



Co-funded by
the European Union

Classification of ESSνSB WC Near Detector Events Using Graph Neural Networks

ν_μ - and ν_e -events for neutrino oscillation studies

Kaare Endrup Iversen - Lund University



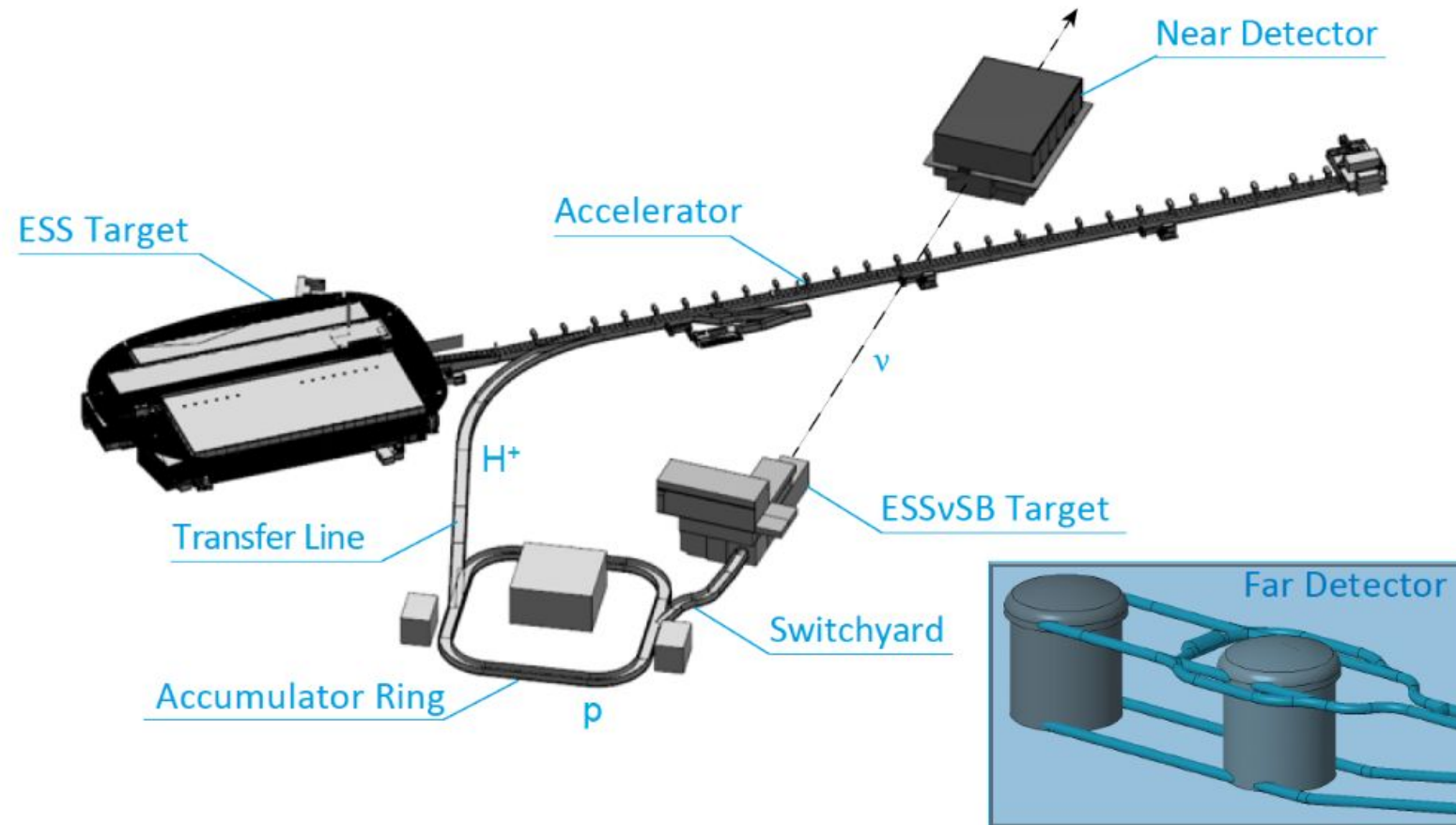
GNN Classification of ESS ν SB events

- ESS ν SB - An accelerator neutrino experiment in Sweden
- Current framework and motivation
- Graph neural networks and implementation
- Classification performance
- Generalisation

Motivation

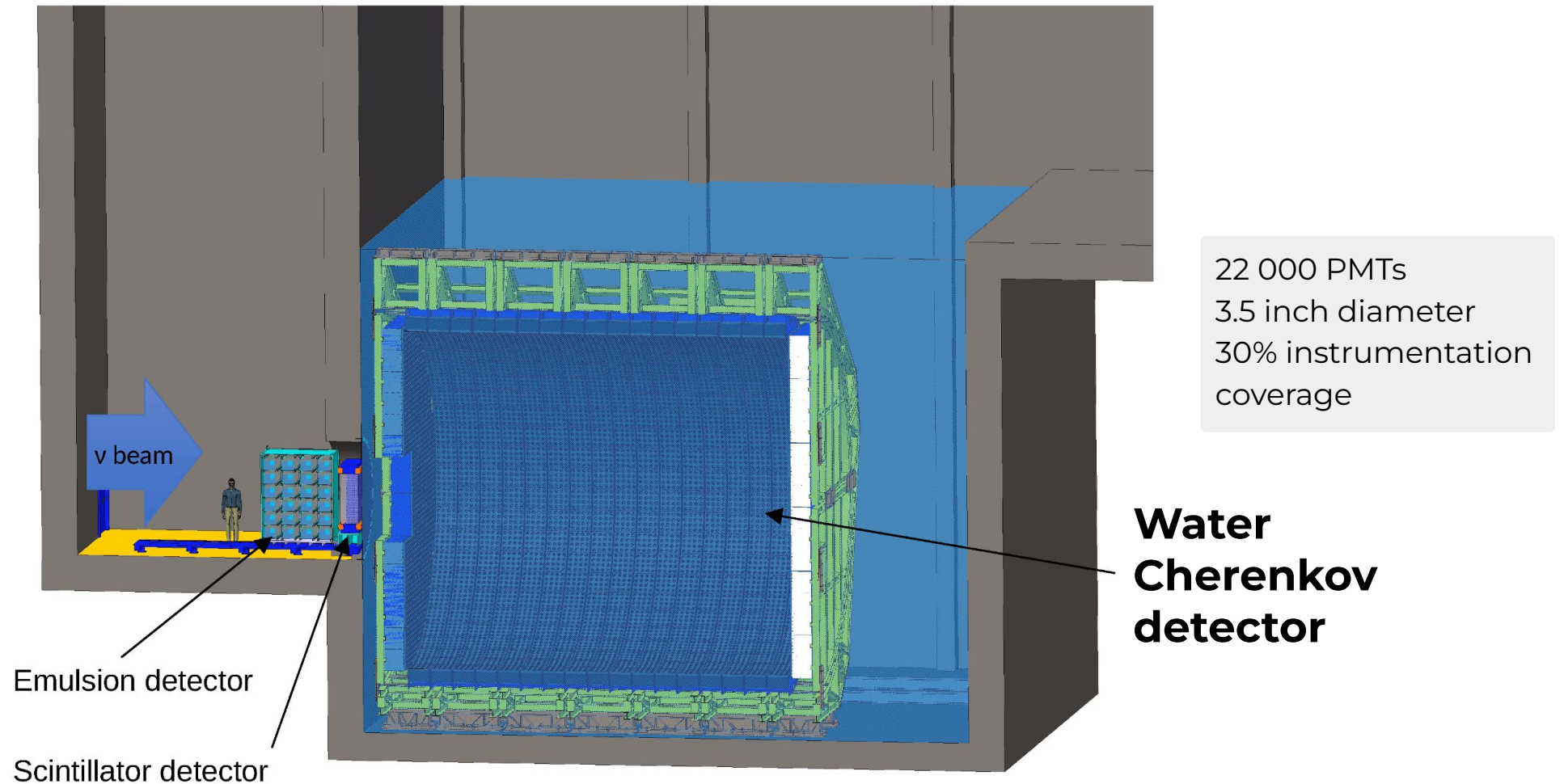


ESSnuSB - Detectors



[The European Spallation Source neutrino Super Beam Conceptual Design Report](#)

ESSnuSB - Near Detectors



The European Spallation Source neutrino Super Beam Conceptual Design Report

Current Framework

Charged Lepton Simulations

WCSIM
<https://github.com/WCSim/WCSim>

Neutrino Interaction Simulations

GENIE
Generator. Nucl. Instrum. Meth. A 614:87–104, 2010

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LLH Based Reconstruction

fiTQun
J. Phys.: Conf. Ser. 888 012066, 2017

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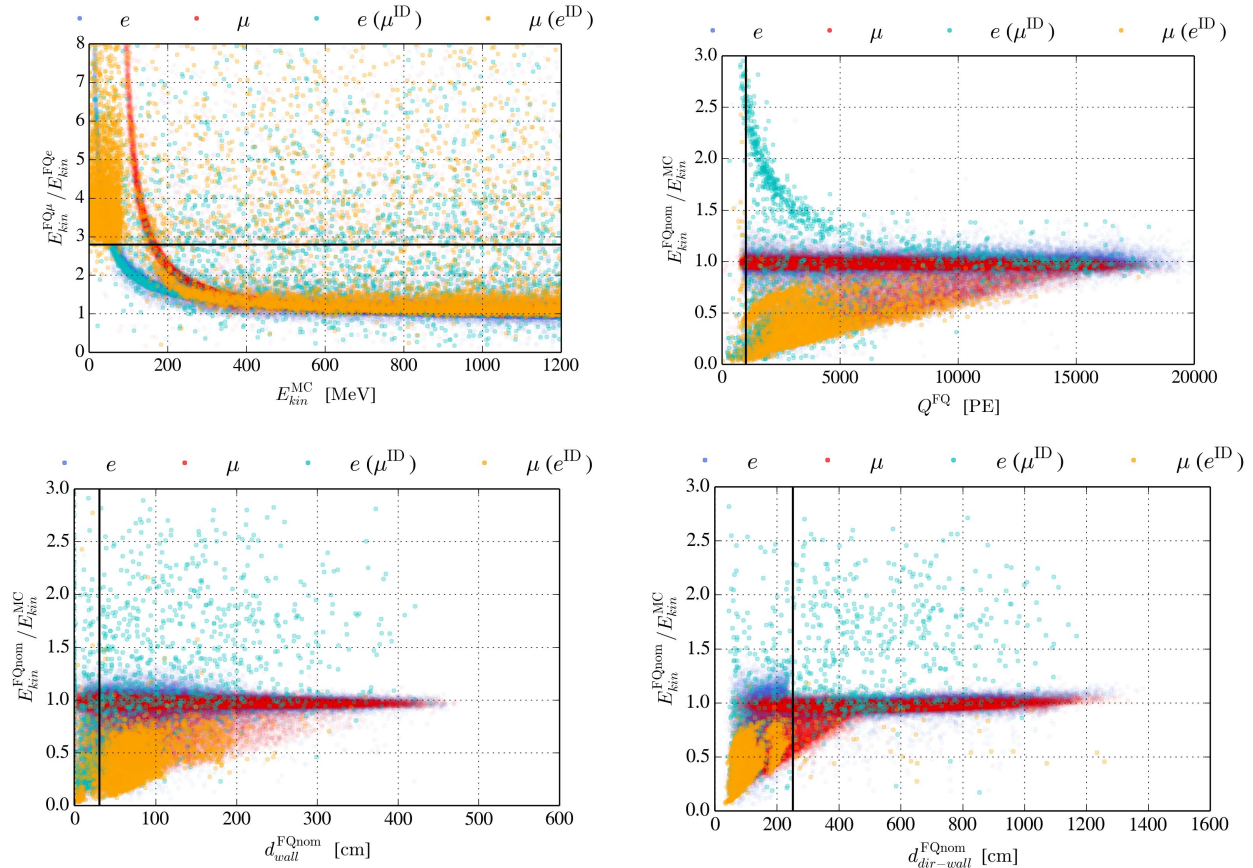
- Likelihood reconstruction takes ~1 CPU min/event
- To explore different detector proposals, fast reconstruction is crucial

Why Do We Need GNN Reconstruction?

- Fast and reliable event reconstruction enables testing of different detector layouts
- LLH-based methods are accurate, but reconstruction is slow
- Multiple reconstruction methods provides a way to cross check and find systematic errors

Data processing

- Cuts based on **reconstructed variables**
- Removes events that are **hard to classify**
- Reduces number of events by a **factor ~2**

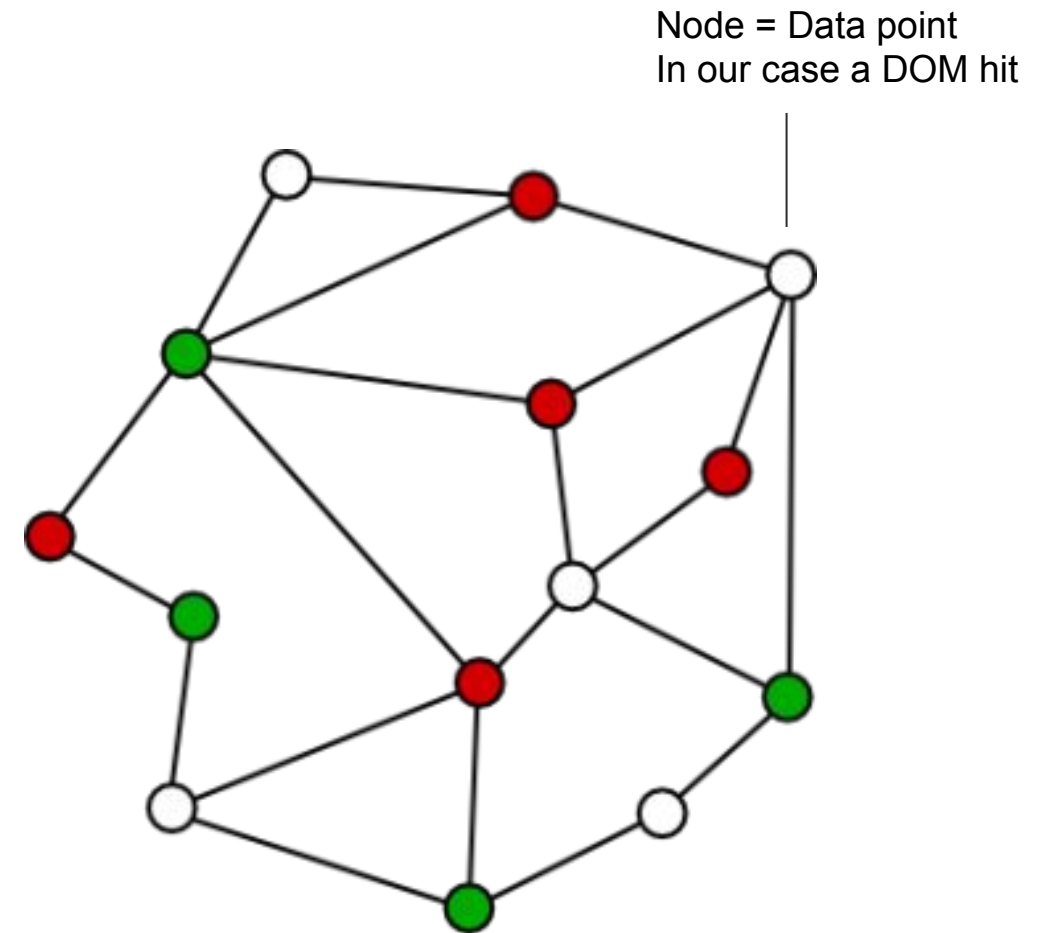


The European Spallation Source neutrino Super Beam Conceptual Design Report

Graph Neural Networks

Graph Neural Networks

- Based on graph theory
- Each **graph** is a **neutrino event**
- Each **PMT hit** is a **node**
- A node features: **xyz, time, charge**
- Suited for non-euclidian data



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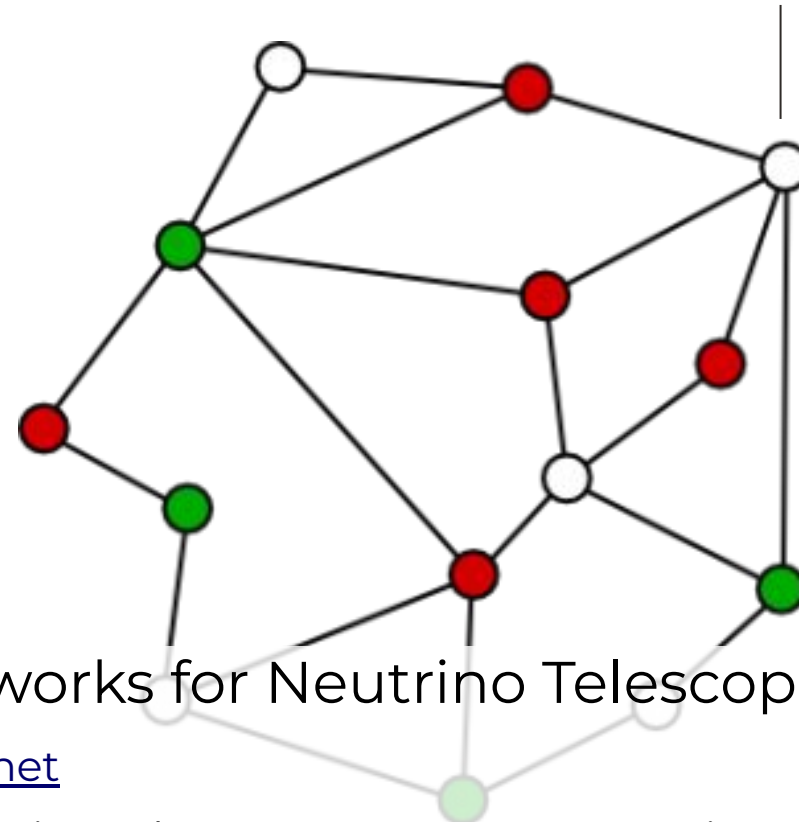


GraphNeT - Graph Neural Networks for Neutrino Telescopes

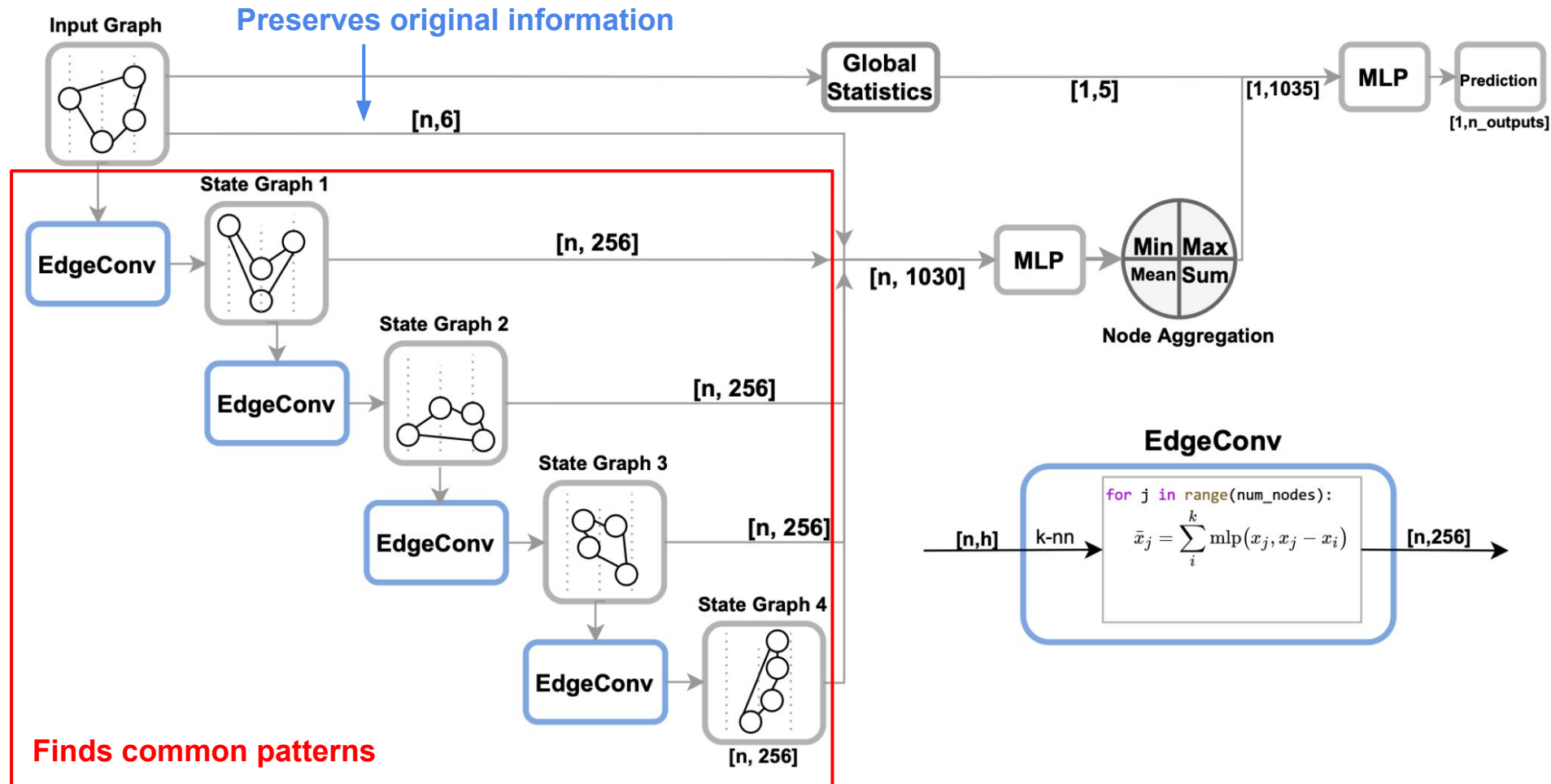
<https://github.com/graphnet-team/graphnet>

GraphNeT: Graph neural networks for neutrino telescope event reconstruction
(Søgaard et al)

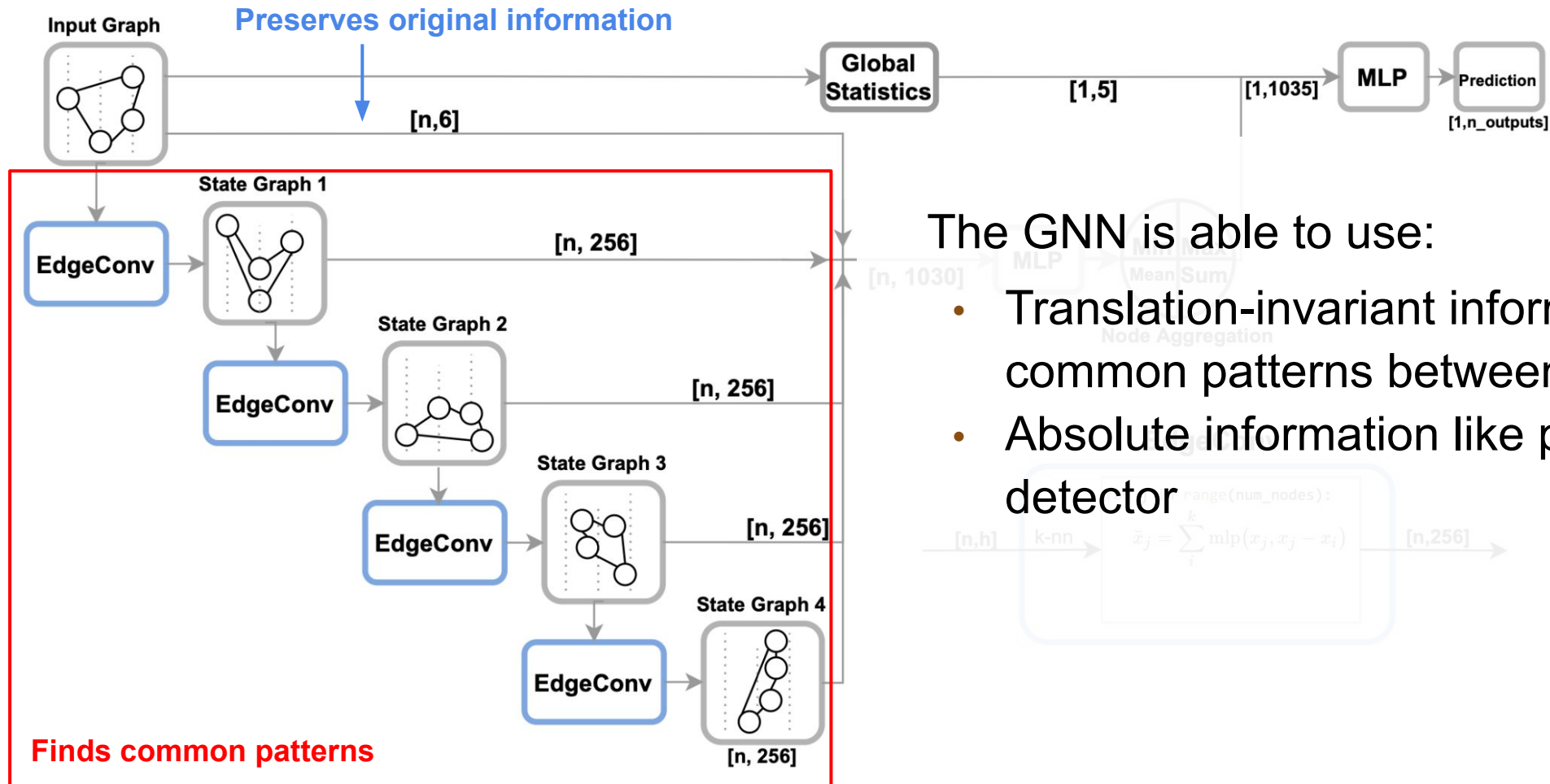
Node = Data point
In our case a DOM hit



Graph Neural Networks - Architecture



Graph Neural Networks - Architecture



The GNN is able to use:

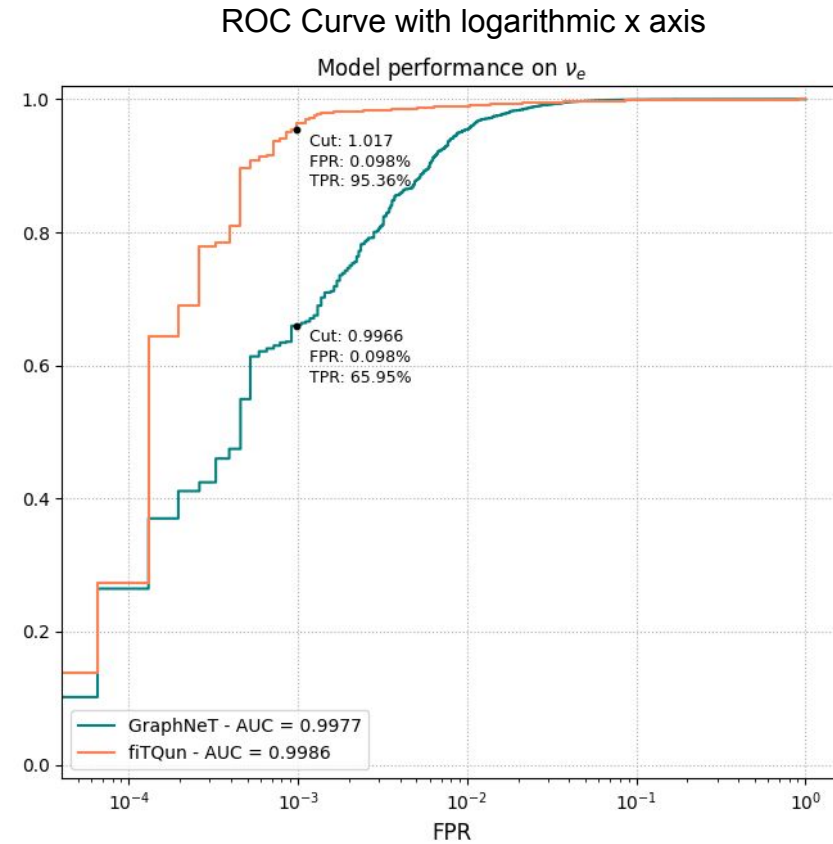
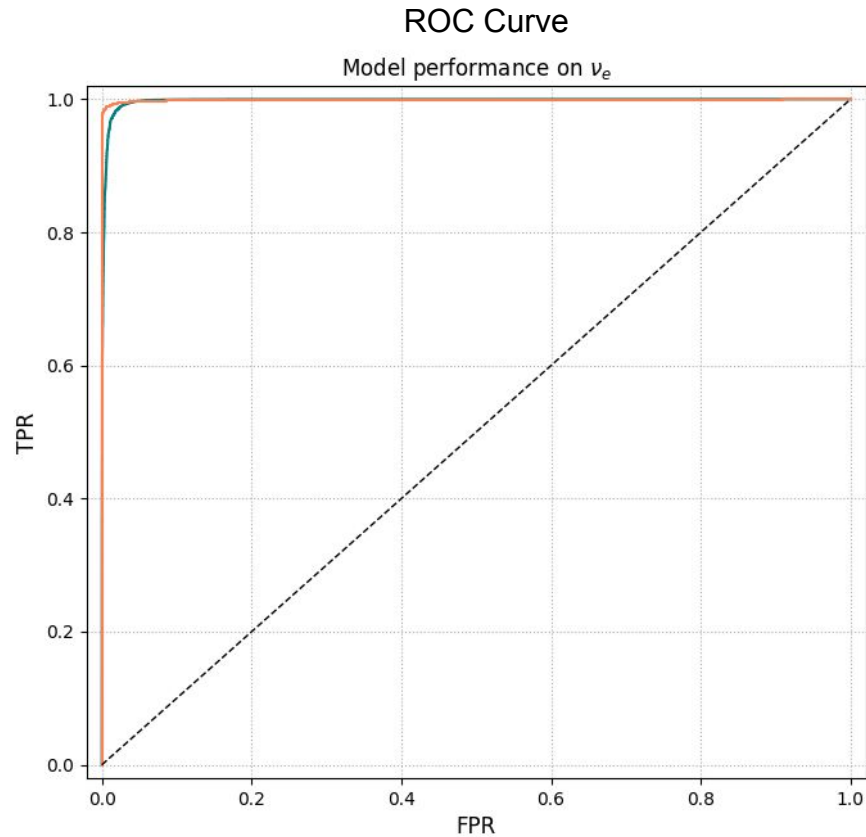
- Translation-invariant information about common patterns between events
- Absolute information like position in the detector

$$\bar{x}_j = \sum_i^k \text{mlp}(x_j, x_j - x_i)$$

Classification Performance

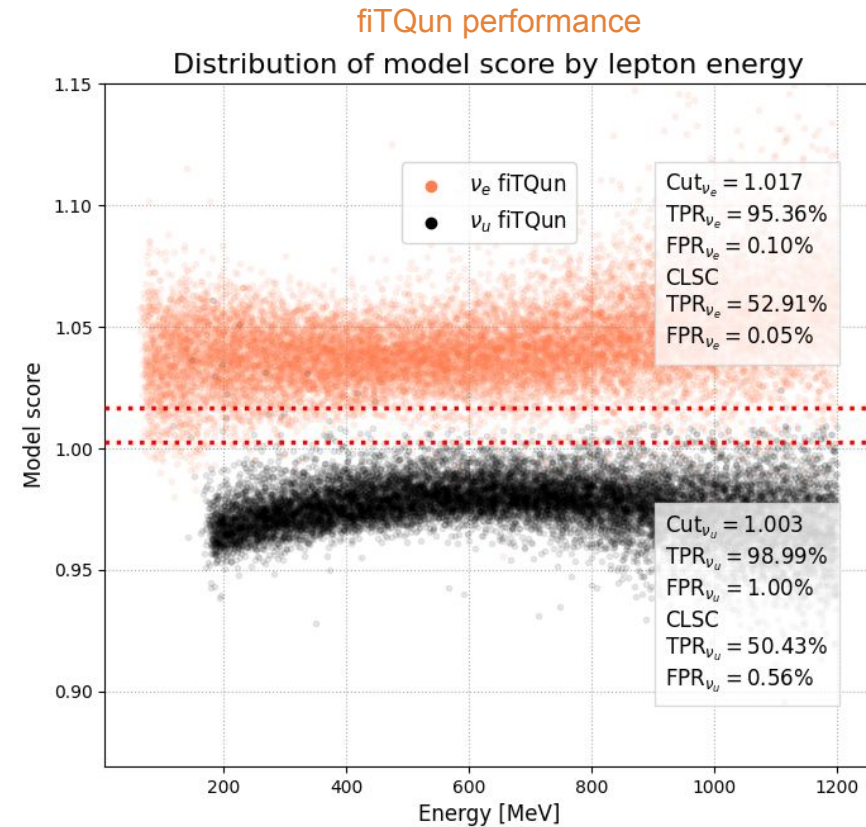
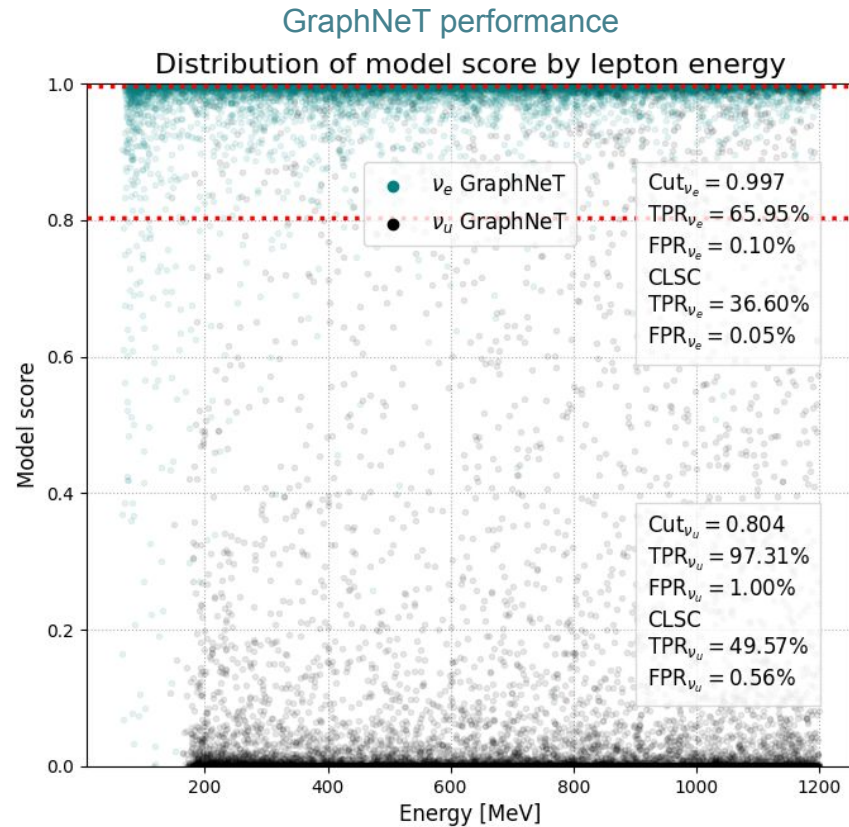
Event Classification Performance

With cuts



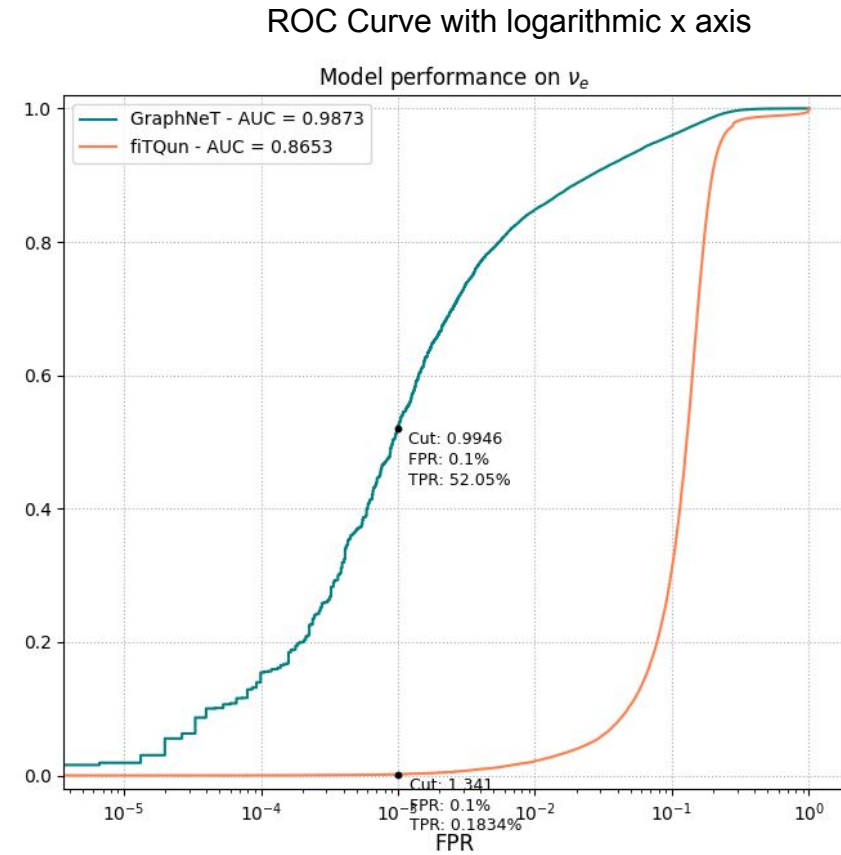
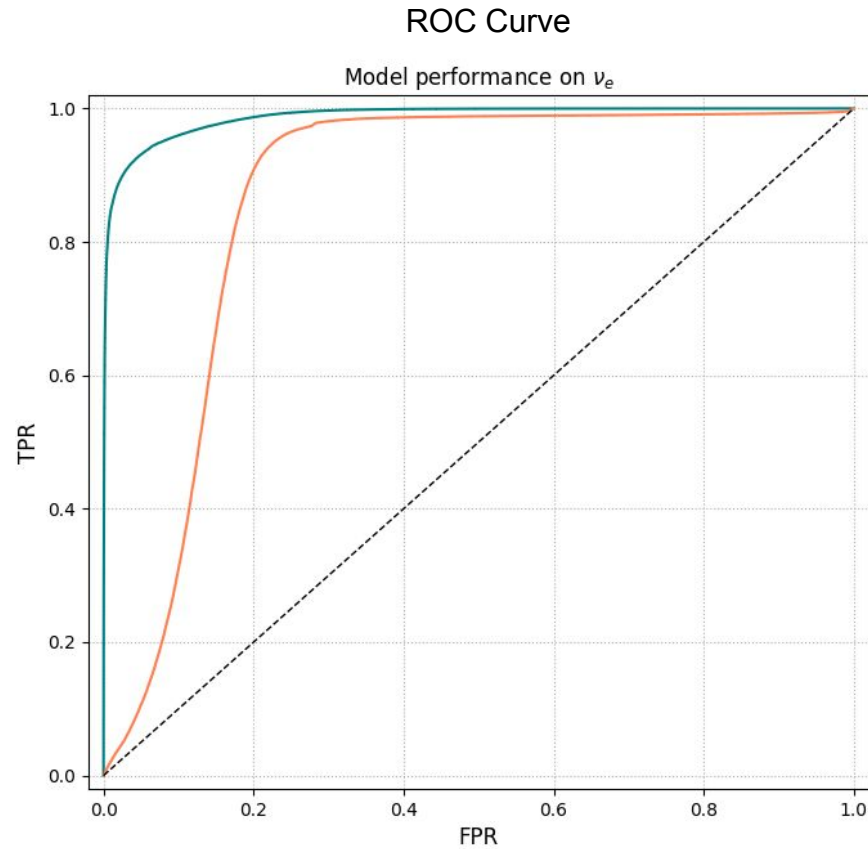
Event Classification Performance

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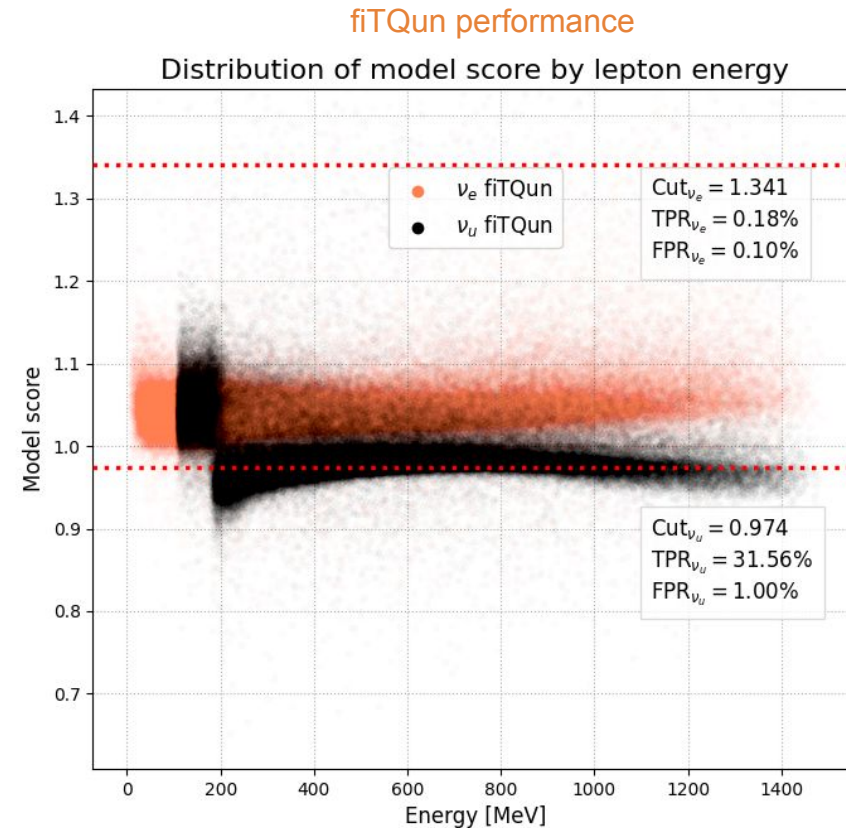
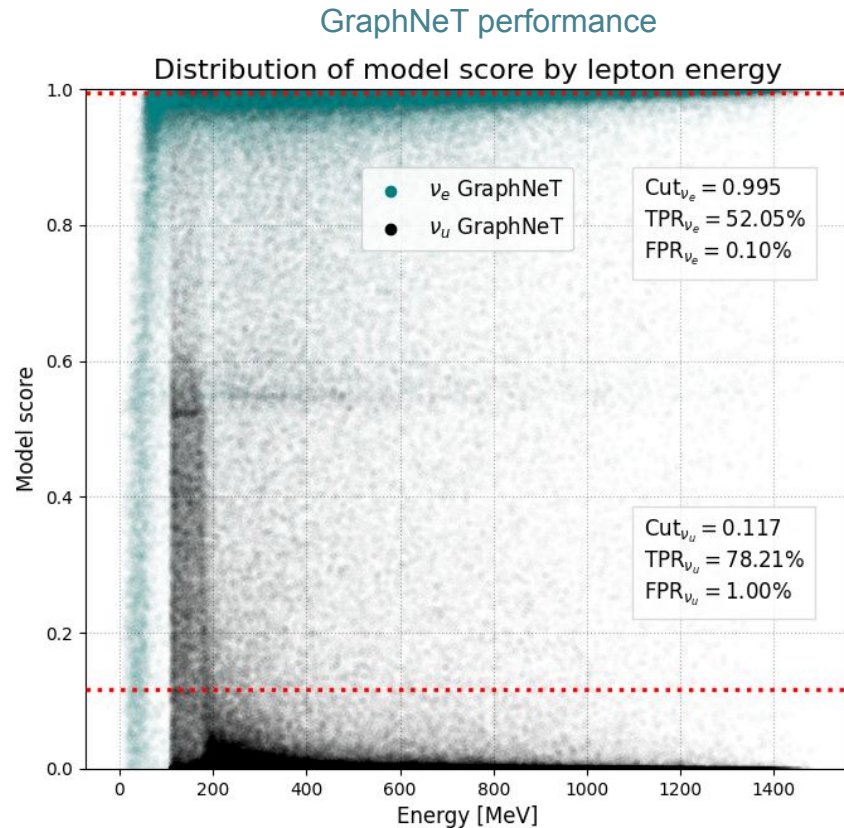
Event Classification Performance

Without data cuts



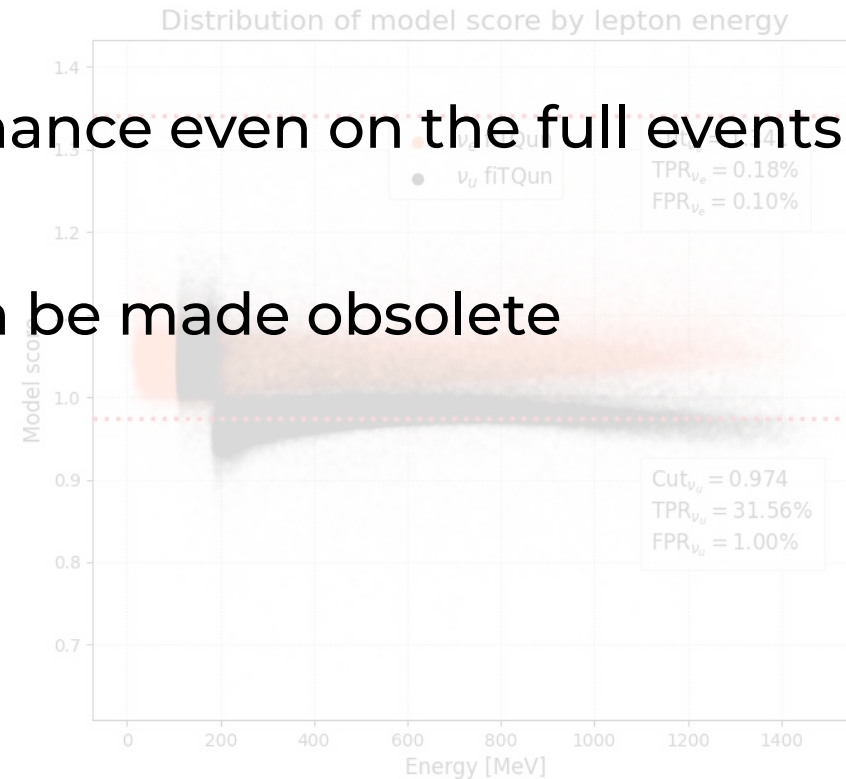
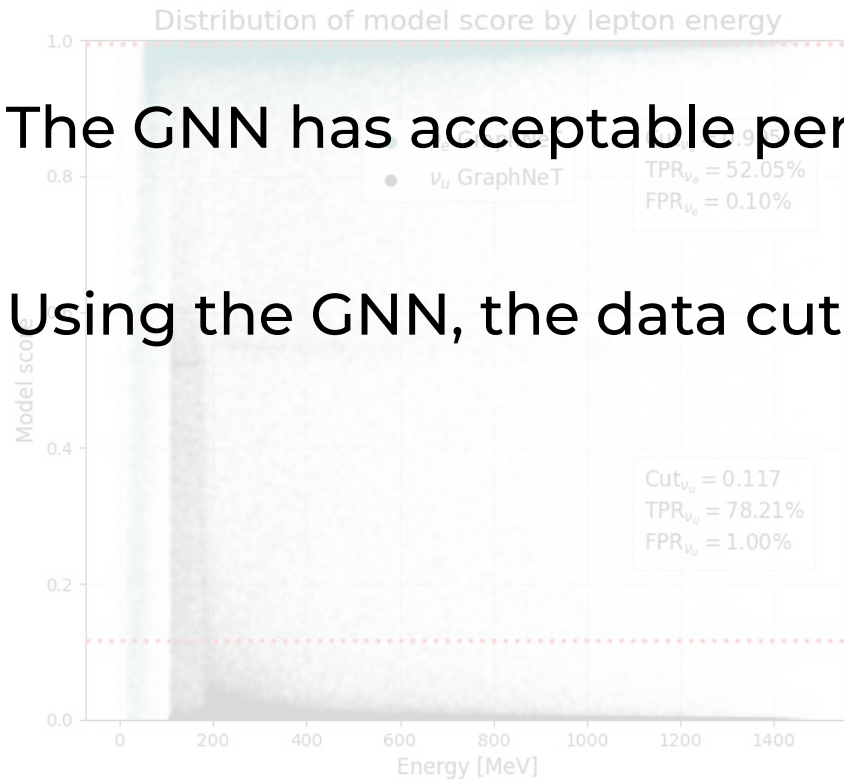
Event Classification Performance

Without data cuts



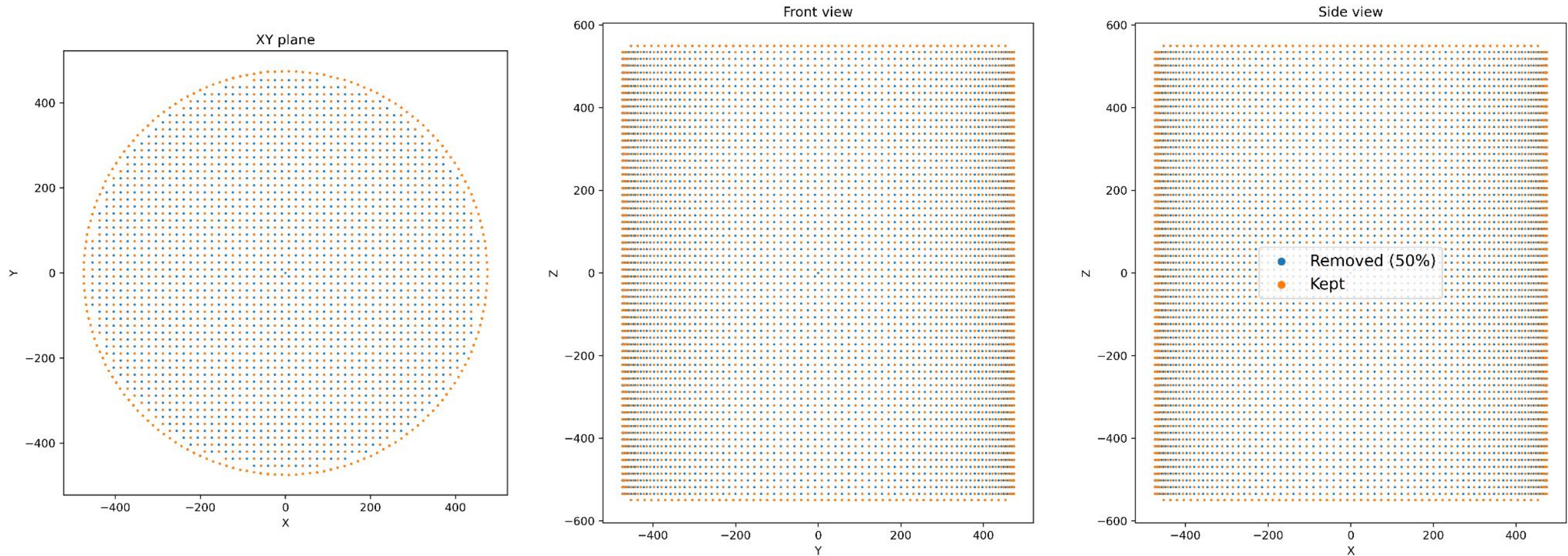
Event Classification Performance

- The GNN has acceptable performance even on the full events
- Using the GNN, the data cuts can be made obsolete

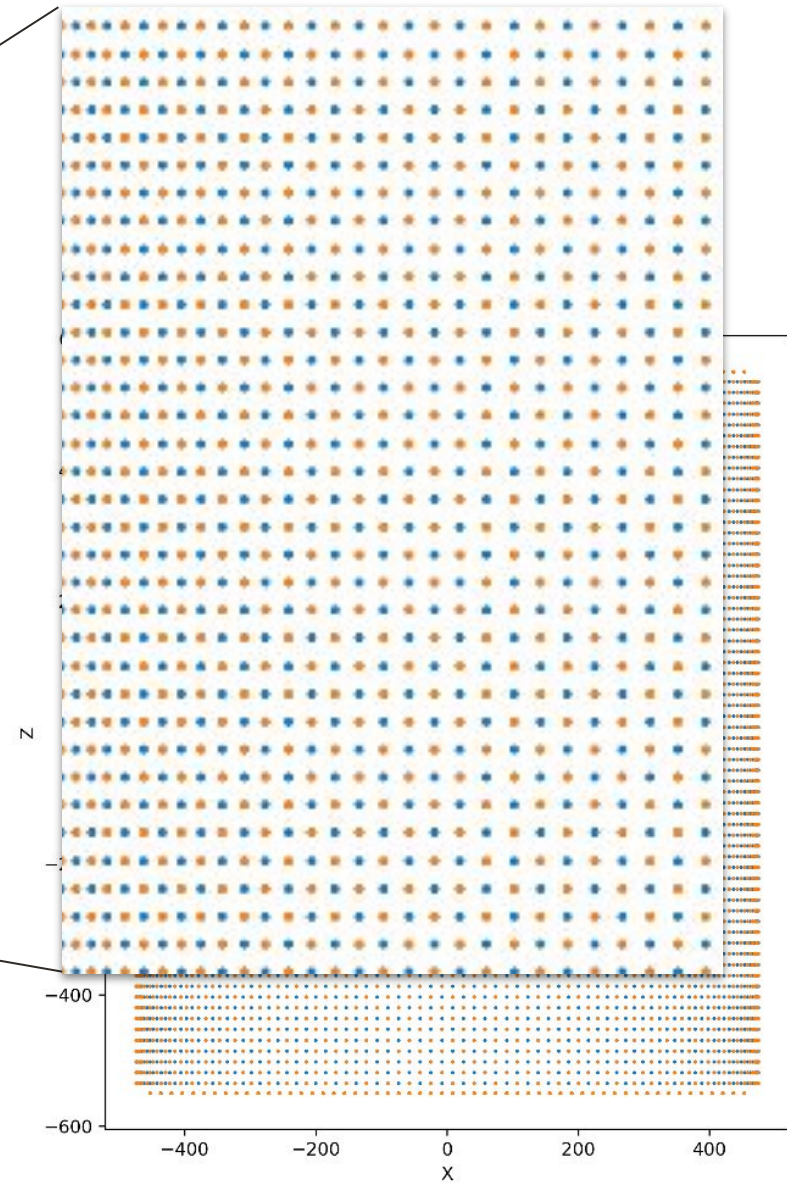
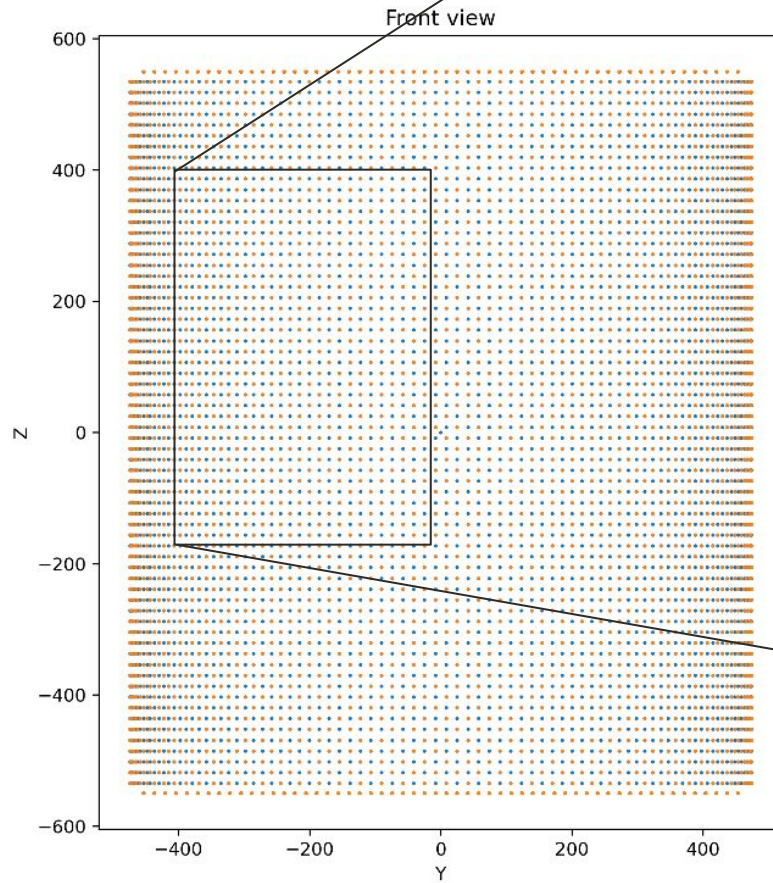
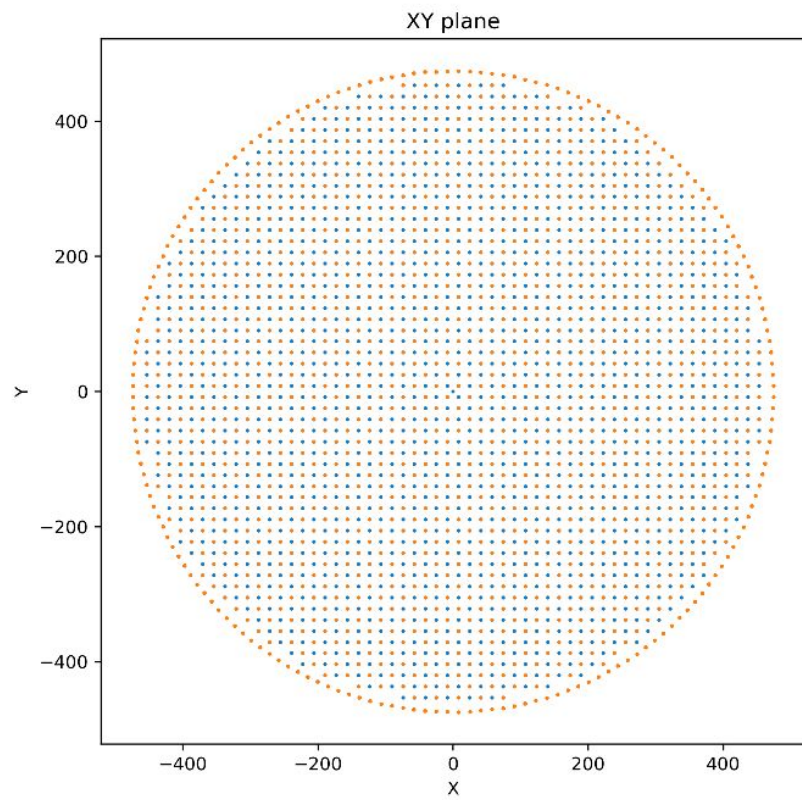


Generalisation

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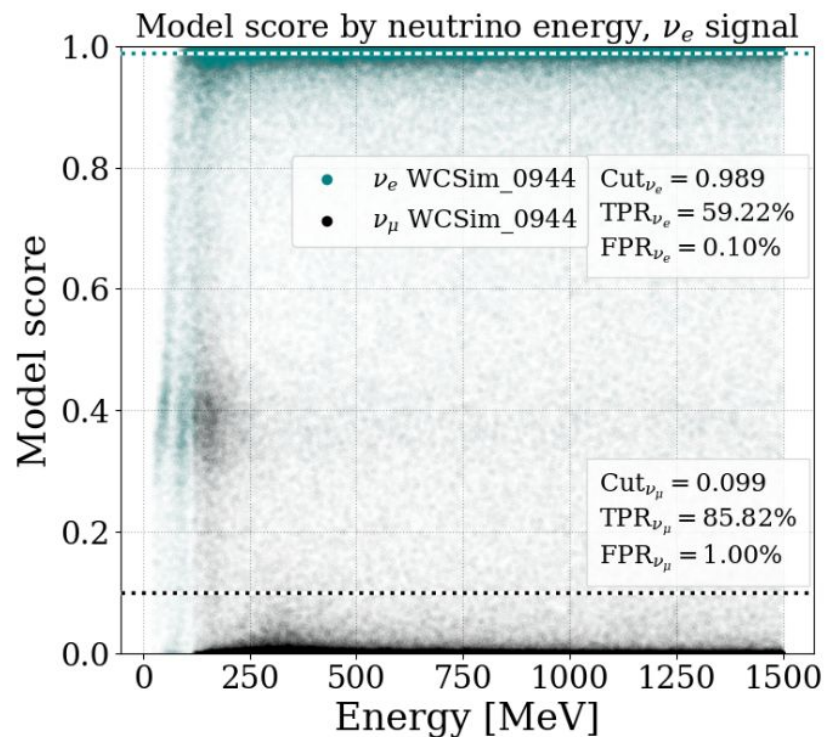


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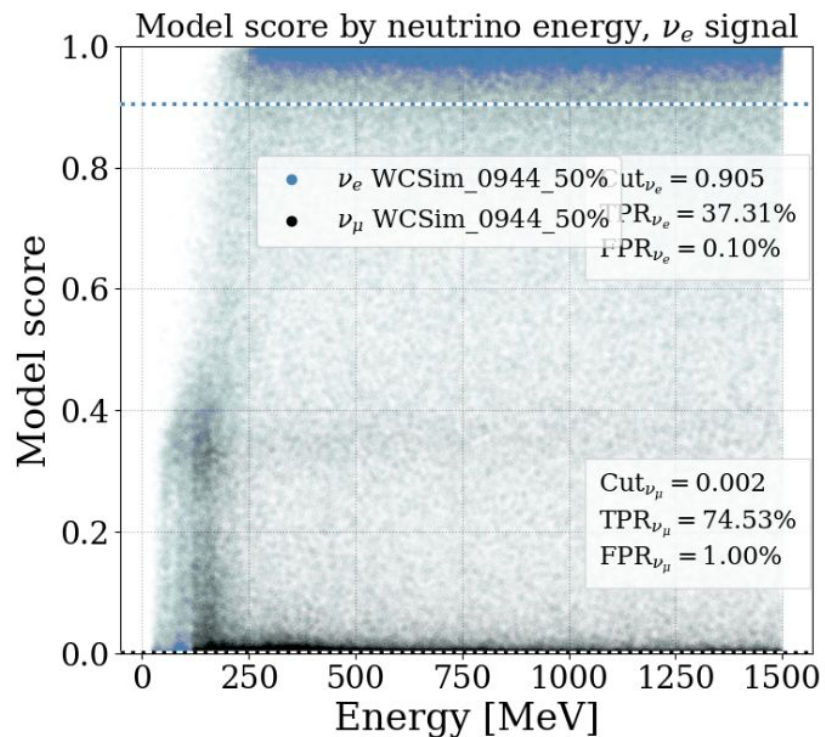


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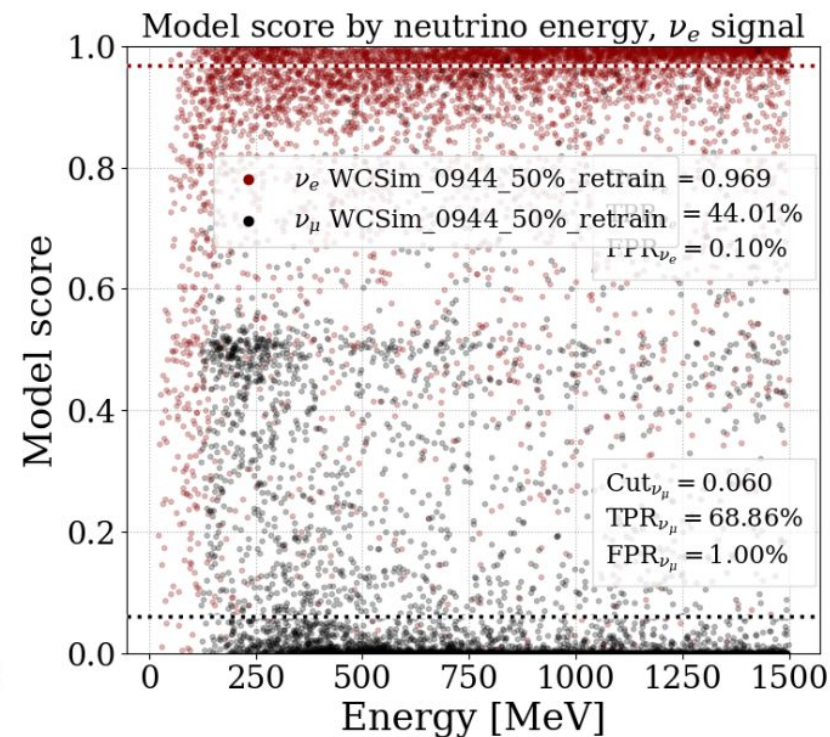
Original PMTs



50% PMTs
Original model

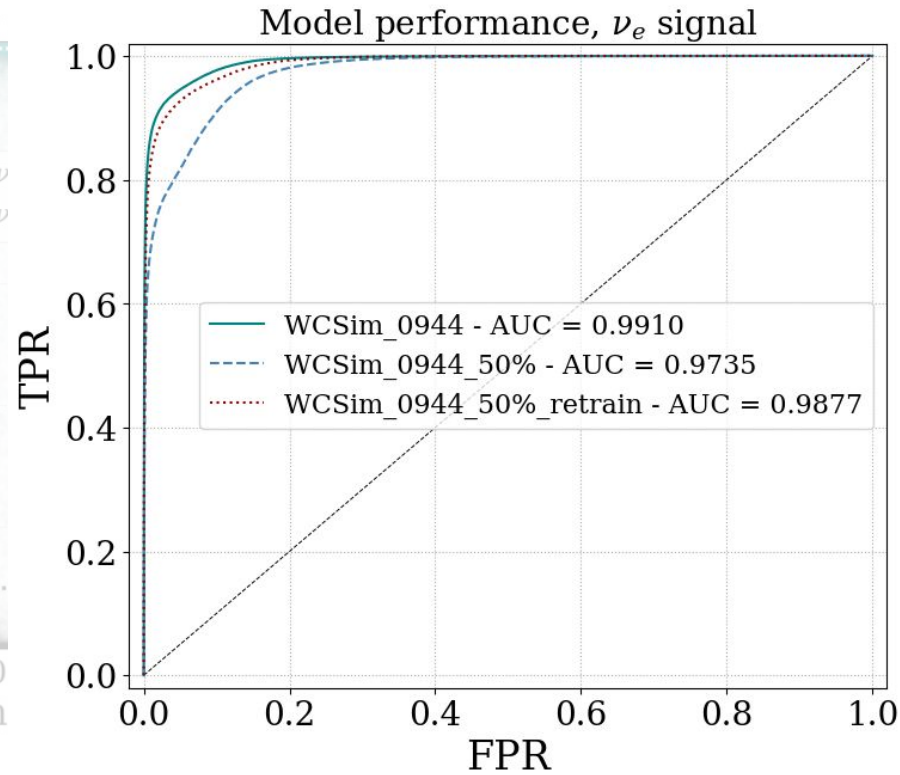
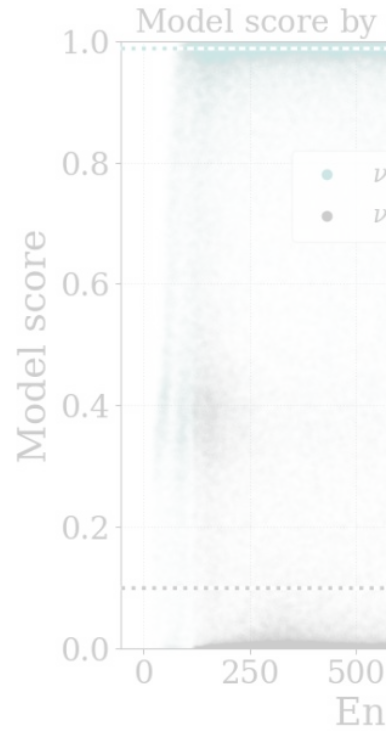


50% PMTs
Retrained model

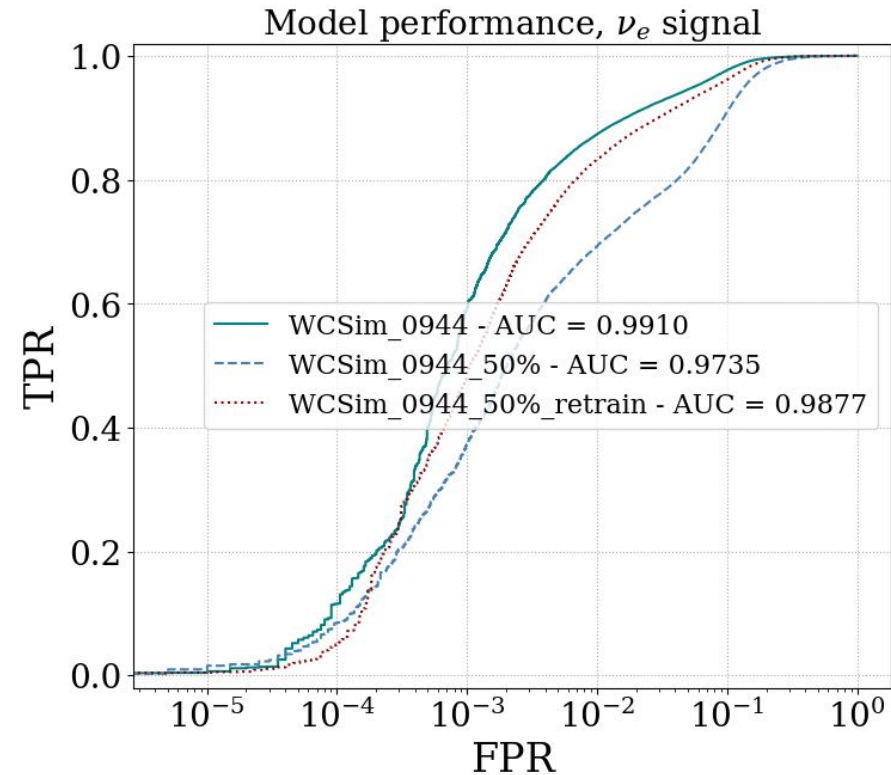


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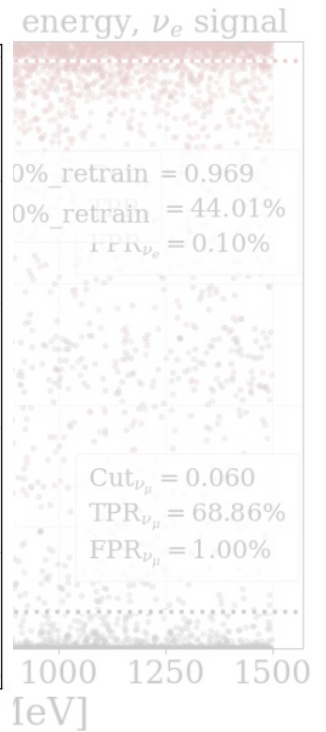
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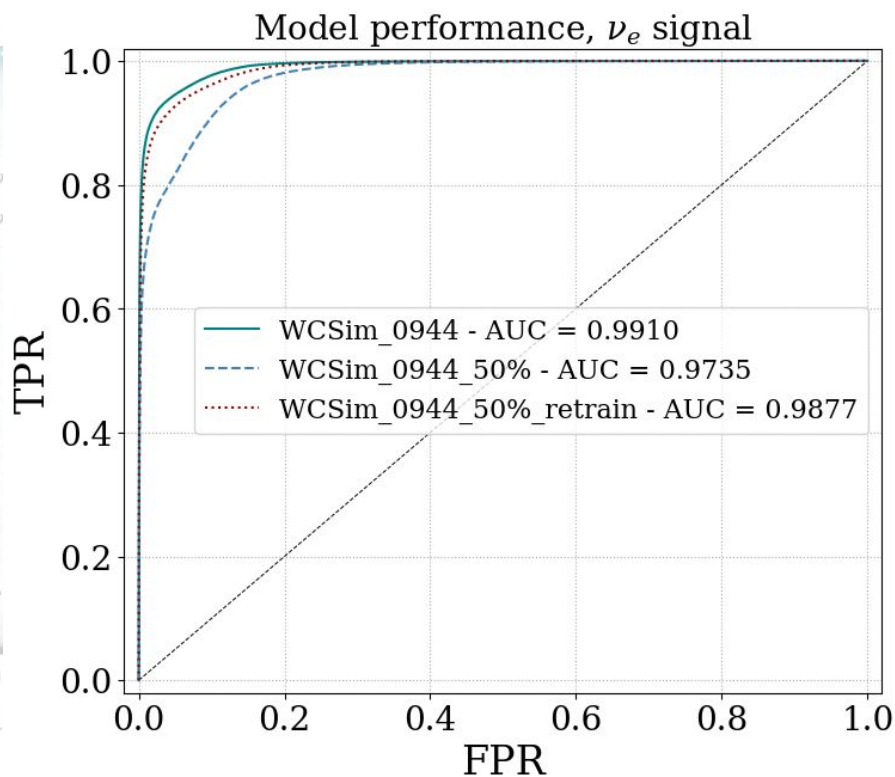
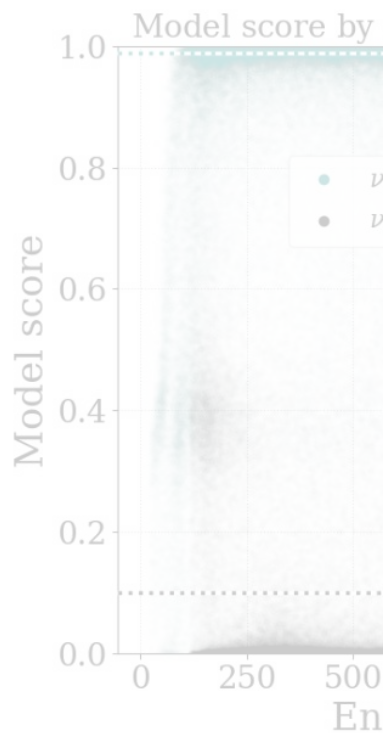


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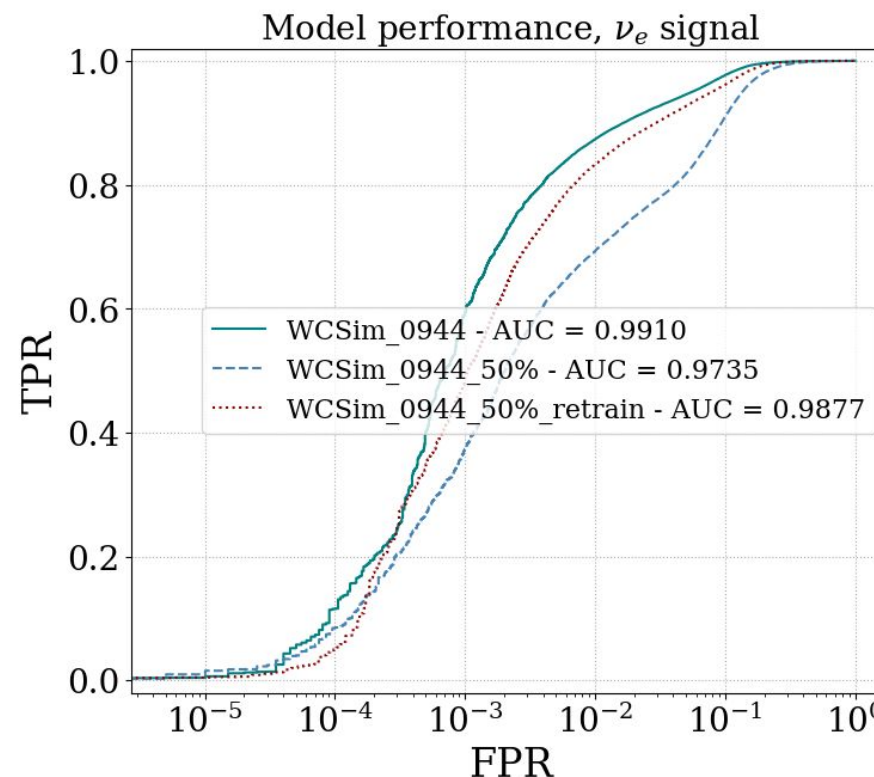


Generalisation

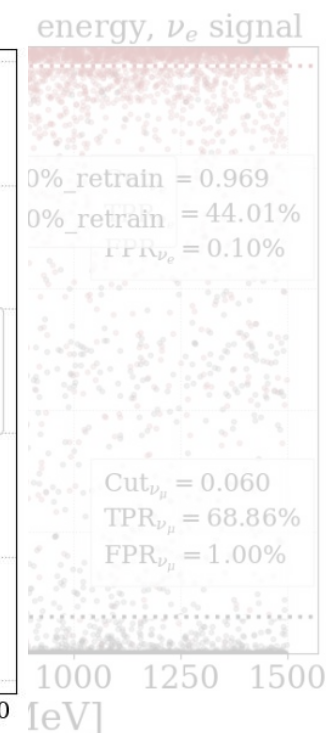
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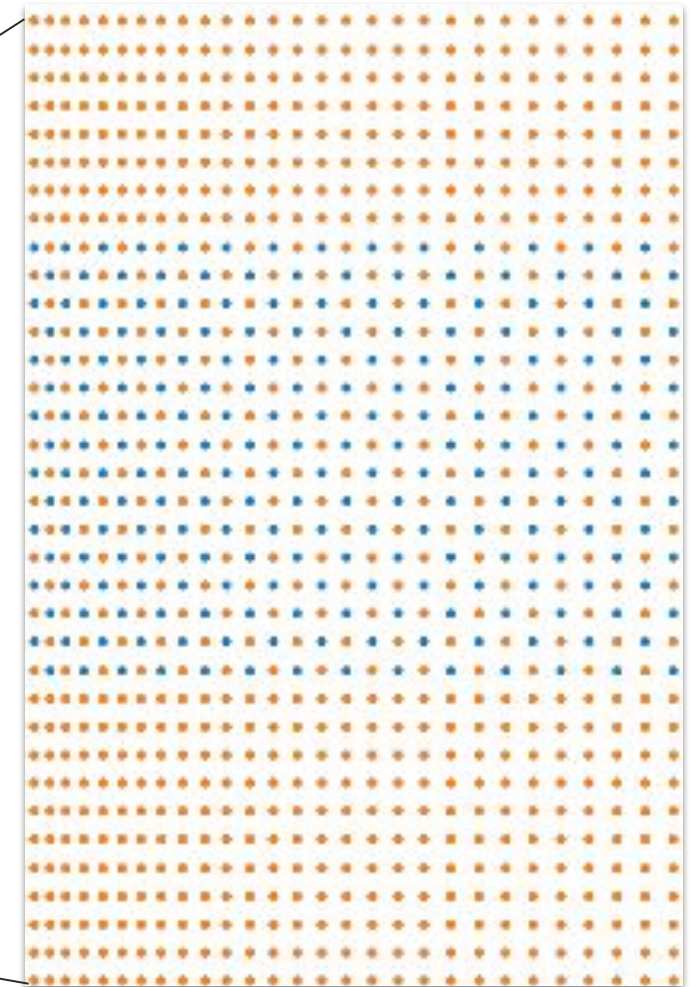
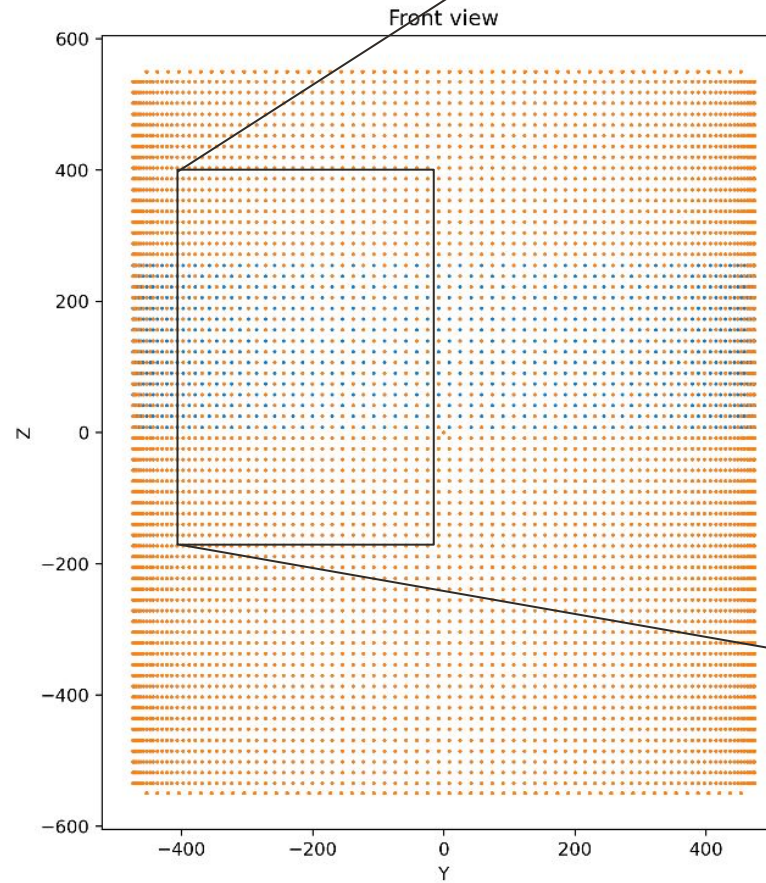
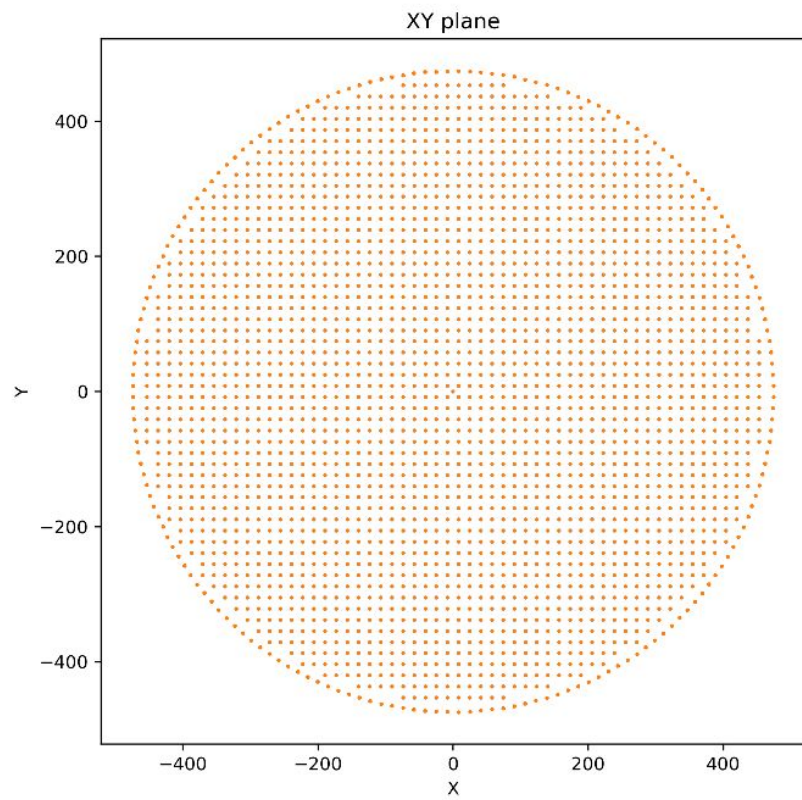


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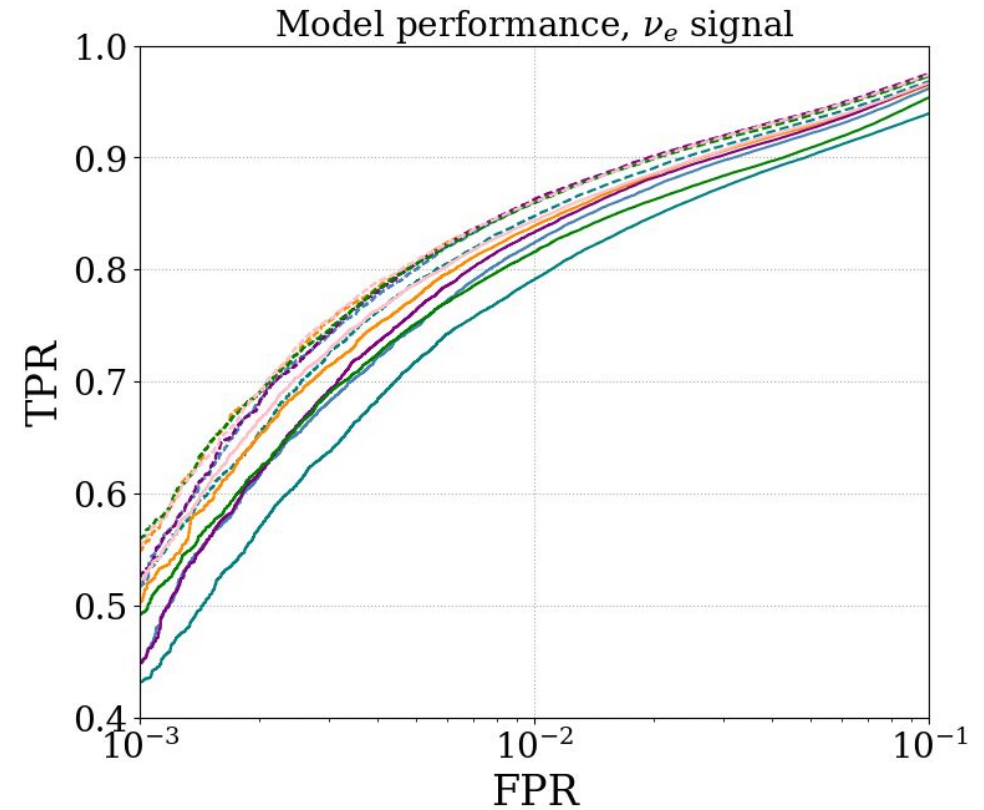
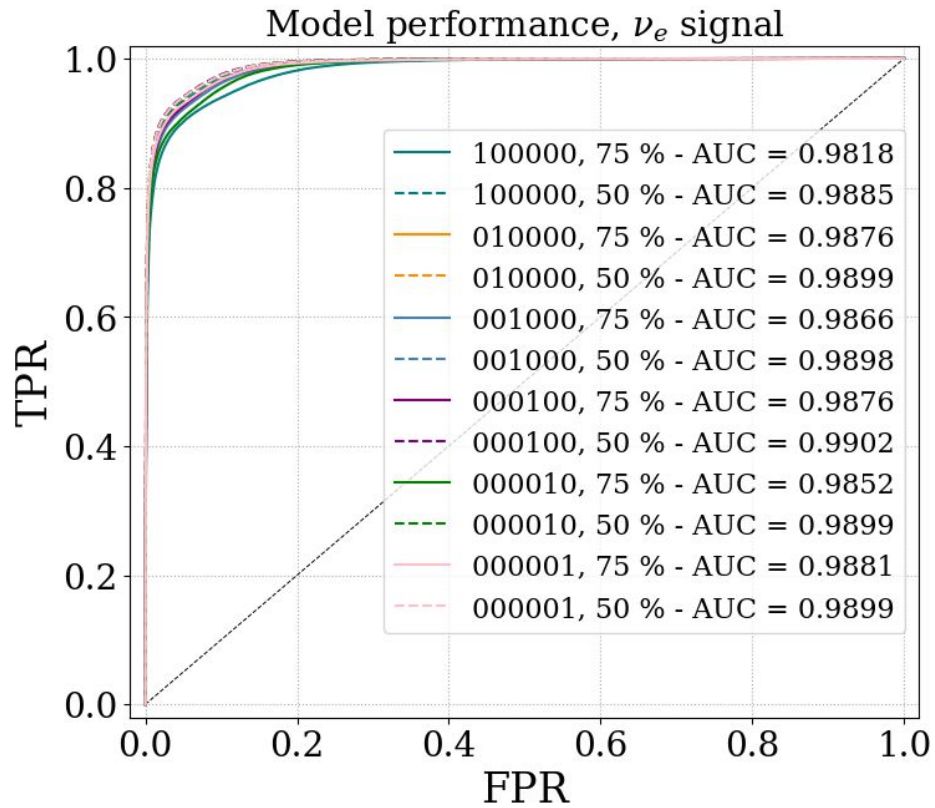


What if we don't switch off all PMTs uniformly?

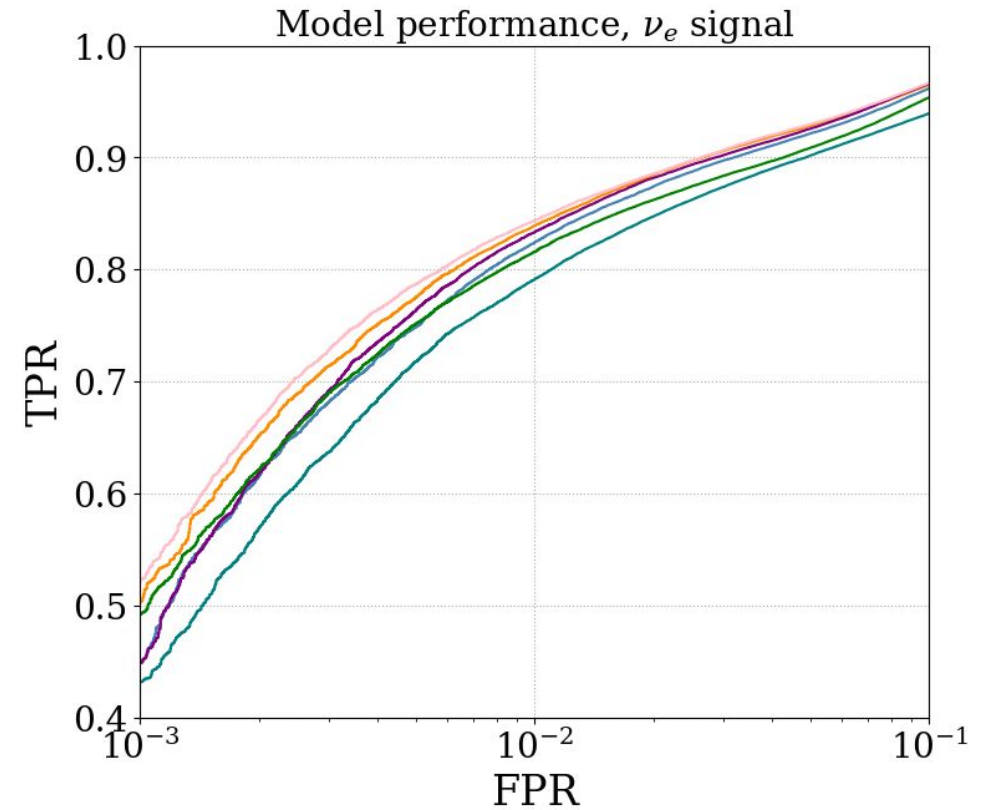
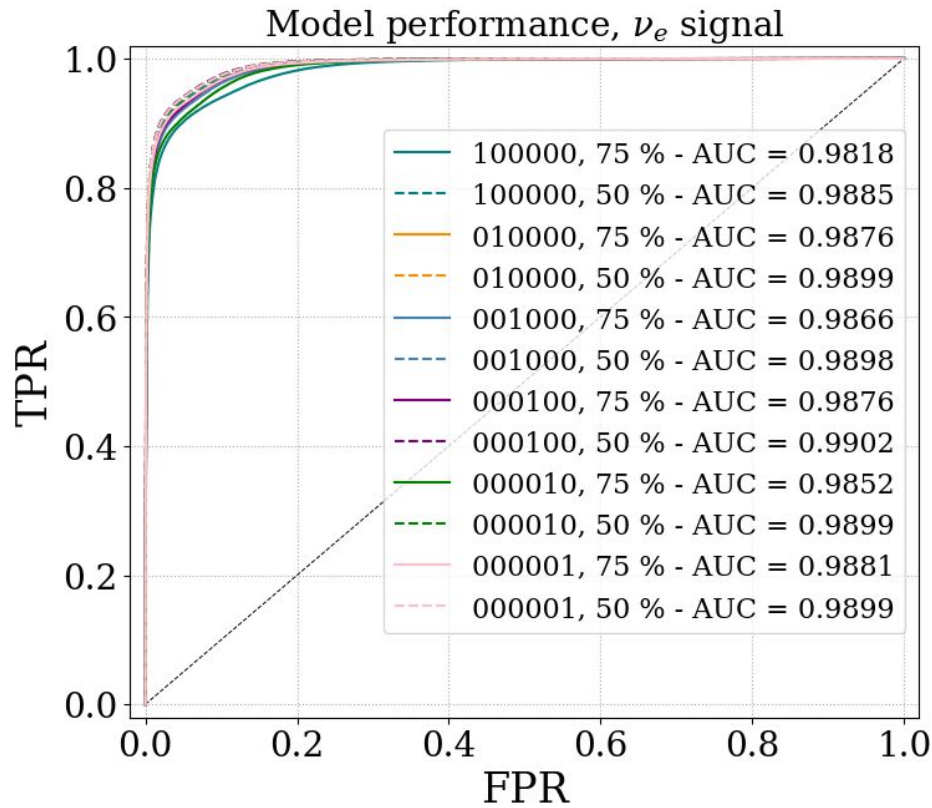
Generalisation



Generalisation



Generalisation



Summary

- GNNs can match fiTQun and outperform on difficult topologies
- Fast reconstruction can aid detector design
 - not all PMTs are created equal

Next steps

- Qualify PMT importance with simulation information
- Use more complex schemes to adjust PMTs
- Investigate event topologies of “difficult” events

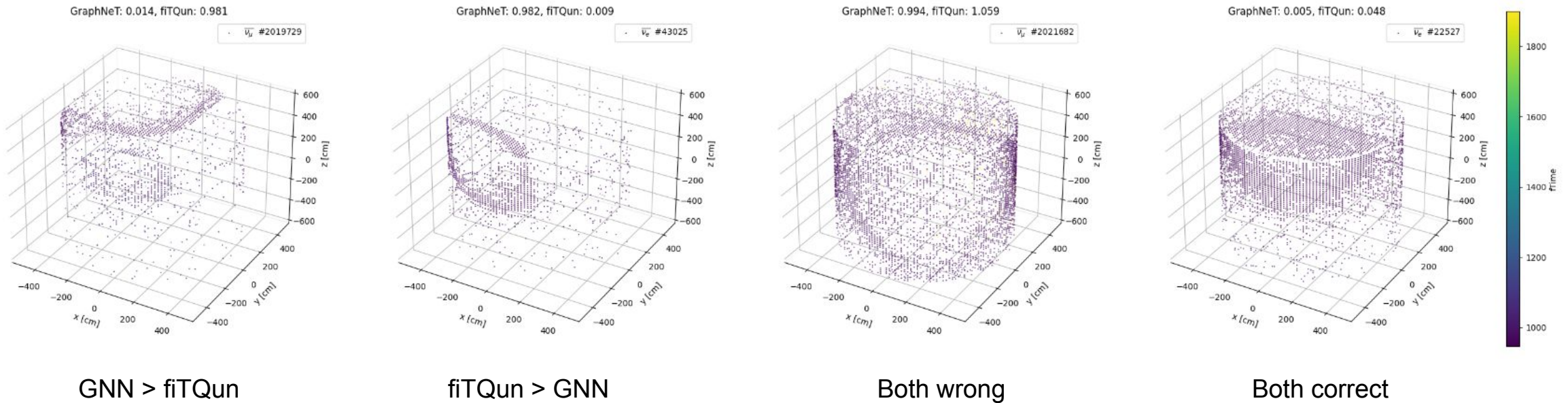
An aerial architectural rendering of a city development. The scene features a winding river in the foreground, a multi-lane highway on the right, and a cluster of buildings in the center. A prominent circular structure with a green logo is visible near the river. The background shows a dense urban area with various buildings and green spaces.

Thank you!

Backup Slides

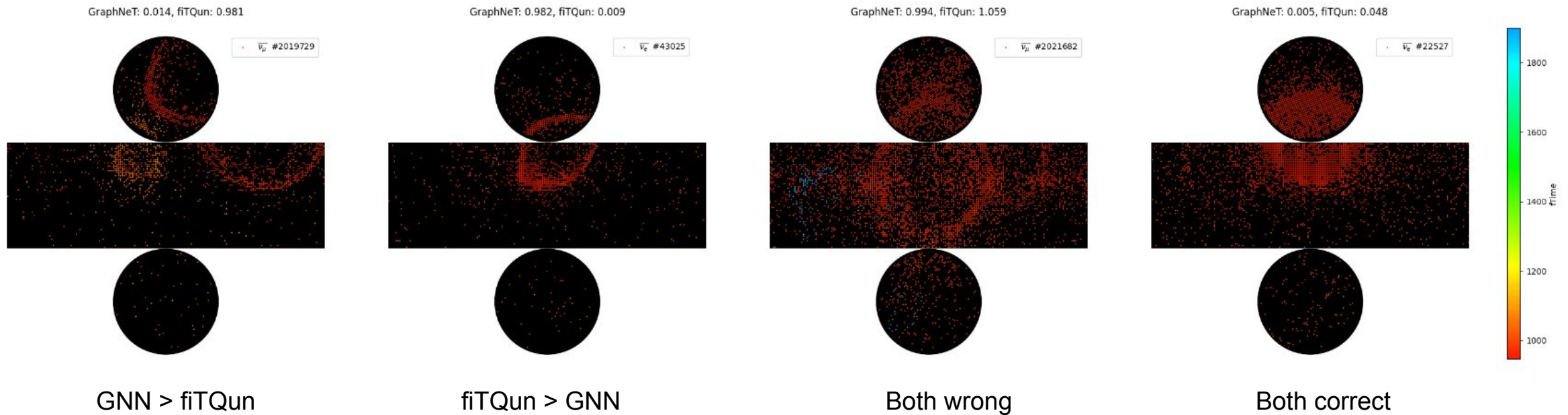
Factors impacting performance - multiple charged lepton signatures

Neutrino event simulations - **with** data cut

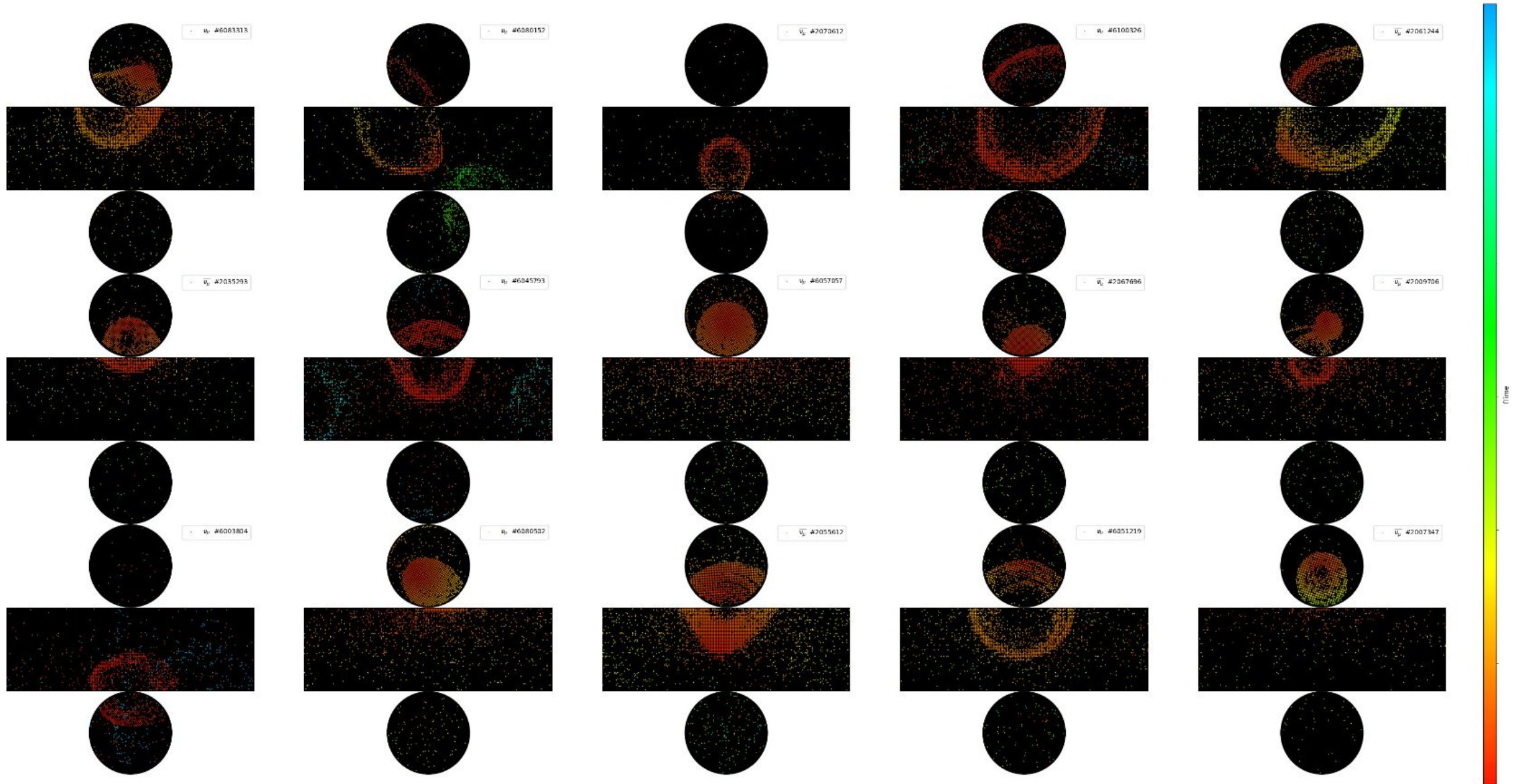


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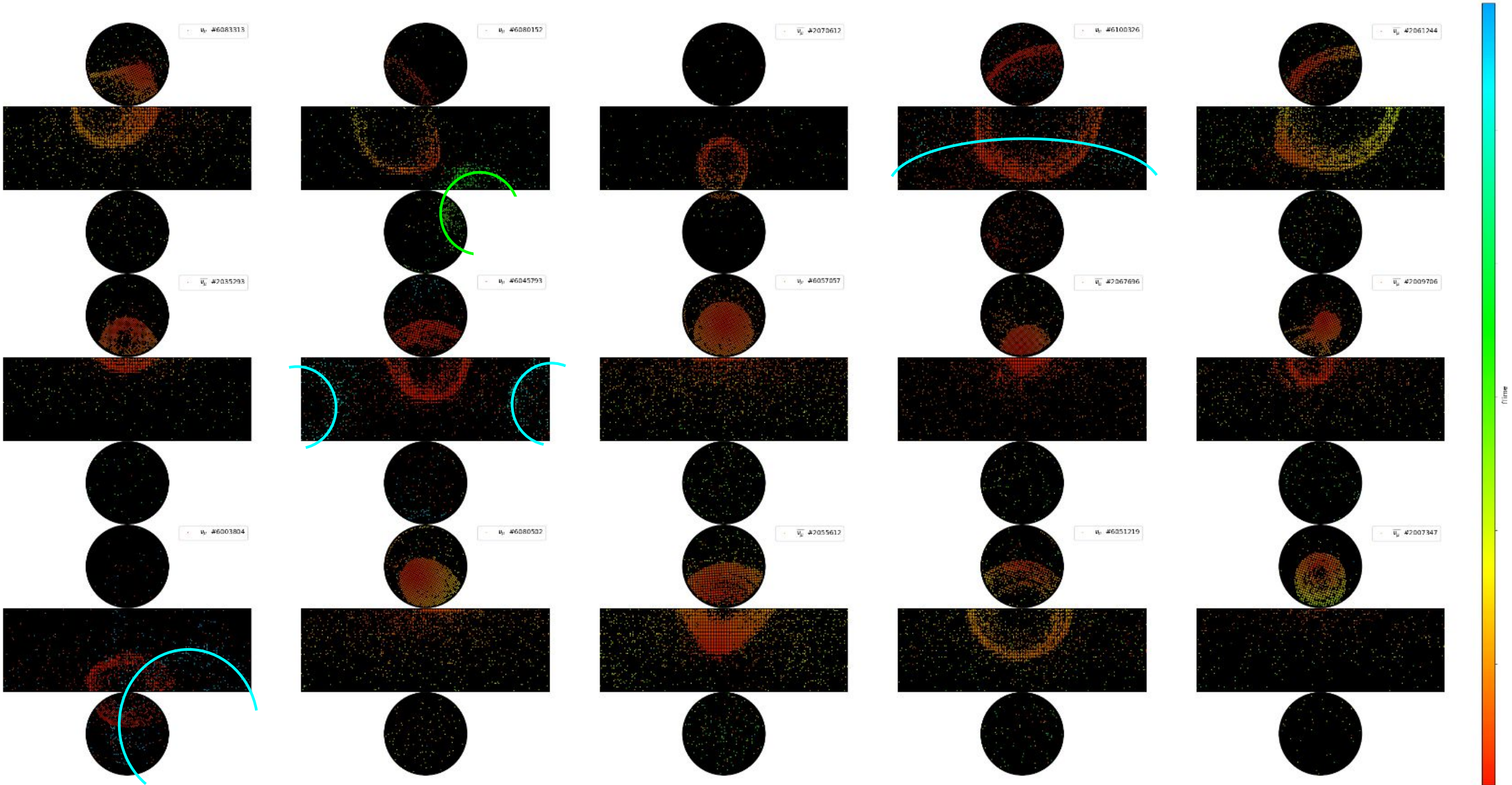
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Event for which GraphNeT performs significantly better than fiTQun



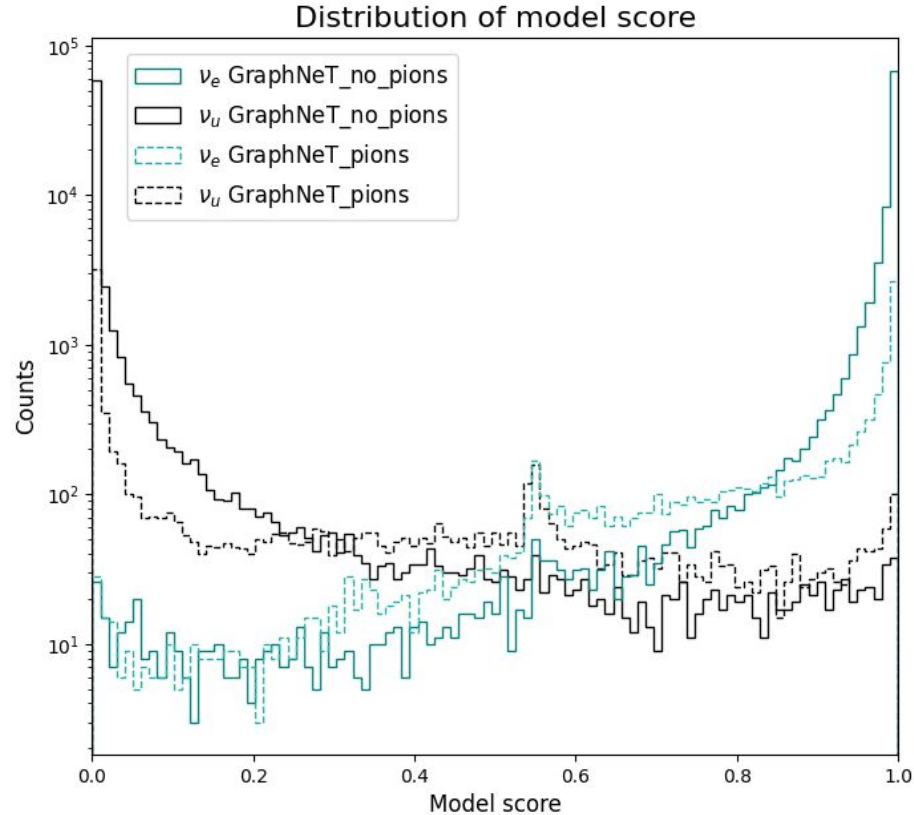
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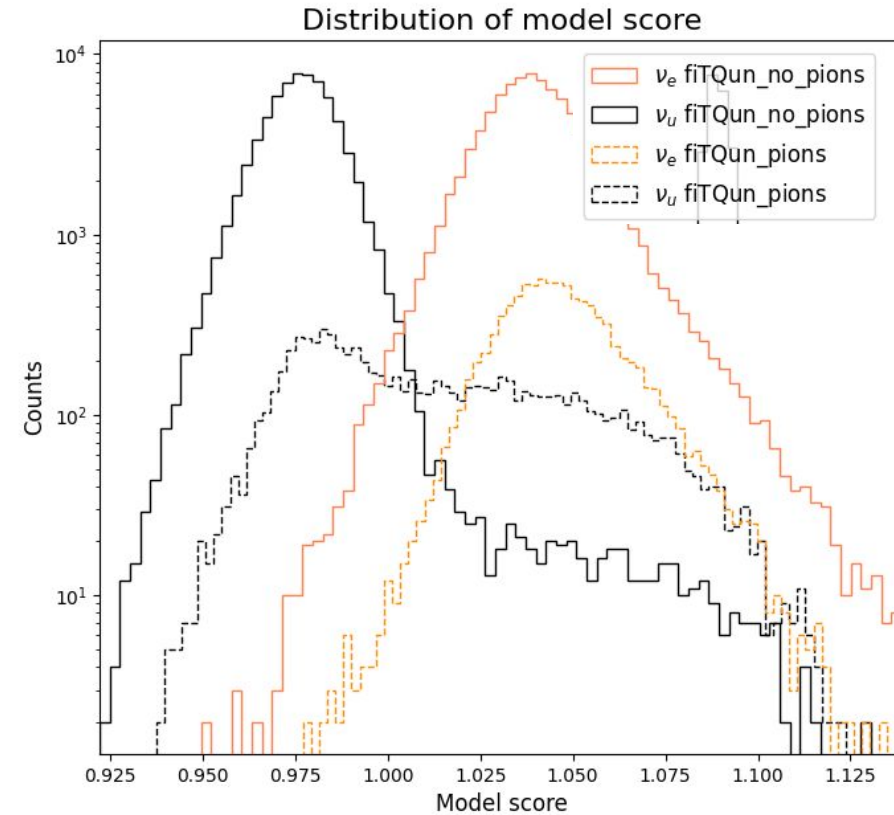
Factors impacting performance - pion creation

Neutrino event simulations - with data cut

GraphNeT performance with logarithmic x axis

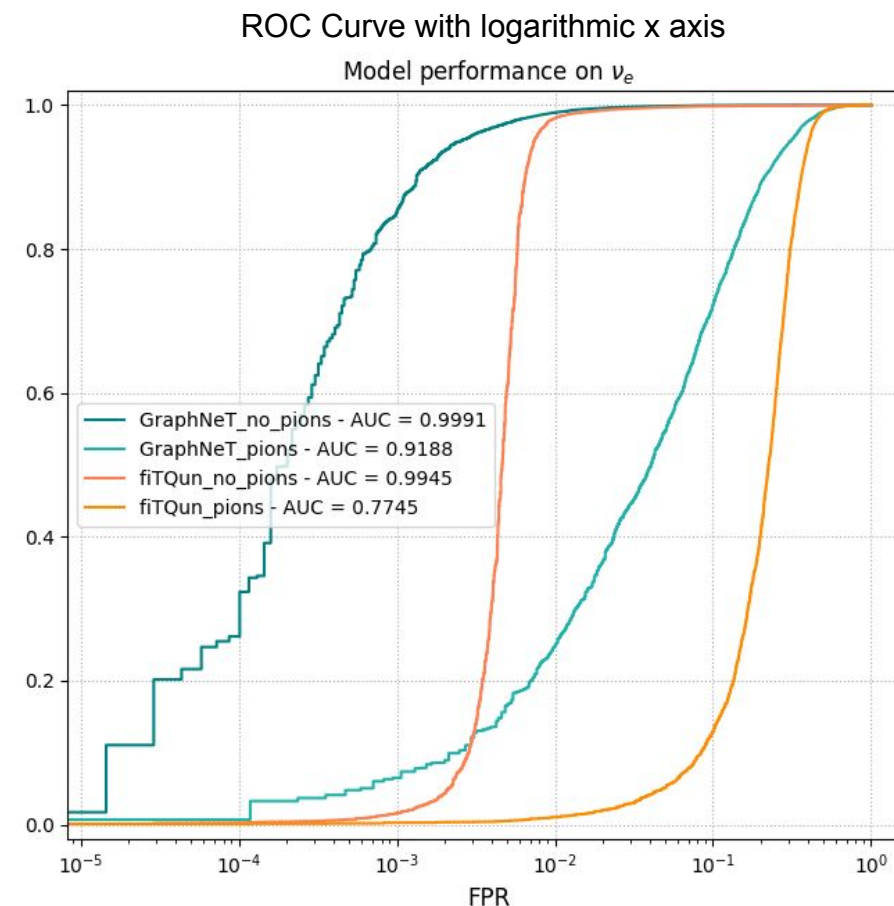
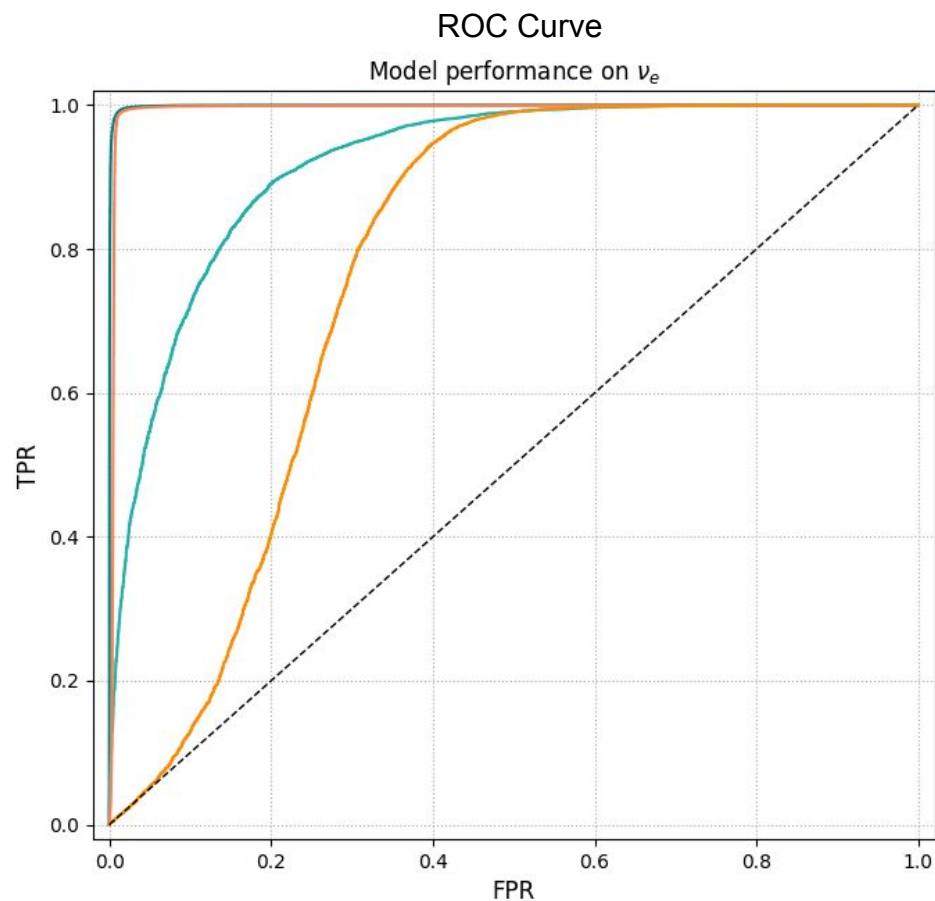


fiTQun performance with logarithmic x axis



Factors impacting performance - pion creation

Neutrino event simulations - with data cut



Event Classification - Performance Measures

The neutrino beam will consist of > 98 % muon neutrinos (at the near detector)

- We can allow more electron neutrino events to be misidentified and still have a pure muon neutrino event sample



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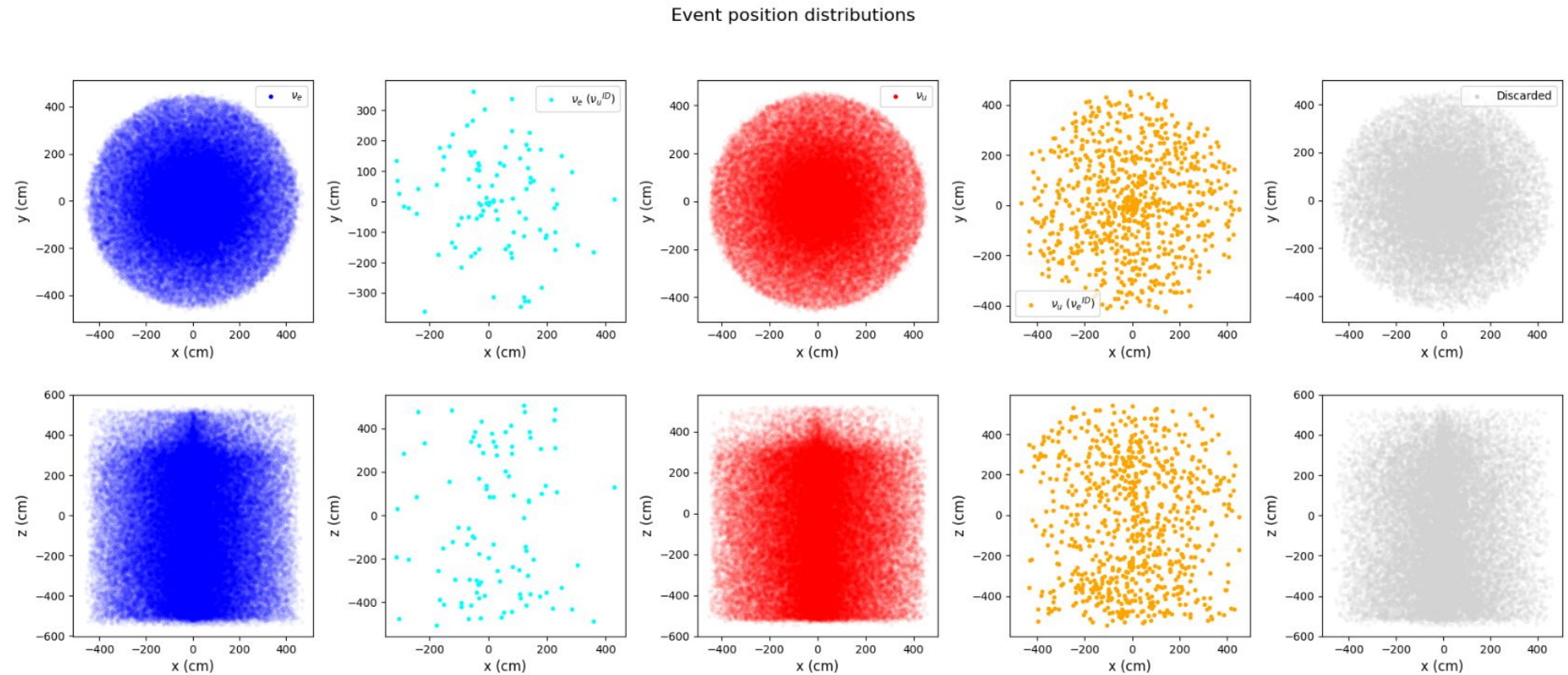
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We select samples that yield FPRs of:

- **1 % for muon neutrinos**
- **0.1 % for electron neutrinos**

Relation between interaction position and reconstruction performance

Neutrino event simulations - with data cut



Relation between interaction position and reconstruction performance

Neutrino event simulations - with data cut

