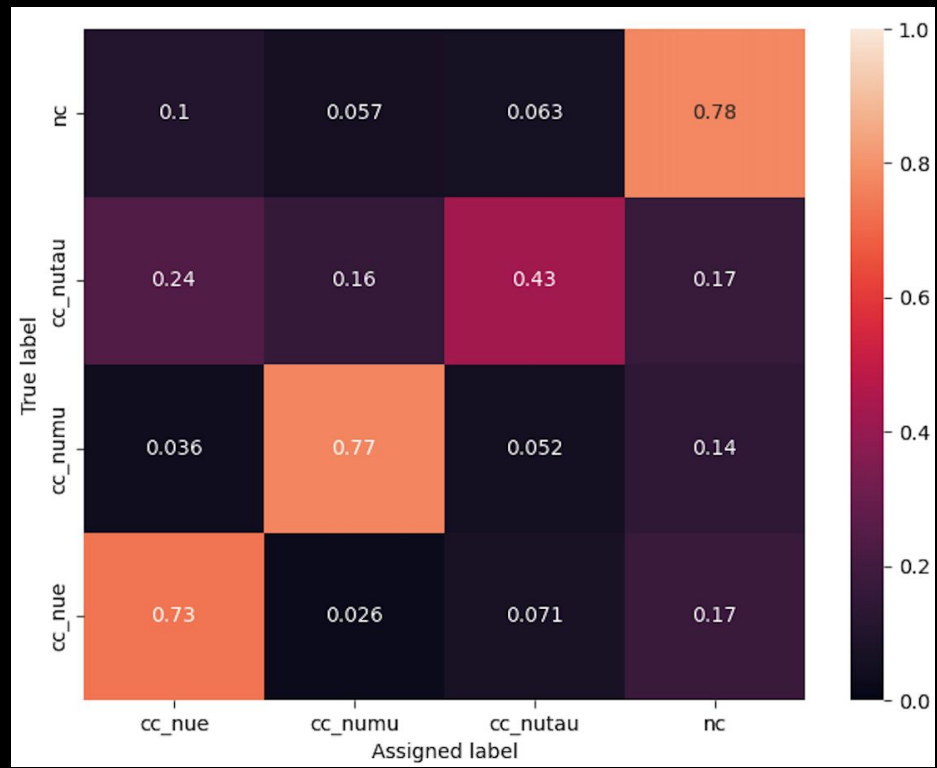
Confusion matrix - **true semantic label to show efficiency (recall/sensitivity)**



Confusion matrix - **predicted event label** to show purity (precision/prediction)



Confusion matrix - **true event label** to show efficiency (recall/sensitivity)



NuGraph it's not doing a great job distinguishing cc_nutau and nc and even gets a decent number of cc_nutau confused with cc_nue

Next steps

- **We need to change the codebase for our training and adapt it to DUNE geometry, get the center of each TPC module & drift directions of each module + decoder (done this week!)**
- NuGraph was originally made for MicroBoone (single TPC module) and DUNE has multiple modules.
- **Add decay channels information**

Conclusions

- DUNE is uniquely capable of providing a high-purity & high-statistics sample of beam and atmospheric ν_τ
- ν_τ are challenging to select and reconstruct, but they provide a needed independent check of the three-flavor model
- Atmospheric ν_τ provide a high-purity window into 1st atmospheric oscillation maximum
- **NuGraph** efficiently reject background detector hits & classify those by particle type. **Currently working on vertexing and event classification.**
- Looking forward for a next generation of **NuGraph** able to reconstruct complex interactions like high energy **tau neutrino interactions.**

[NuGraph public repo](#)

THANK YOU!

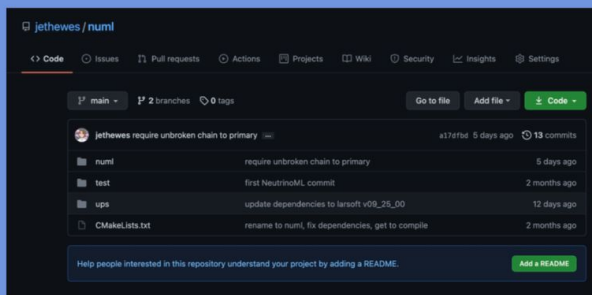
Backup

NuGraph: is a state-of-the-art graph neural network for semantically labelling detector hits in neutrino physics experiments.

NuML: toolkit for writing physics event records to an HDF5 file format.

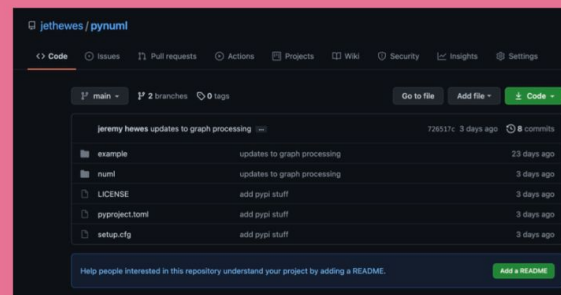
PyNuML: Provide a generic, accessible, efficient and flexible solution for many of the necessary tasks in leveraging ML for particle physics

<https://github.com/vhewes/numl>



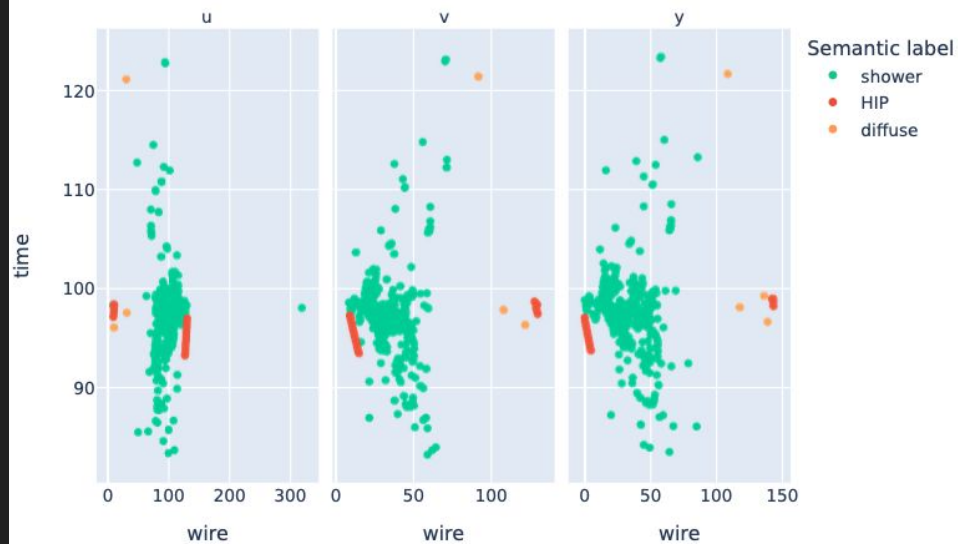
- Hold low-level information such as simulated particles, hits, true energy depositions etc.
- Generic data structure can be shared across experiments

<https://github.com/vhewes/pynuml>

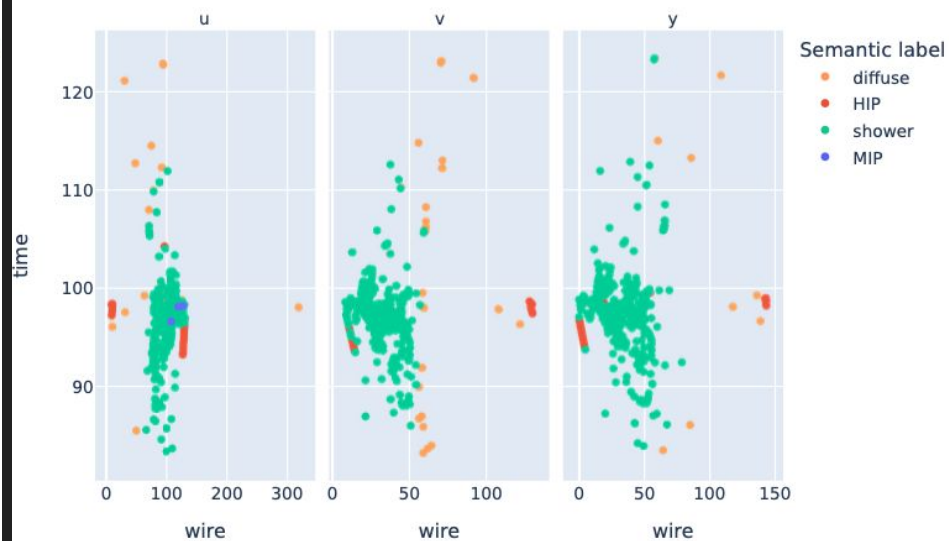


- Define particle ground truth labels for Geant4-simulated particles
- Arrange detector hits into ML objects, ie. graphs, CNN pixel maps, etc

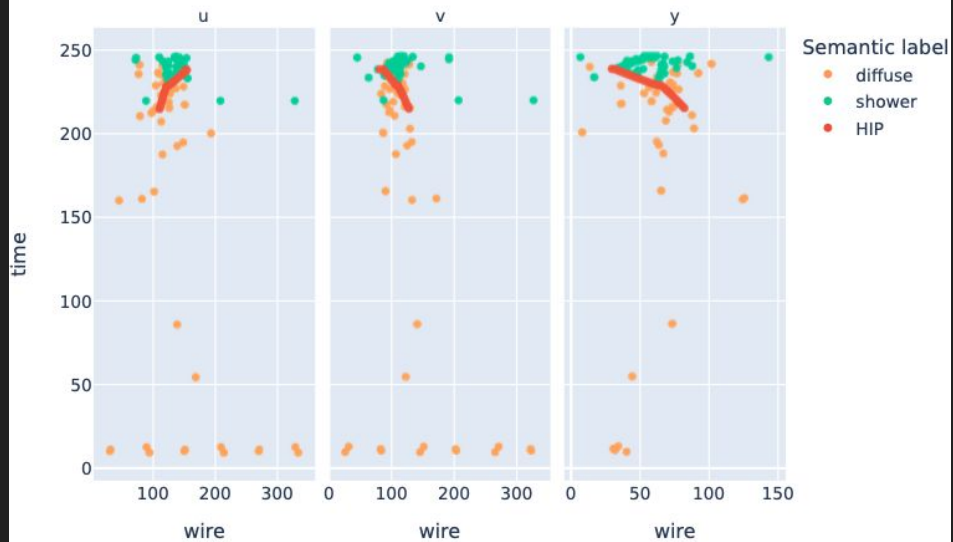
True semantic labels



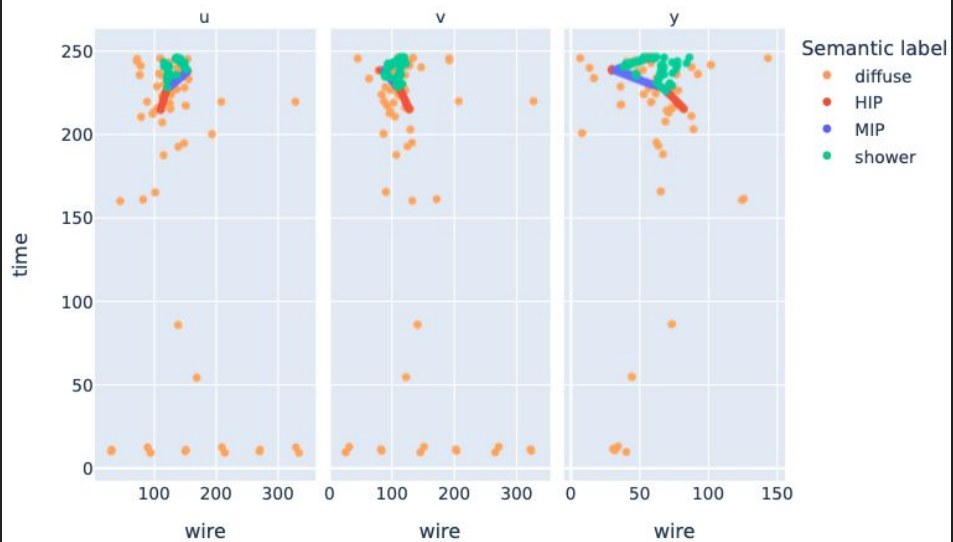
Predicted semantic labels



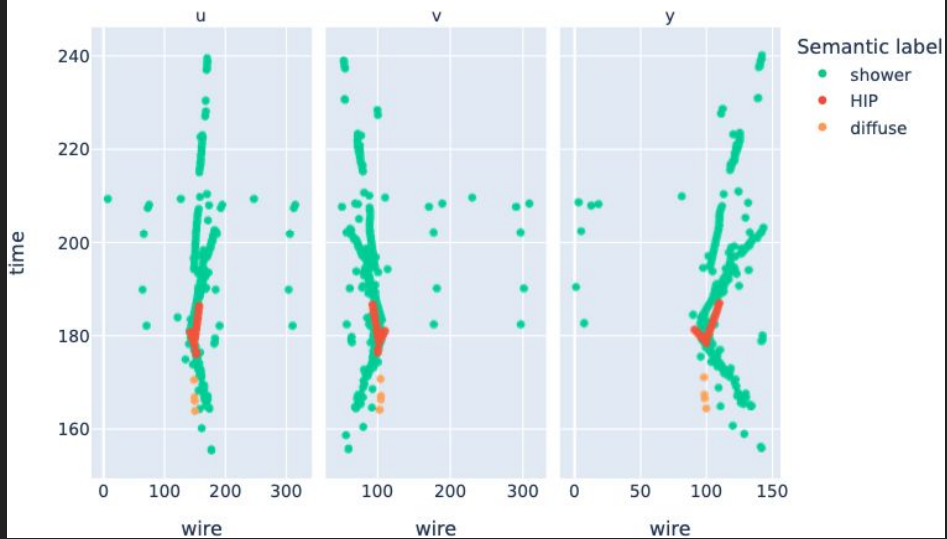
True semantic labels



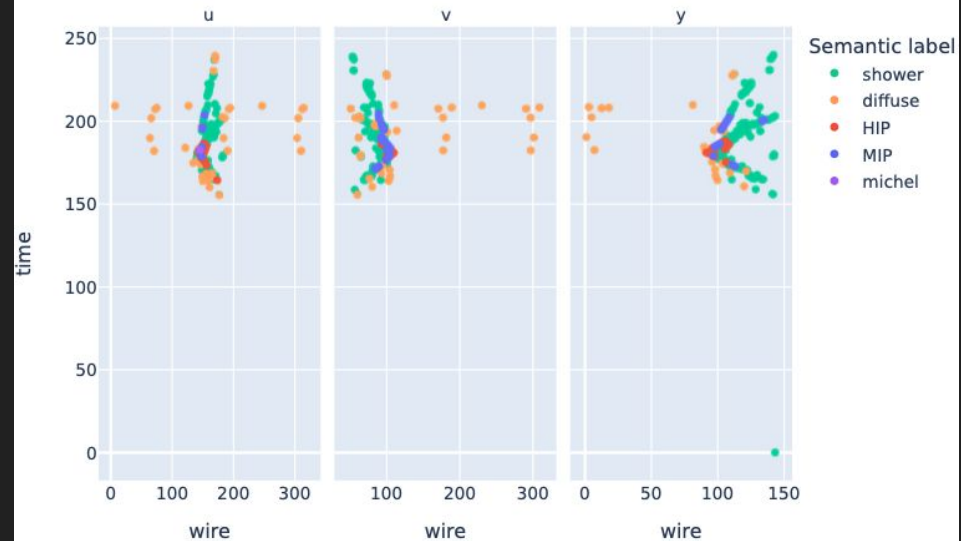
Predicted semantic labels

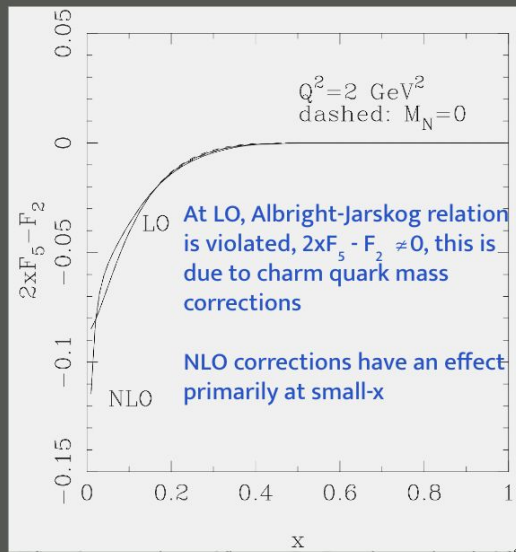
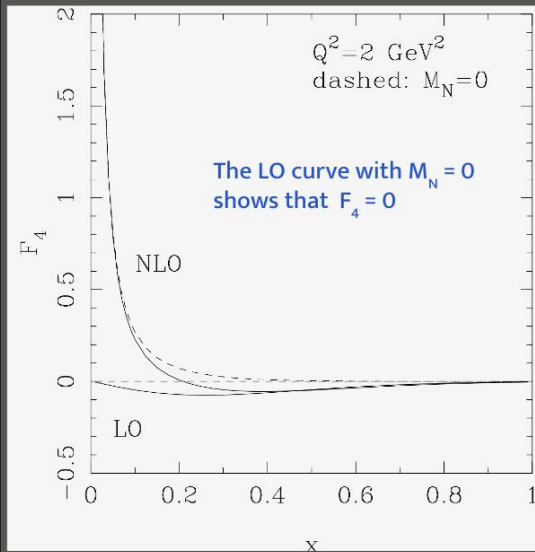


True semantic labels



Predicted semantic labels





Structure functions

The Callan-Gross relations:

$$2xF_1 = F_2$$

$$-xF_3 = F_2$$

At LO, in the limit of massless quarks & target hadrons, Albright-Jarlskog pointed:

$$2xF_5 - F_2 = 0$$

$F_4 = 0$, also holds when the nucleon target is replaced by a lepton target.

$$F_{1N}(x) = W_{1N}(\nu, Q^2)$$

$$F_{2N}(x) = \frac{Q^2}{2xM_N^2} W_{2N}(\nu, Q^2)$$

$$F_{3N}(x) = \frac{Q^2}{xM_N^2} W_{3N}(\nu, Q^2)$$

$$F_{4N}(x) = \frac{Q^2}{2M_N^2} W_{4N}(\nu, Q^2)$$

$$F_{5N}(x) = \frac{Q^2}{2xM_N^2} W_{5N}(\nu, Q^2)$$

- At NLO, $F_4 \sim 1\%$ of F_5 , AJ relations are good approximations to the NLO result, arXiv:hep-ph/0605295.
- Both of the figures show that in evaluations of the total charged current cross section, the naive AJ relations are good approximations to the NLO results. This is true at low energies, where ν_τ cross-section does not probe small-x and at high energies where F_4, F_5 are suppressed, anyway.