



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



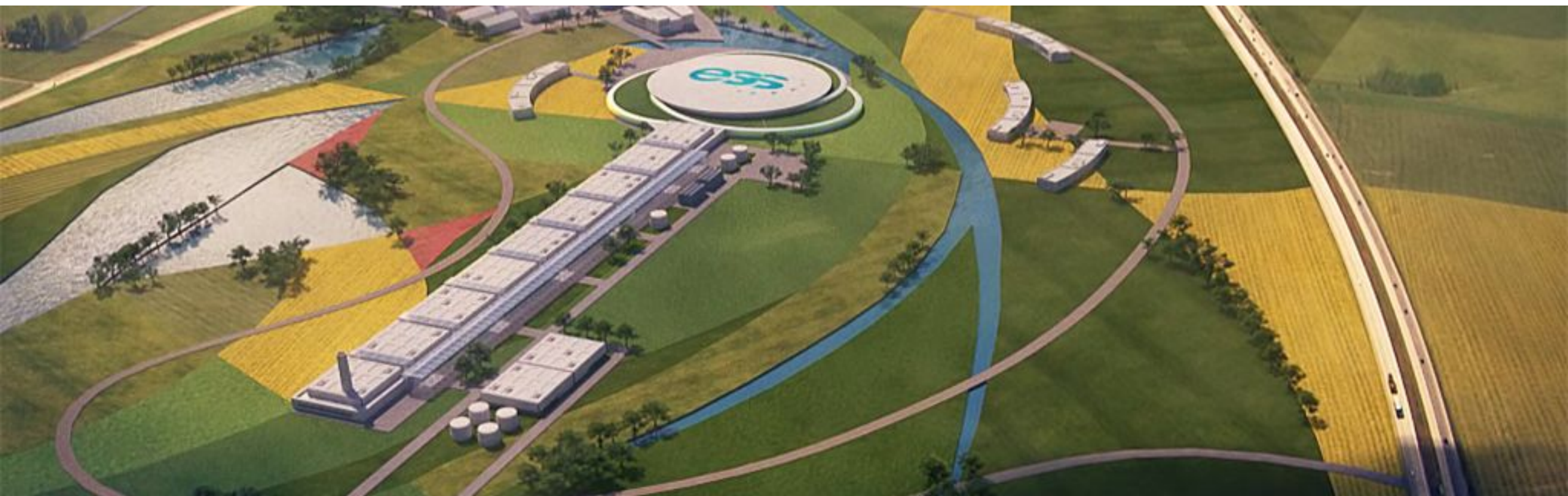
Outline

- ESS ν SB - An accelerator neutrino experiment in Sweden
- GNN Implementation
- Classification Performance
- Generalisation

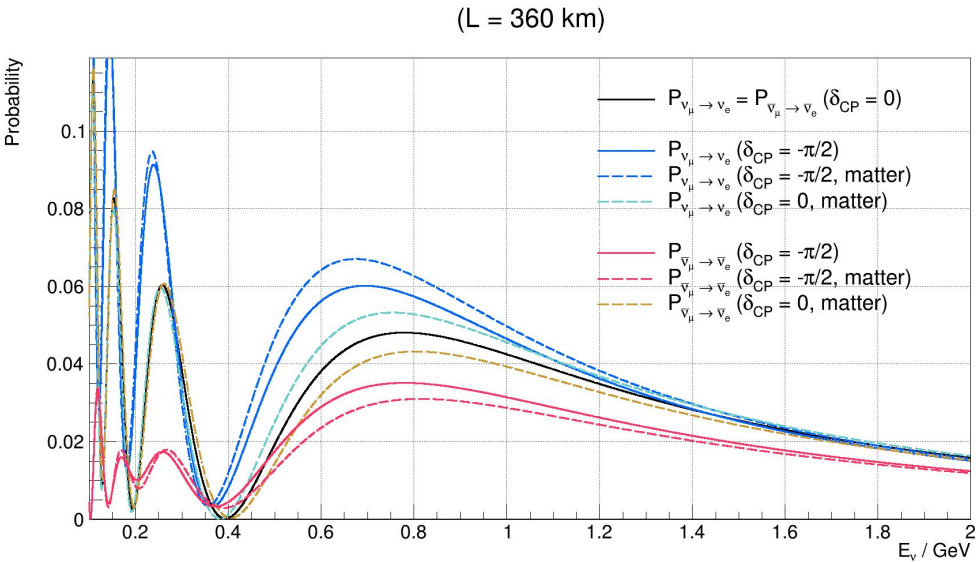
Motivation



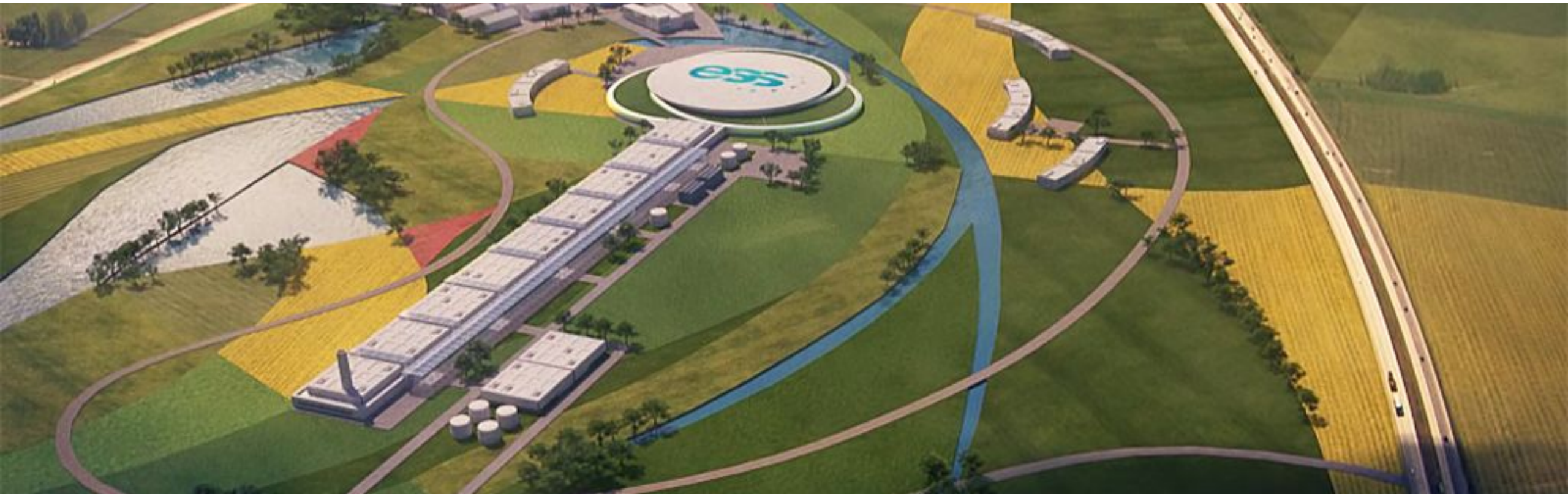
5 MW proton beam



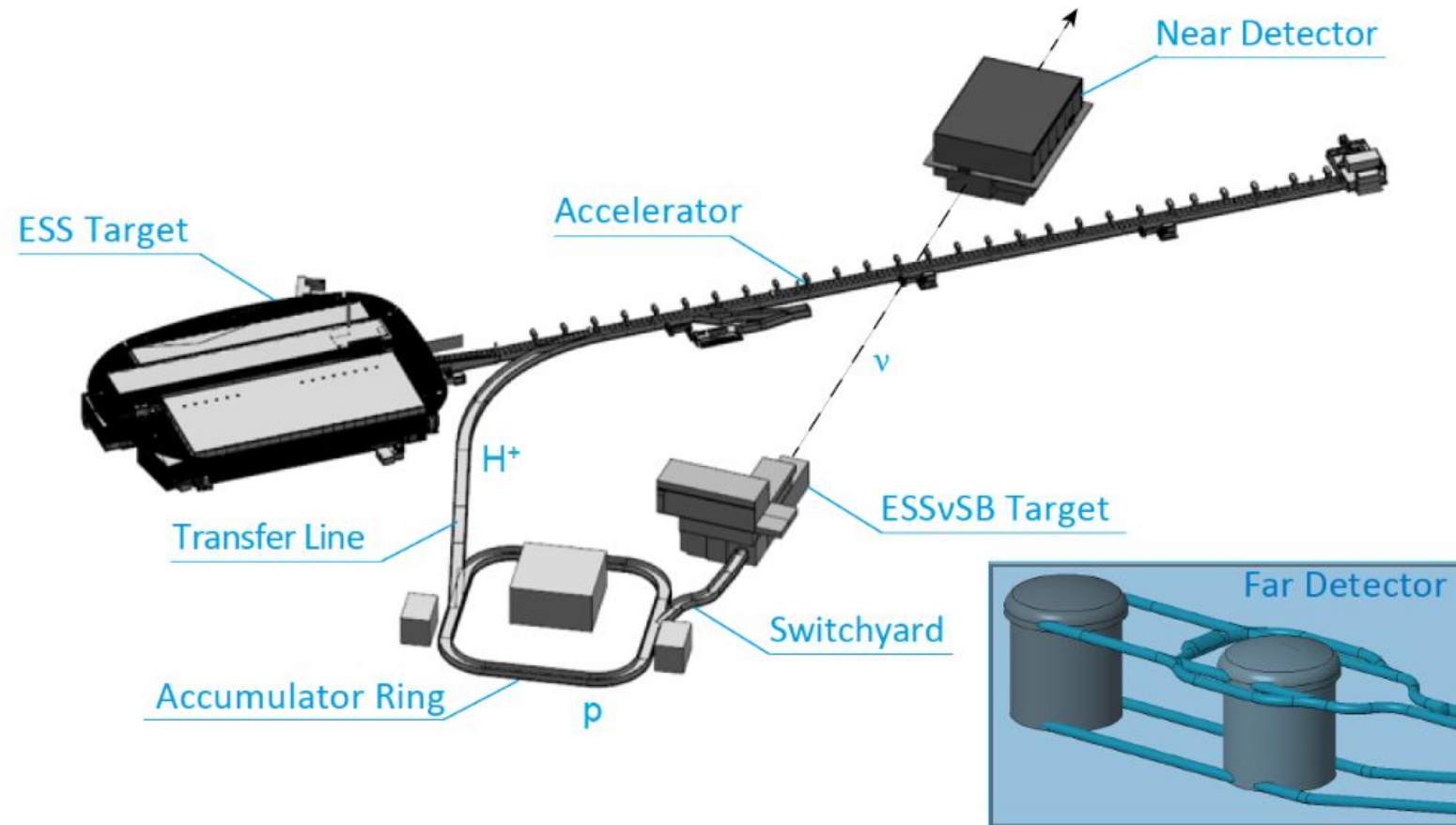
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The ESSnuSB Design Study, Universe 2023, 9(8), 347

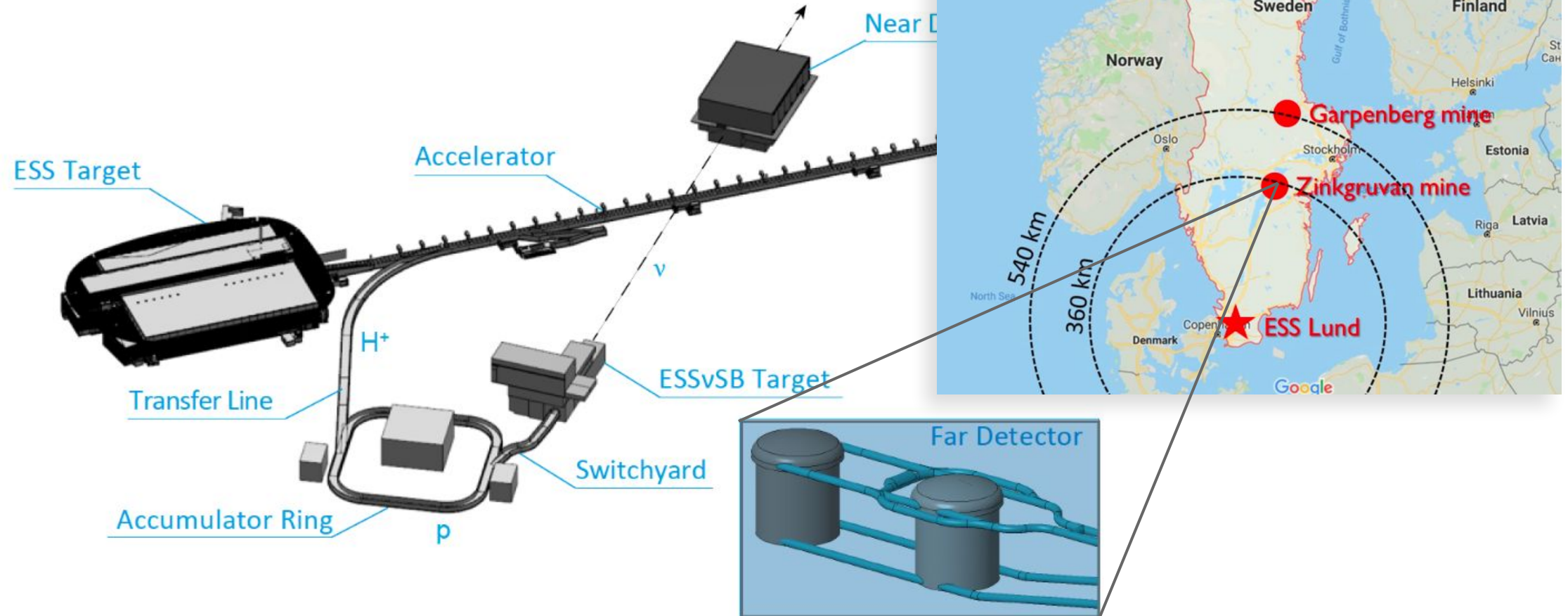


ESSnuSB - Detectors



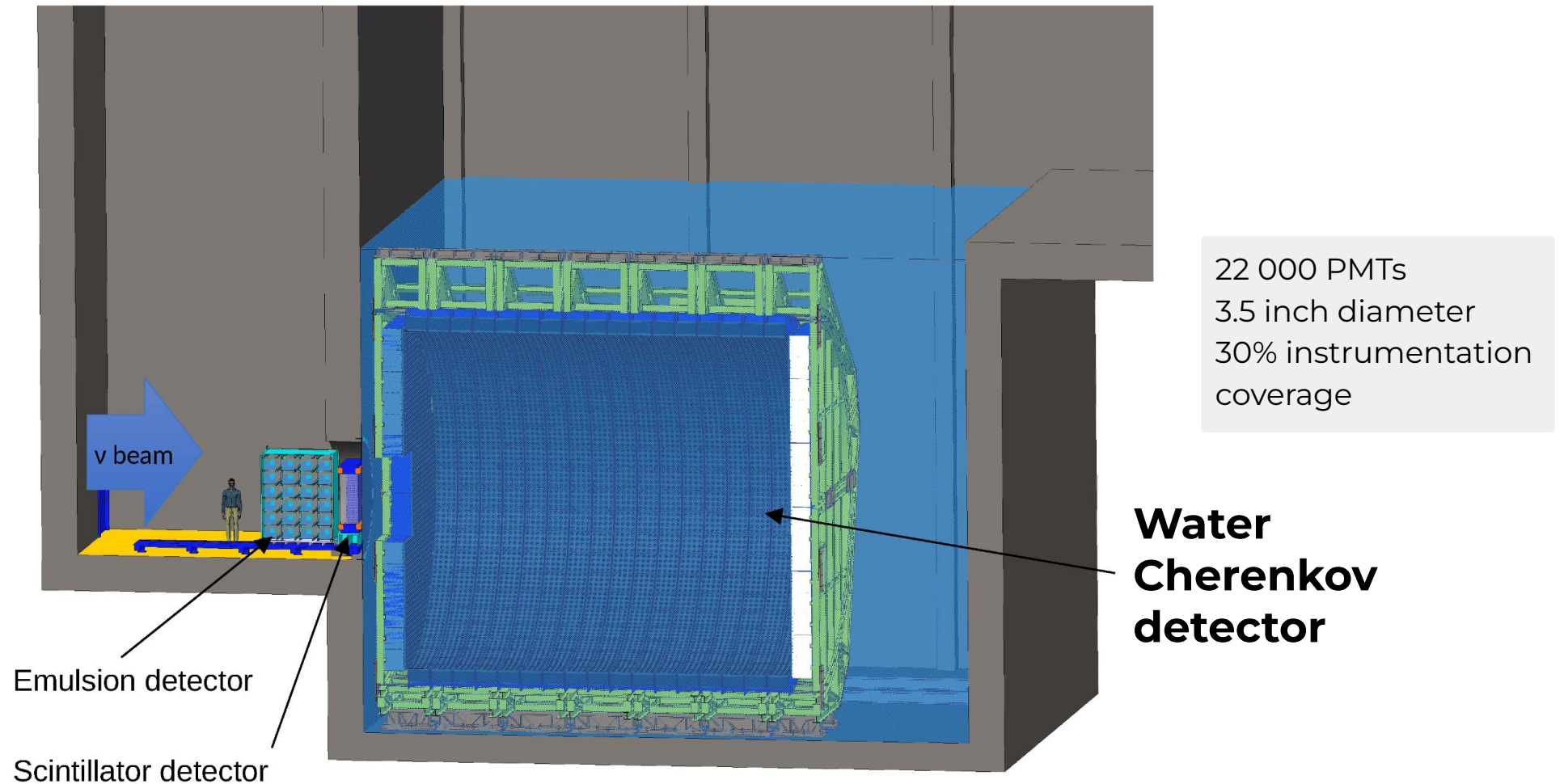
The European Spallation Source neutrino Super Beam Conceptual Design Report

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

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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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Challenges

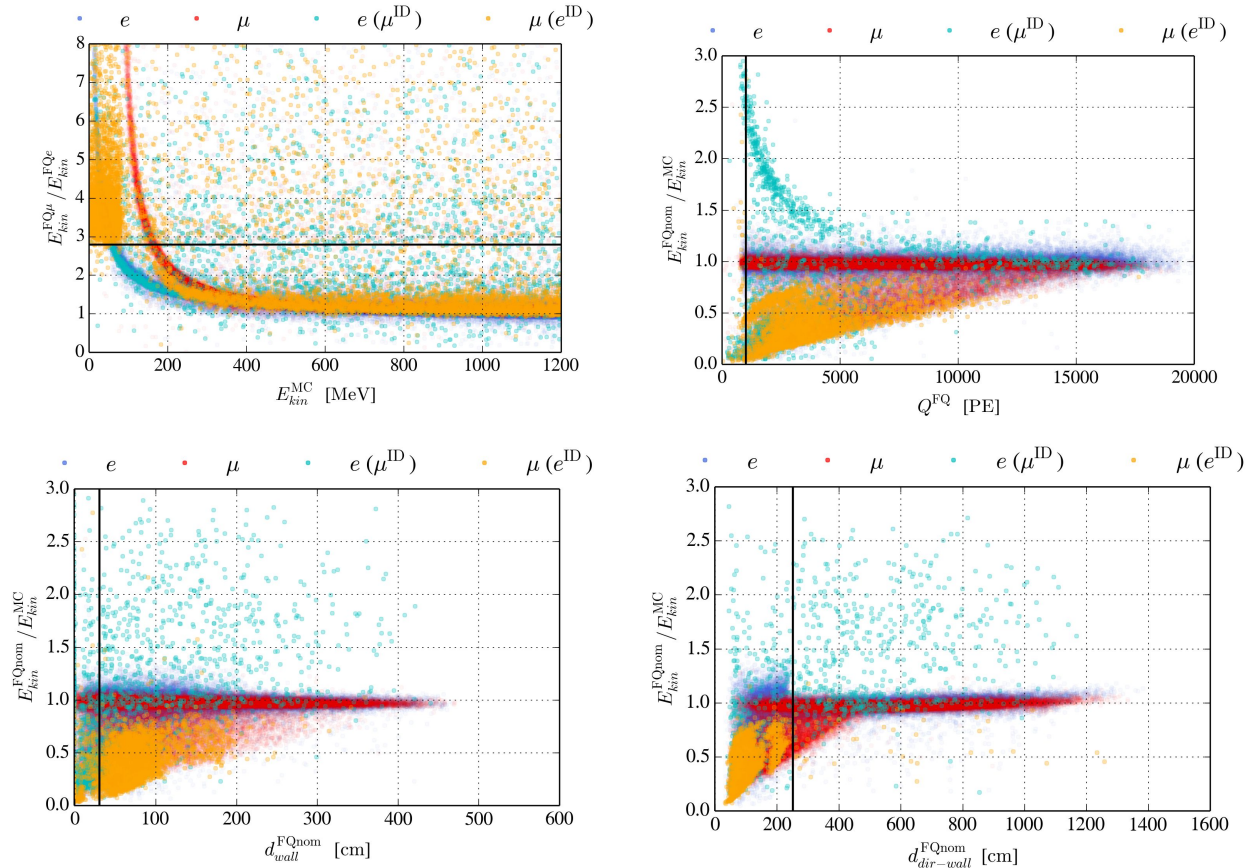
- Likelihood reconstruction takes ~1 CPU min/event
- To explore different detector proposals, fast reconstruction is crucial

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 events by a factor ~ 2



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GNN Implementation

GNN Framework



GraphNeT - Graph Neural Networks for Neutrino Telescopes

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

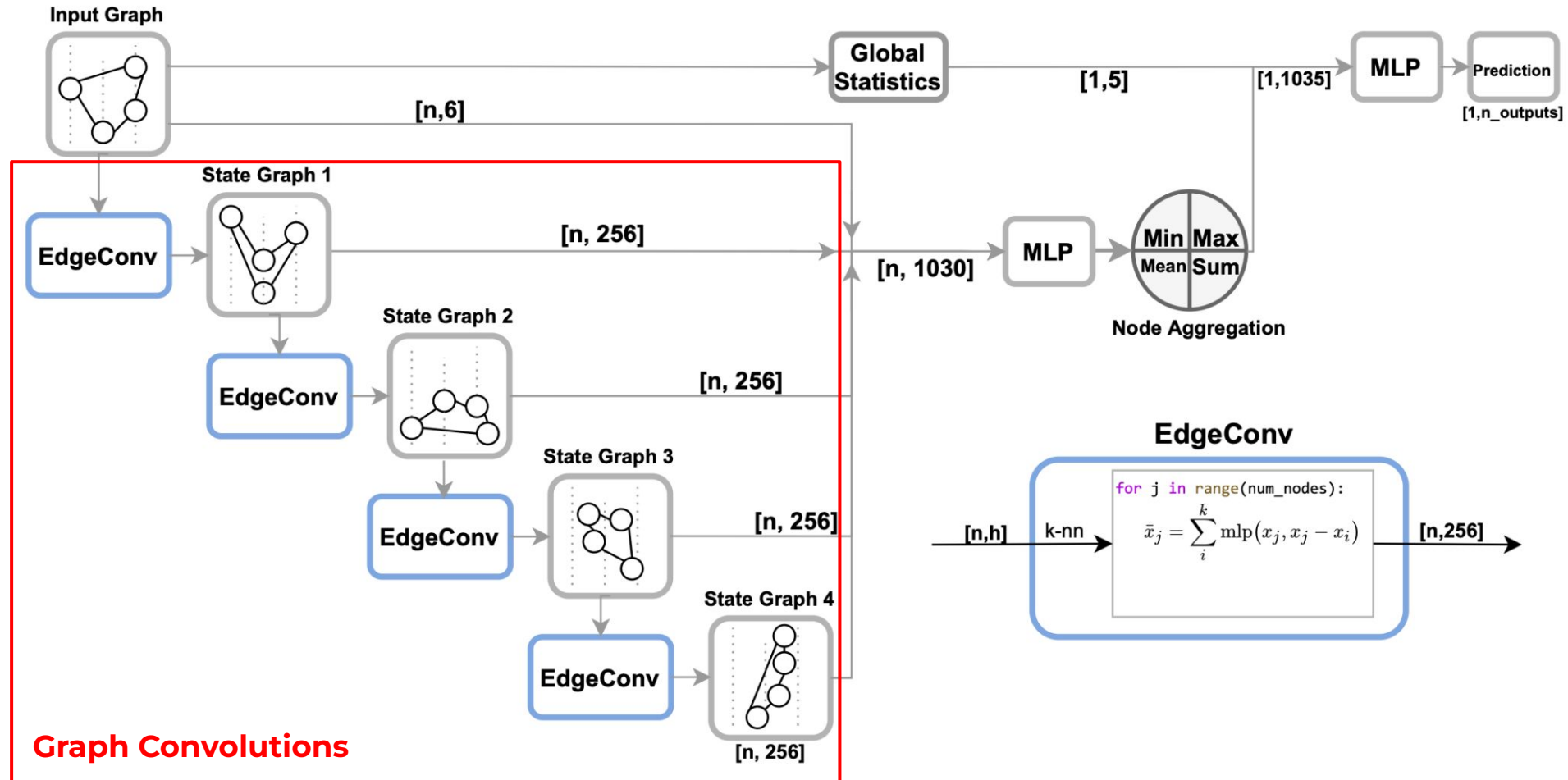
GraphNeT: Graph neural networks for neutrino telescope event reconstruction
(Søgaard et al)

1 PMT = 1 node

Node features: xyz, time, charge

Target: PID

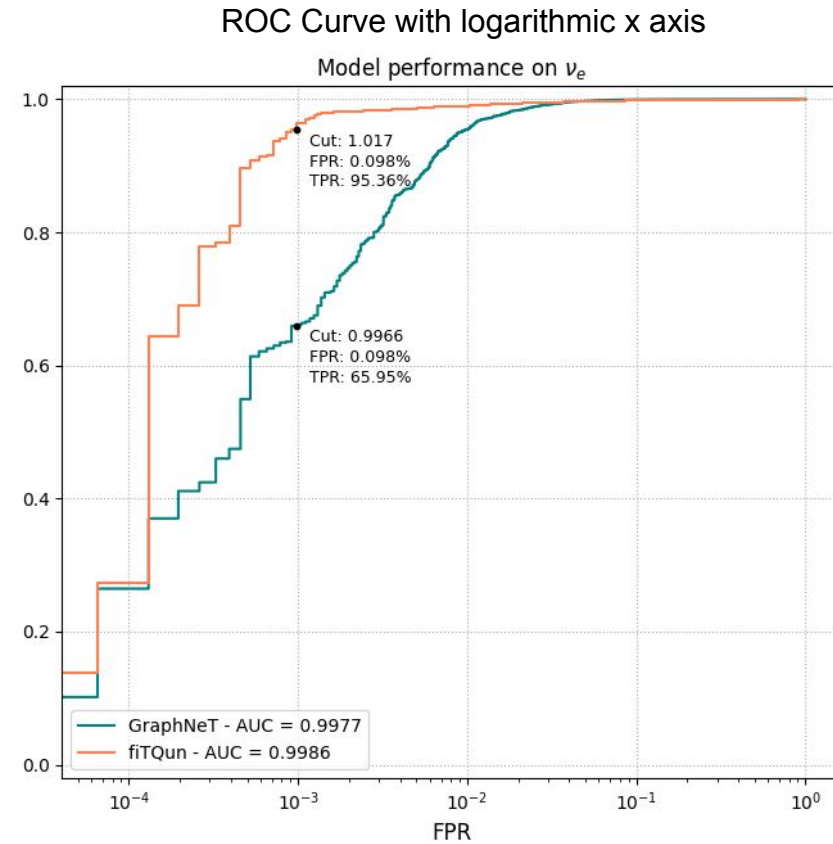
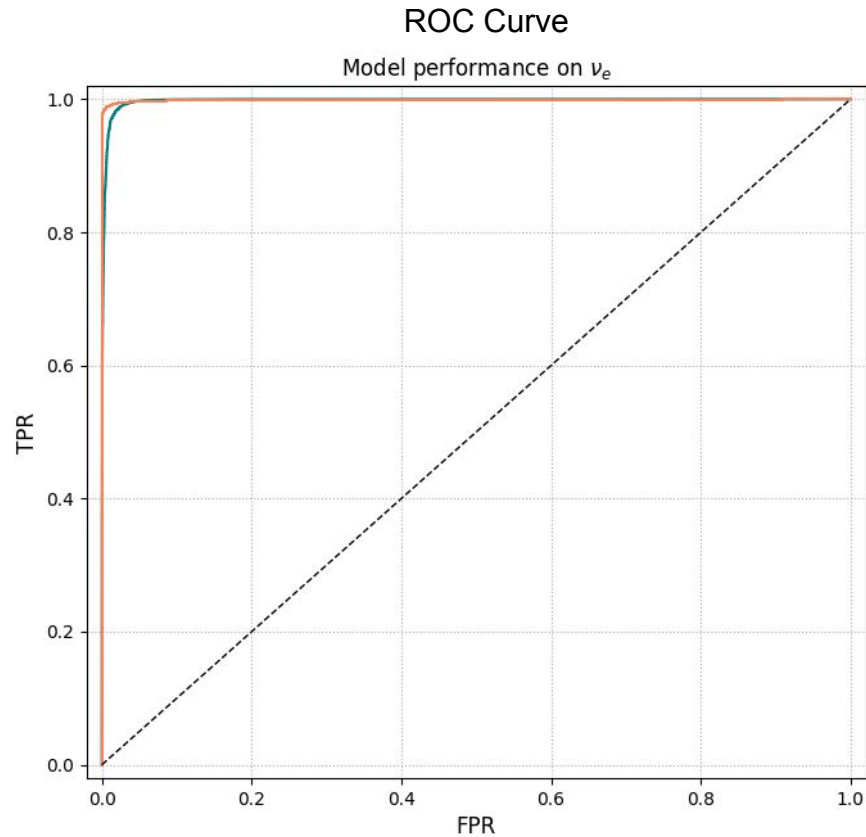
GNN Architecture



Classification Performance

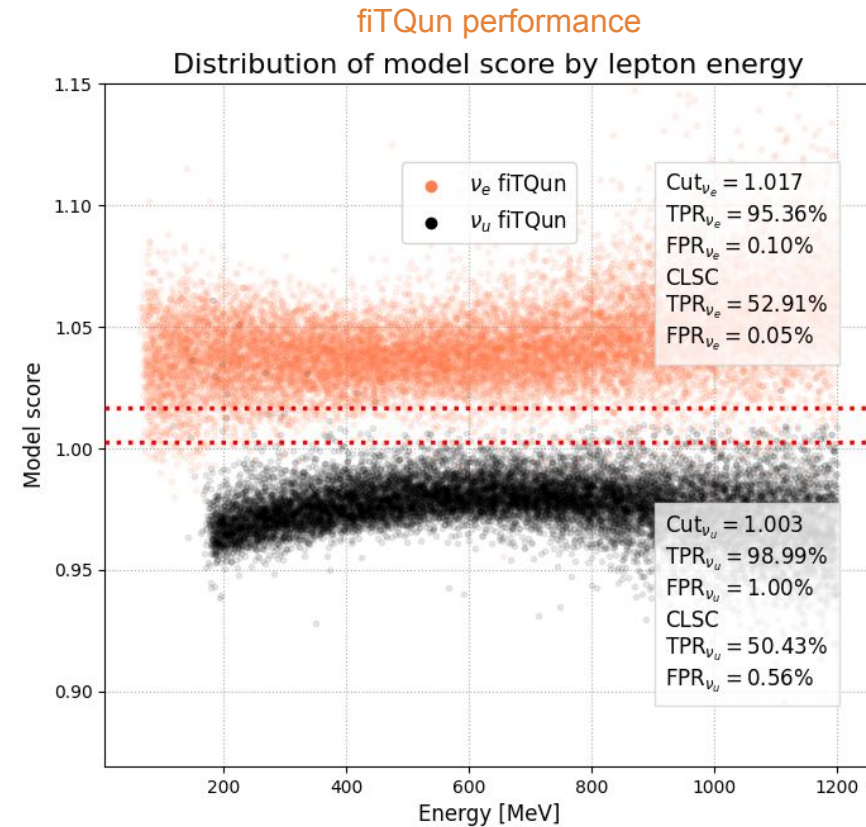
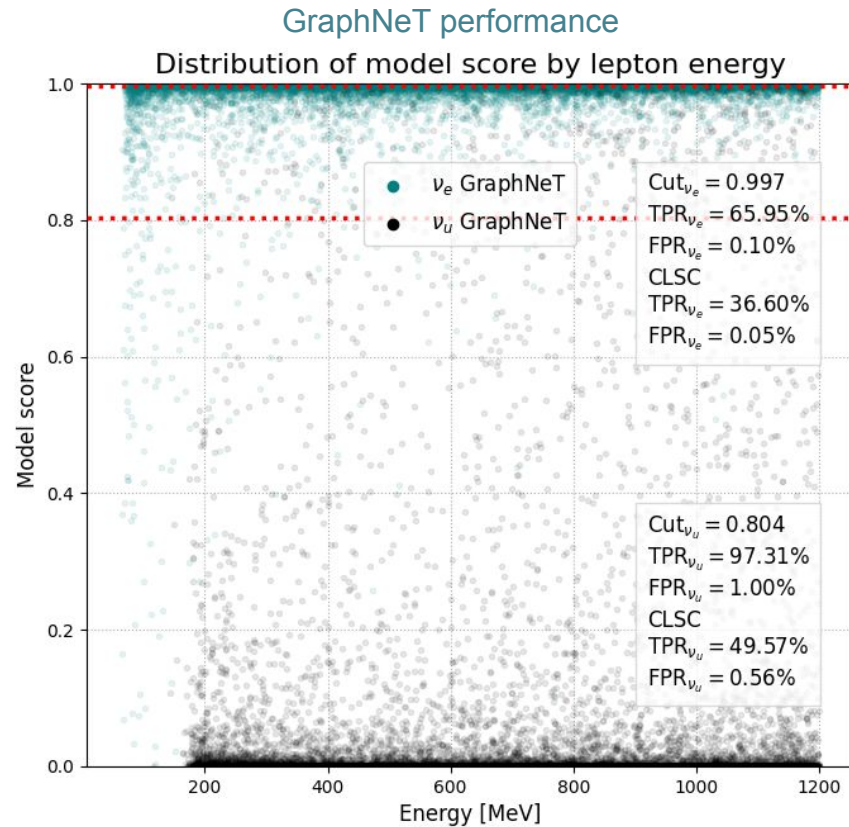
Event Classification Performance

With cuts



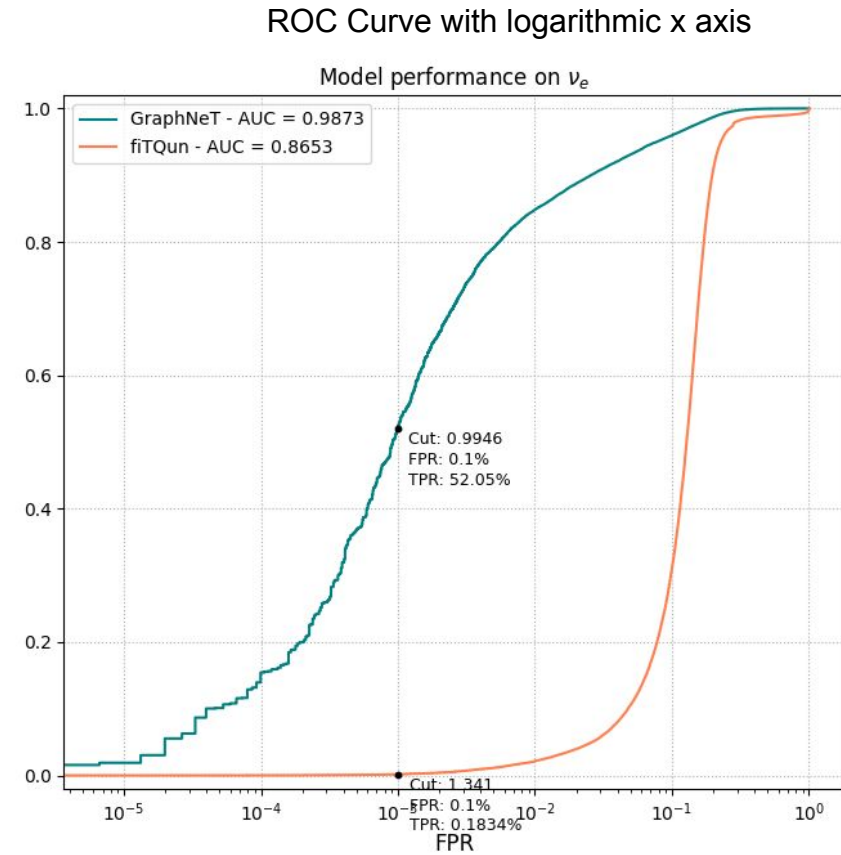
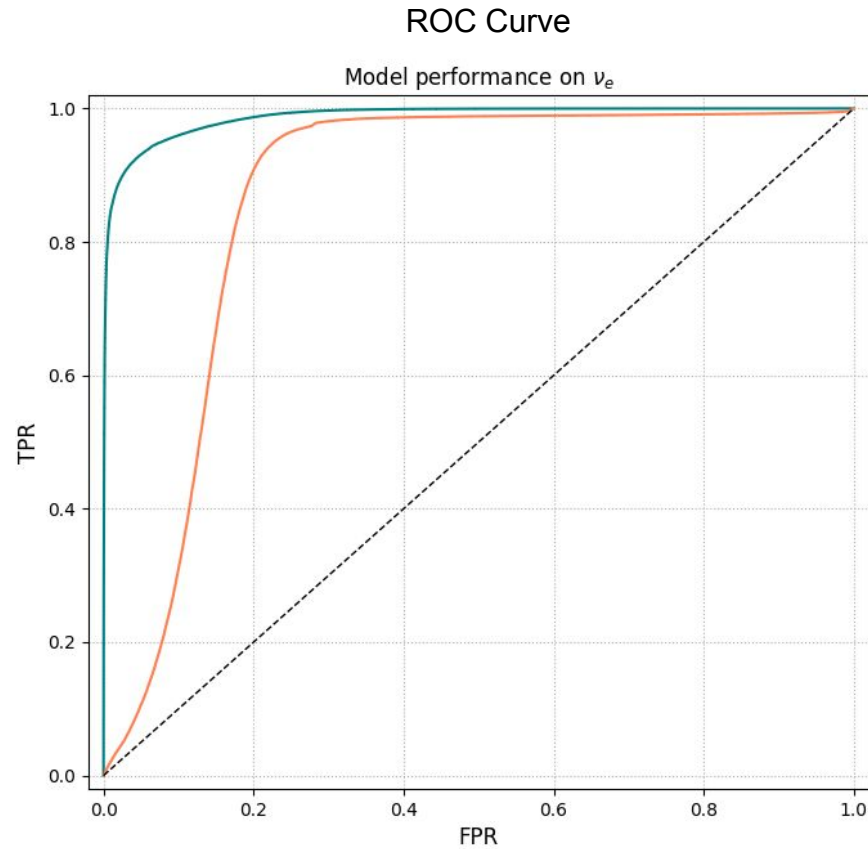
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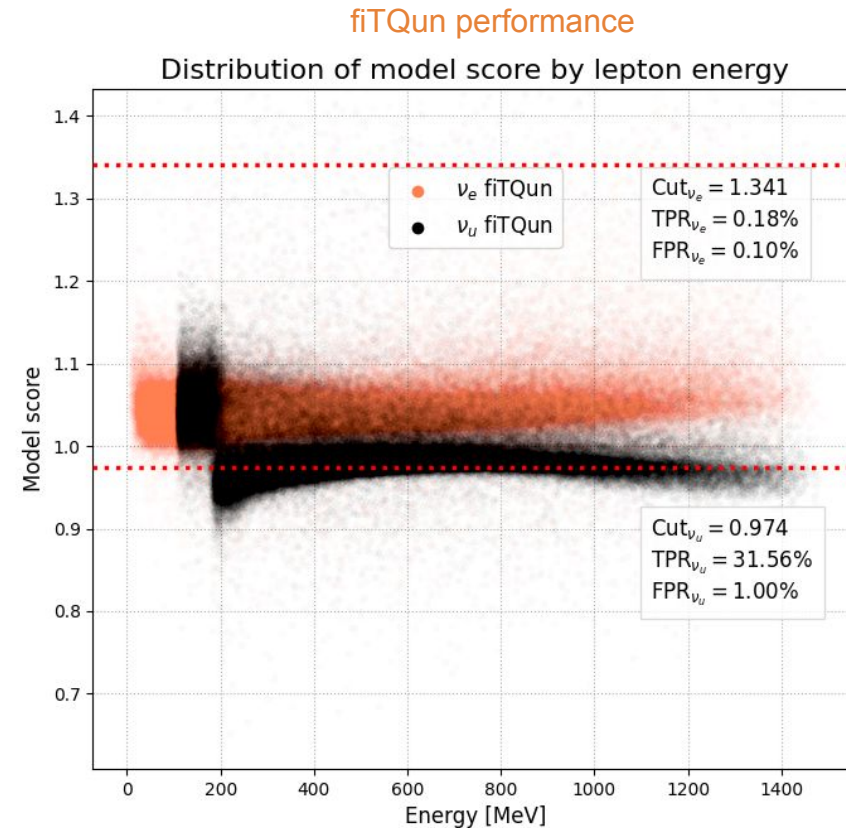
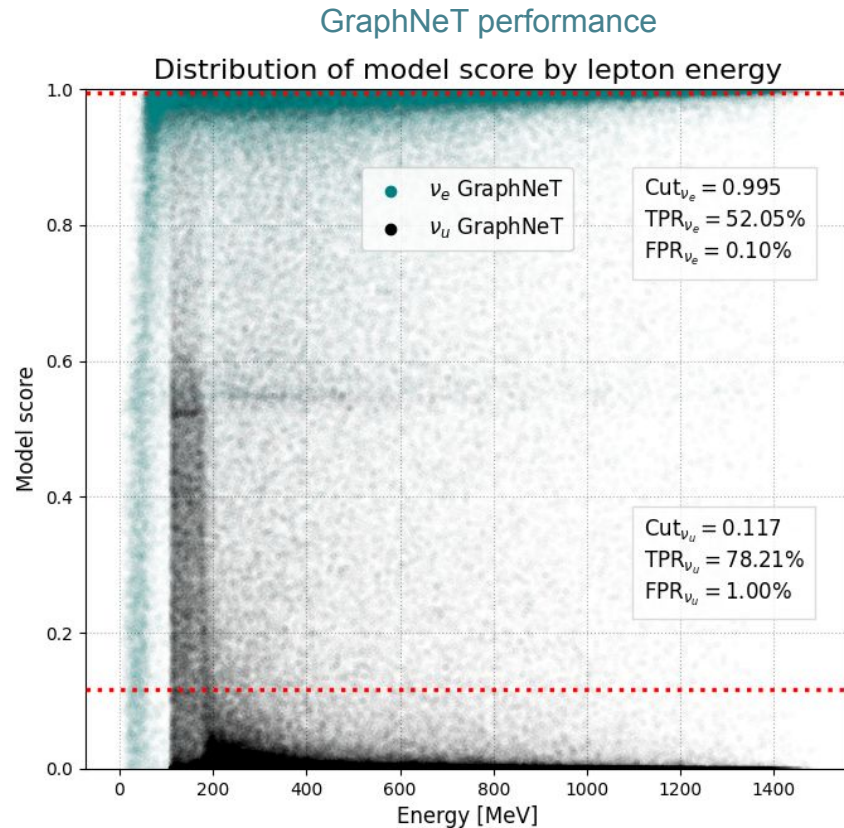
Event Classification Performance

Without data cuts



Event Classification Performance

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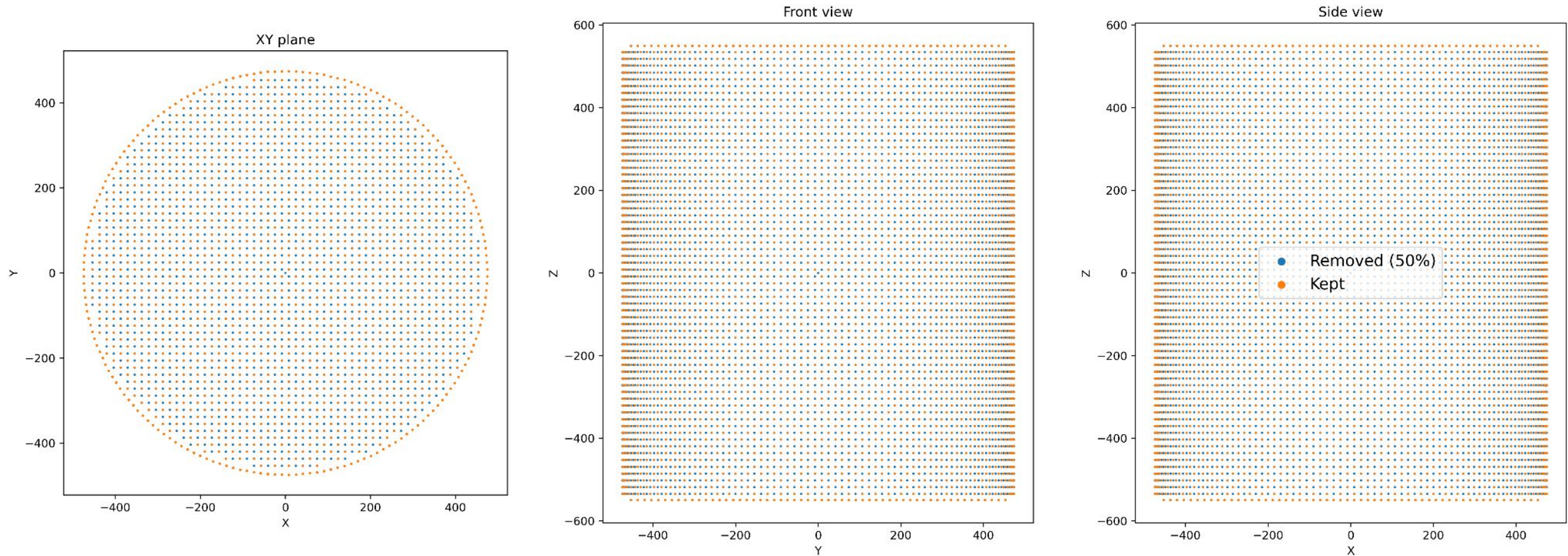


Event Classification Performance

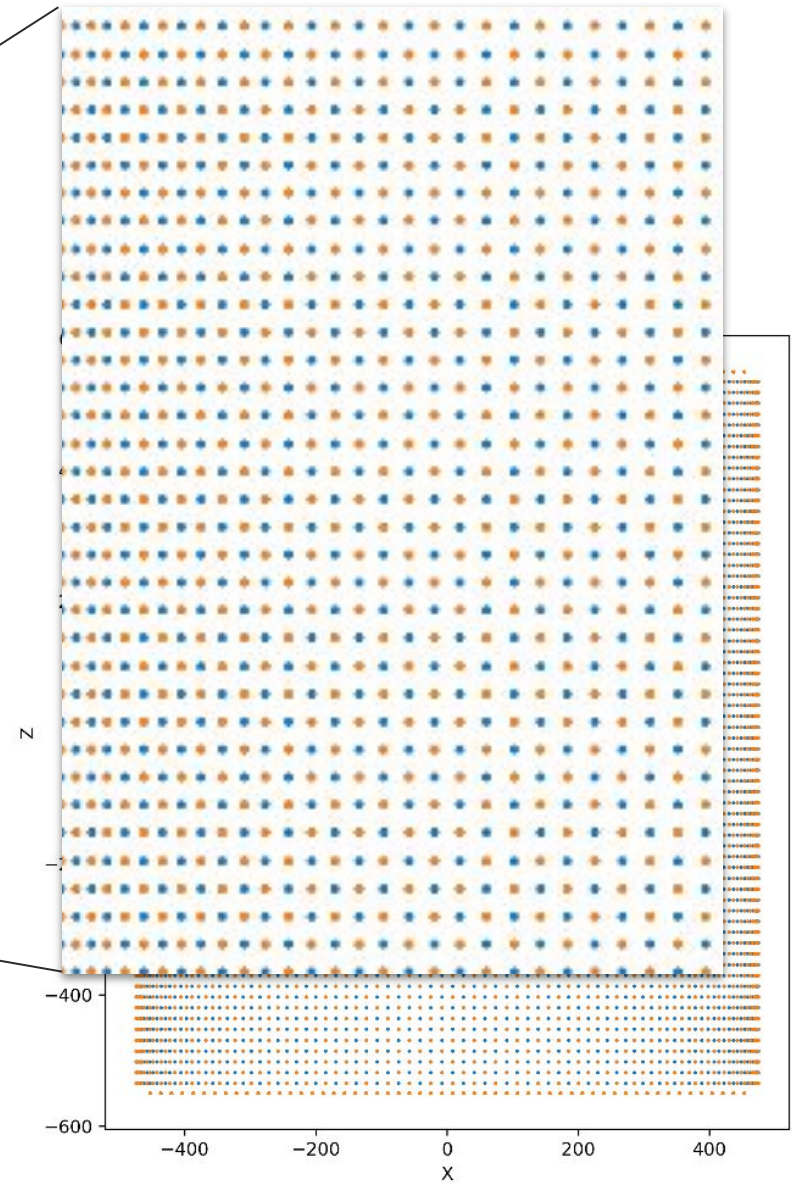
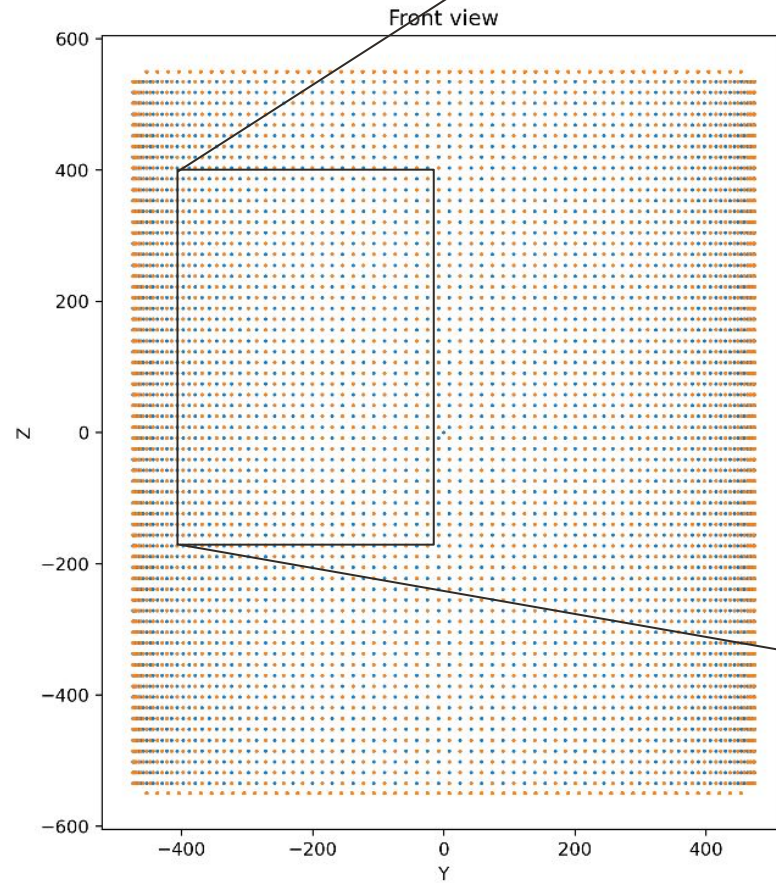
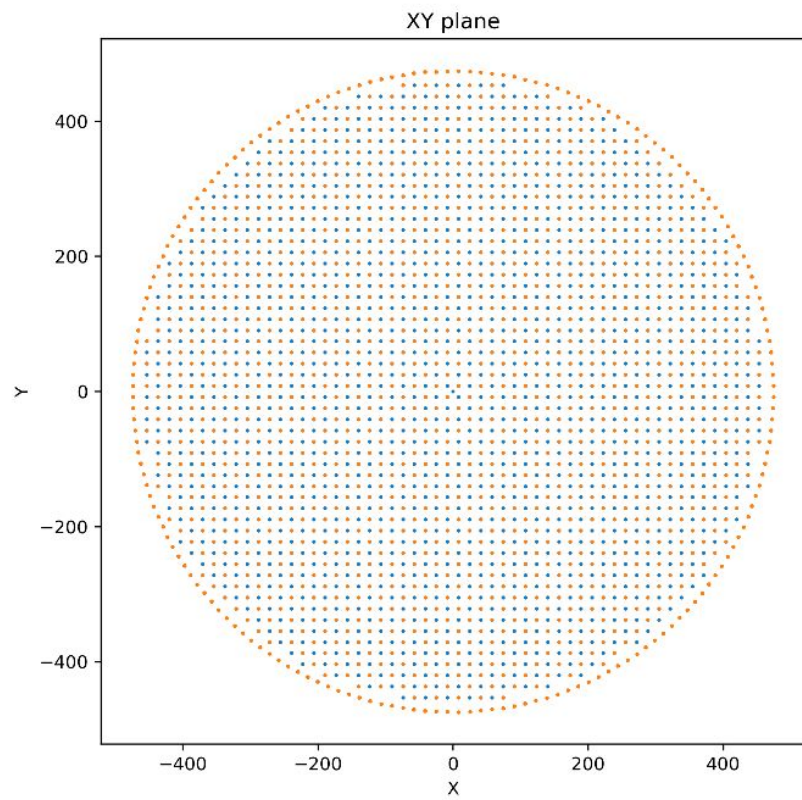
- The GNN has acceptable performance even on the full events
- Using the GNN, the data cuts can be made obsolete

Generalisation

Generalisation

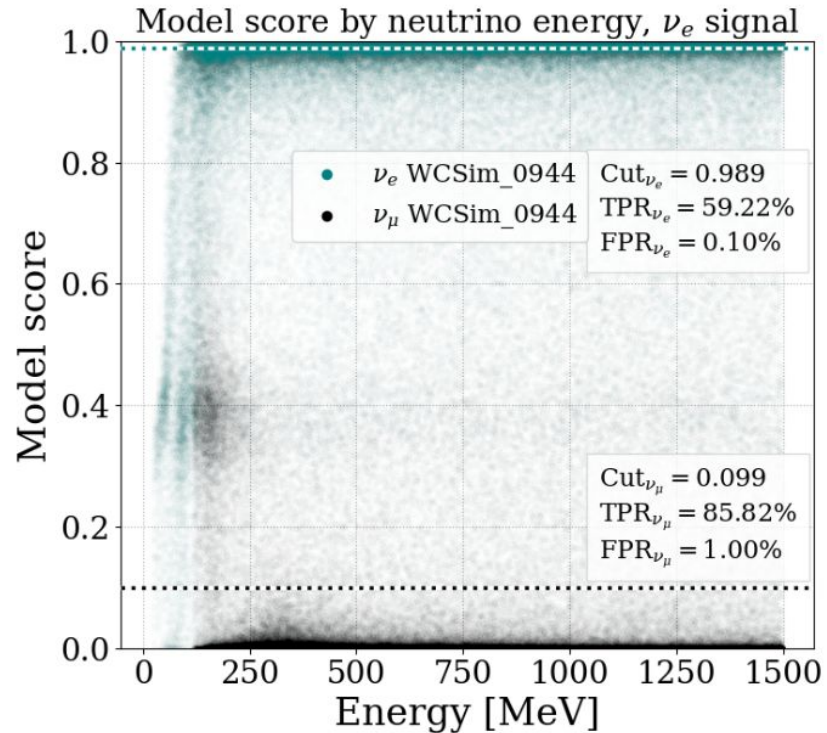


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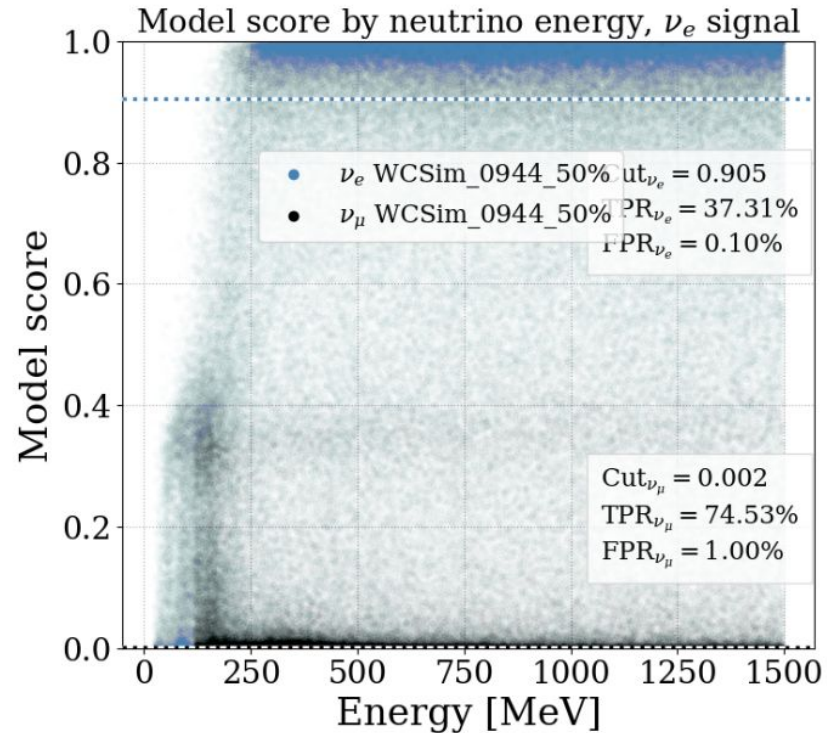


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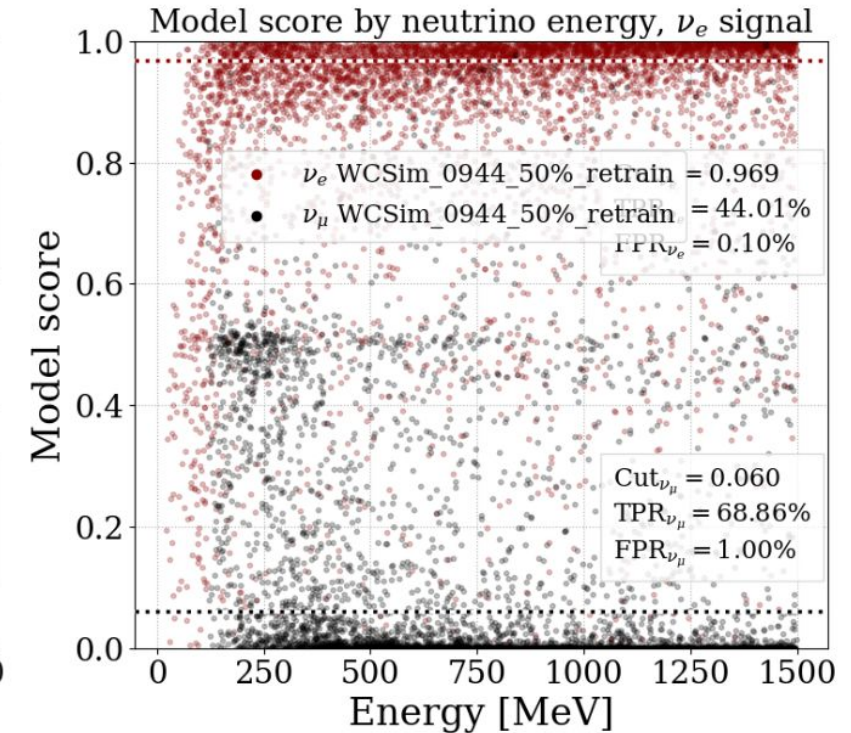
Original PMTs



50% PMTs
Original model

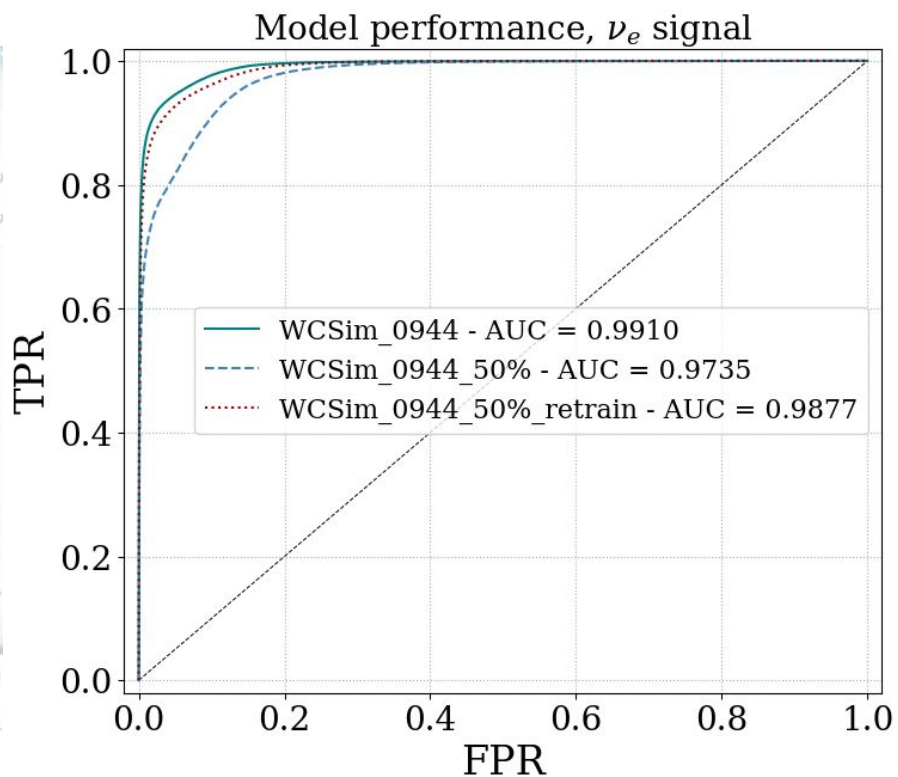
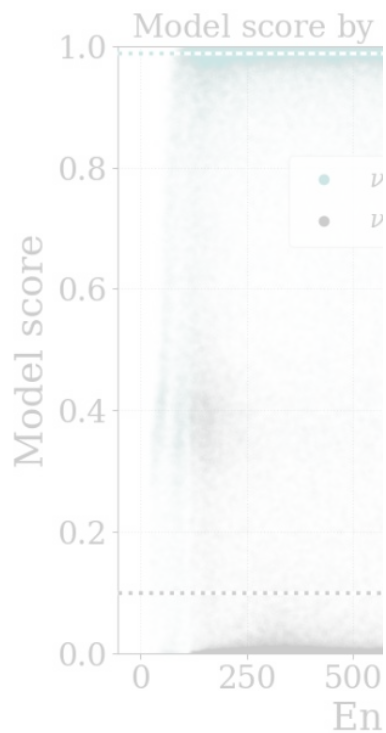


50% PMTs
Retrained model

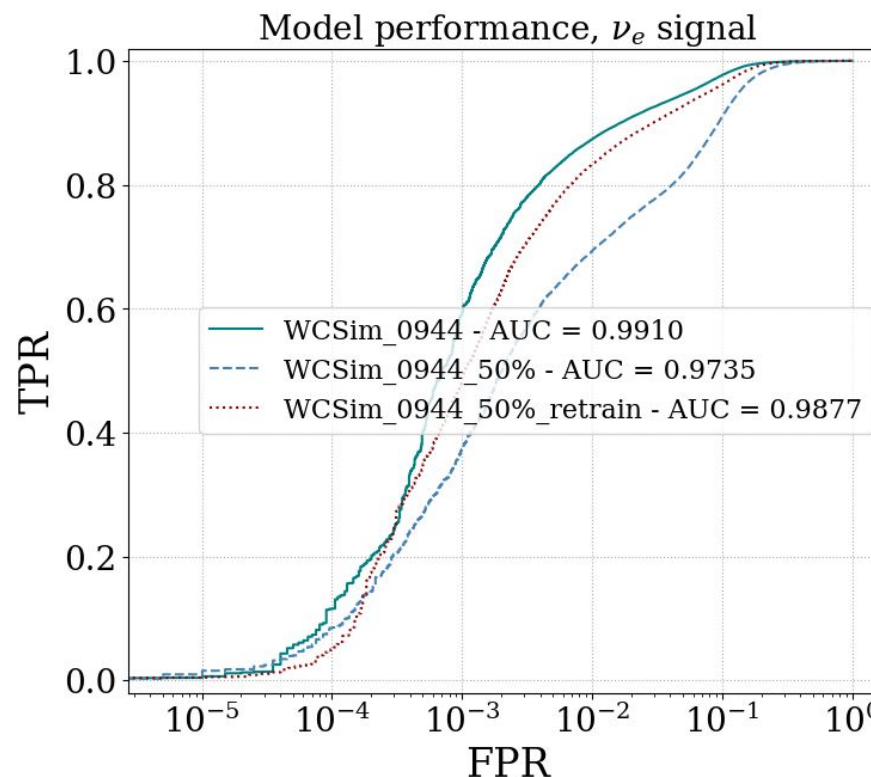


Generalisation

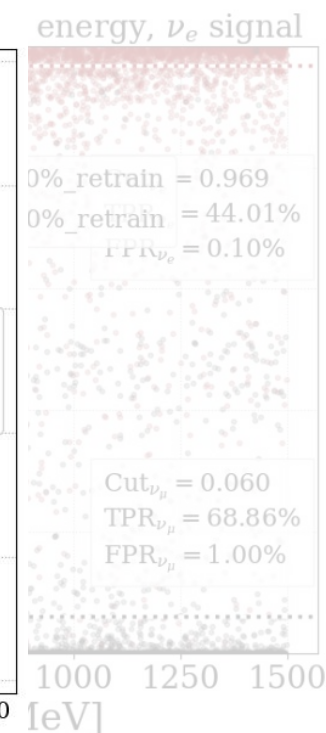
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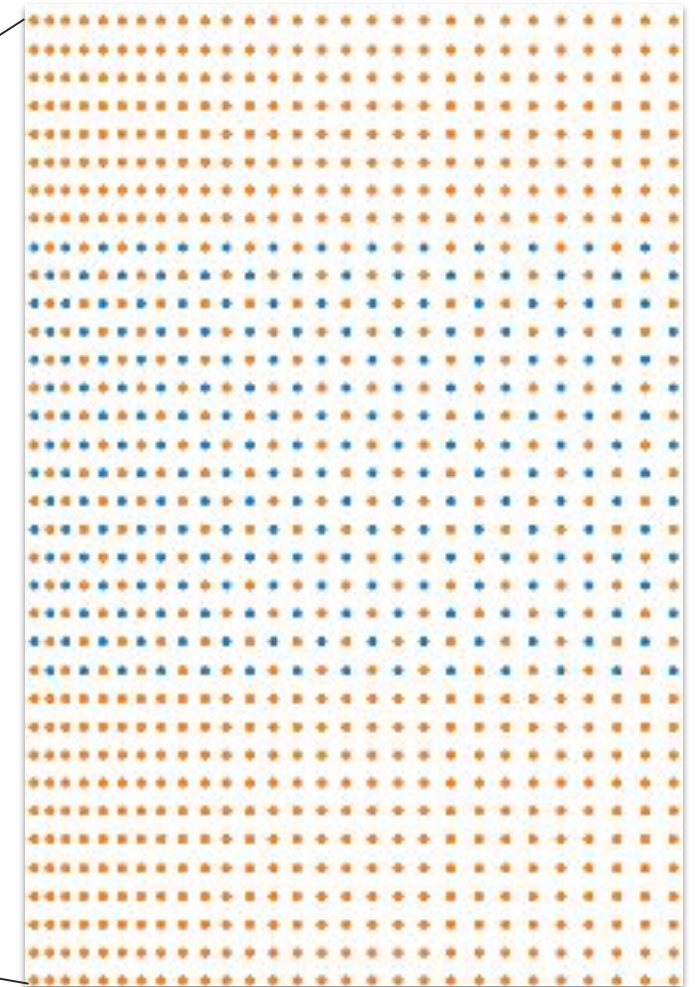
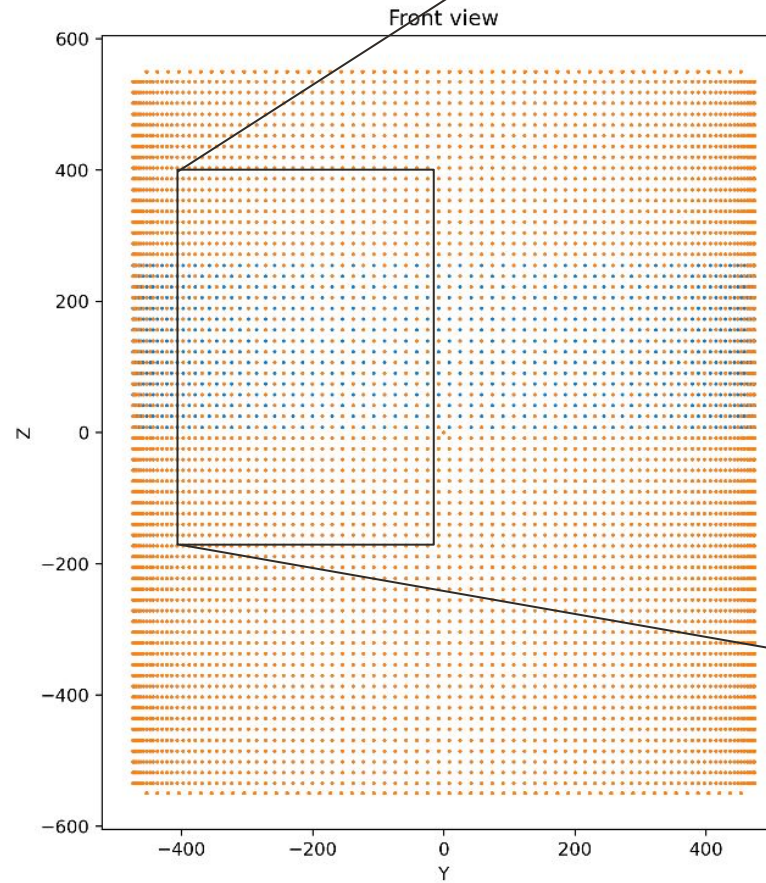
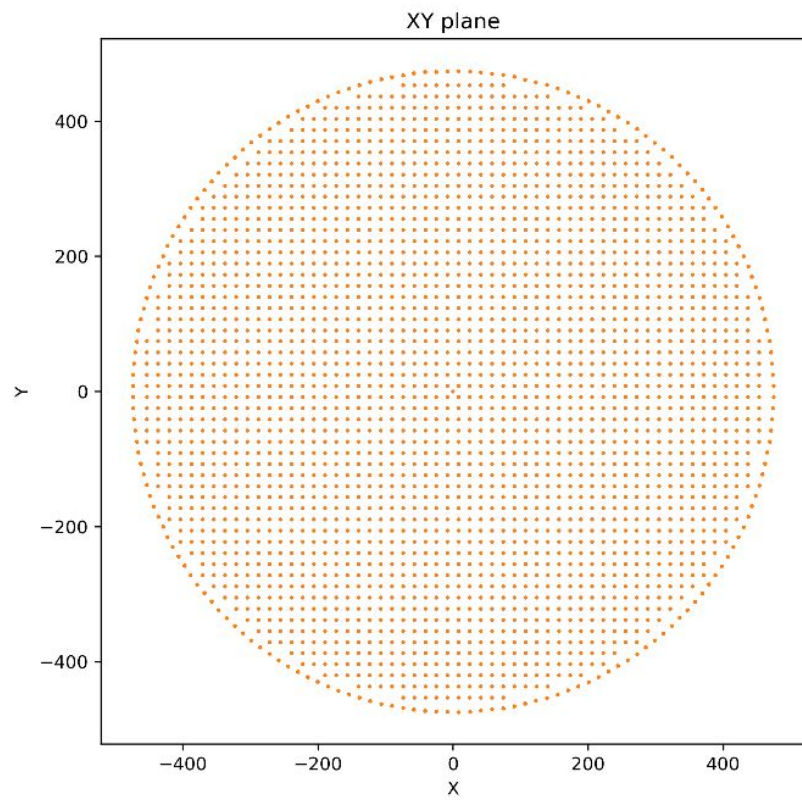
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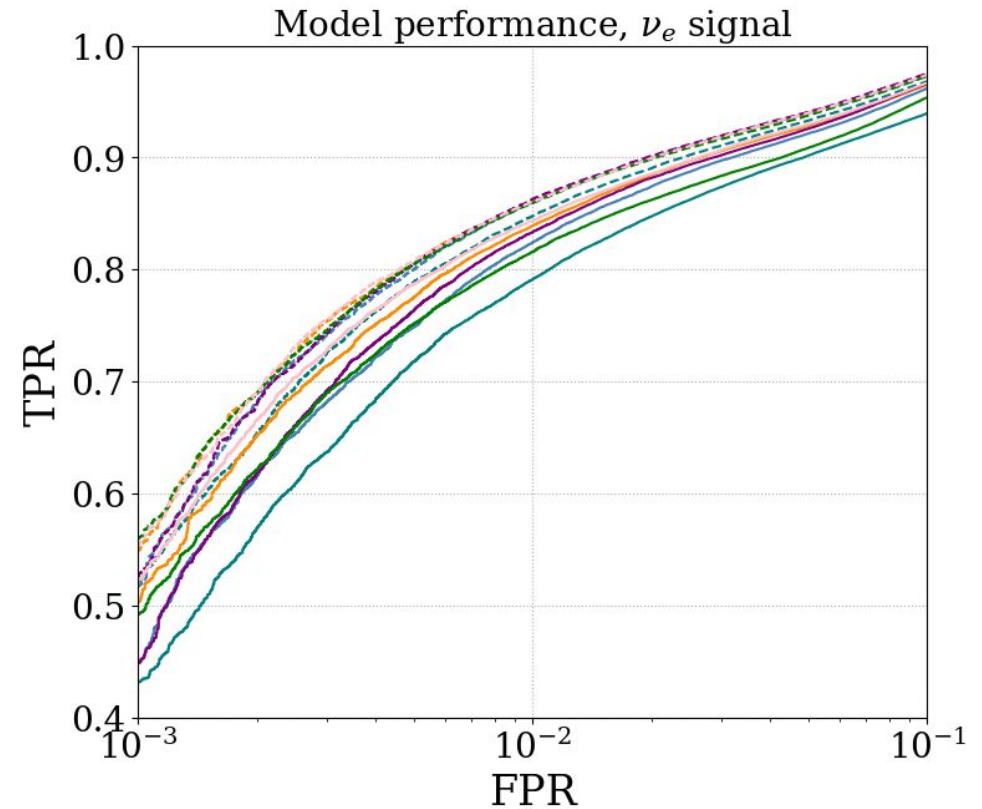
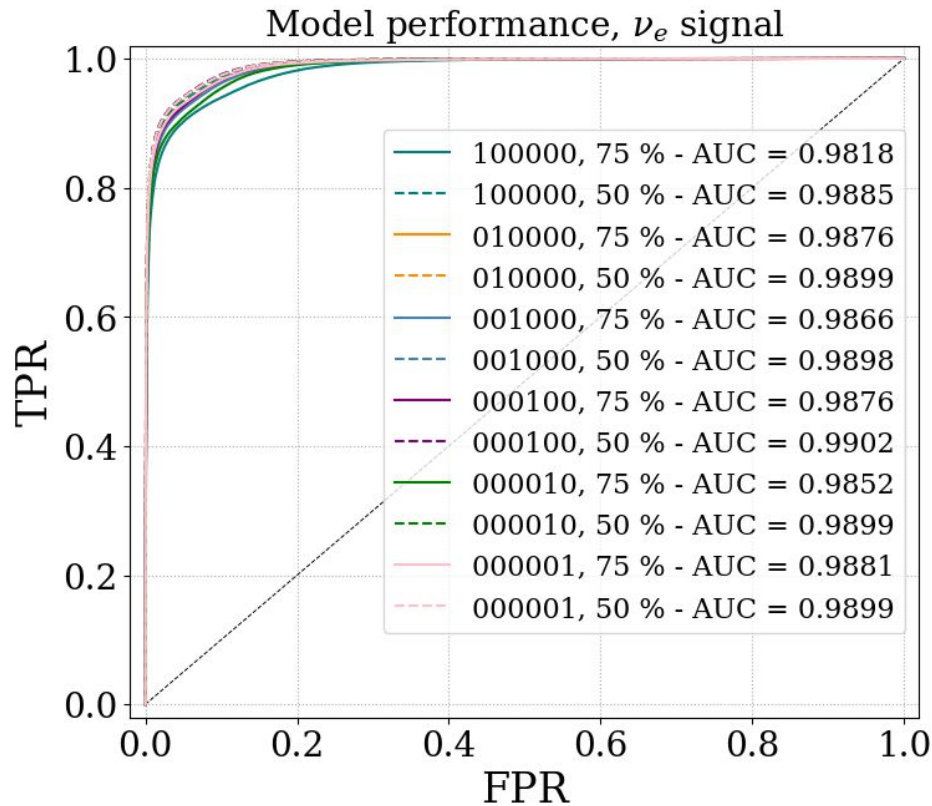
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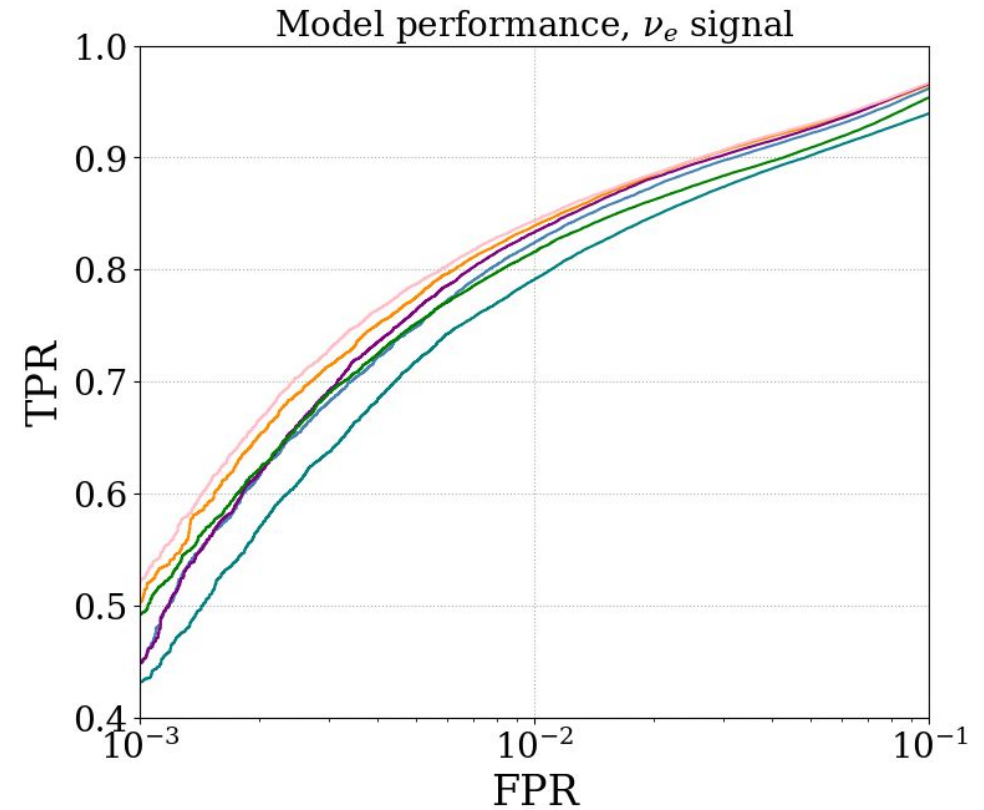
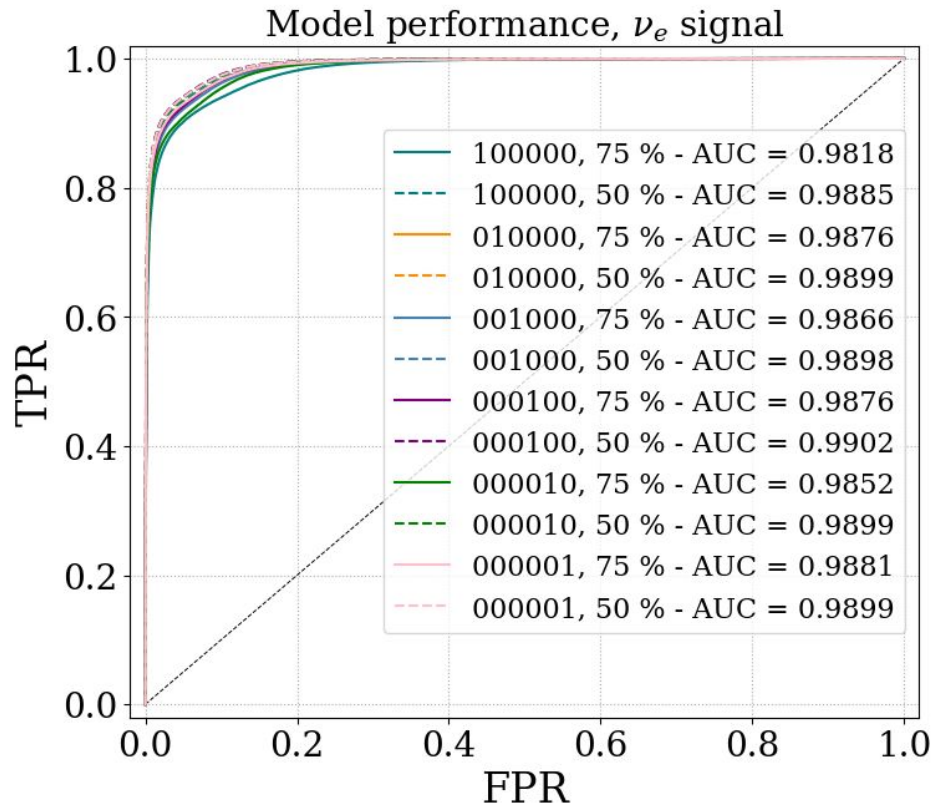


Generalisation



But which event topologies?

Generalisation



Summary

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

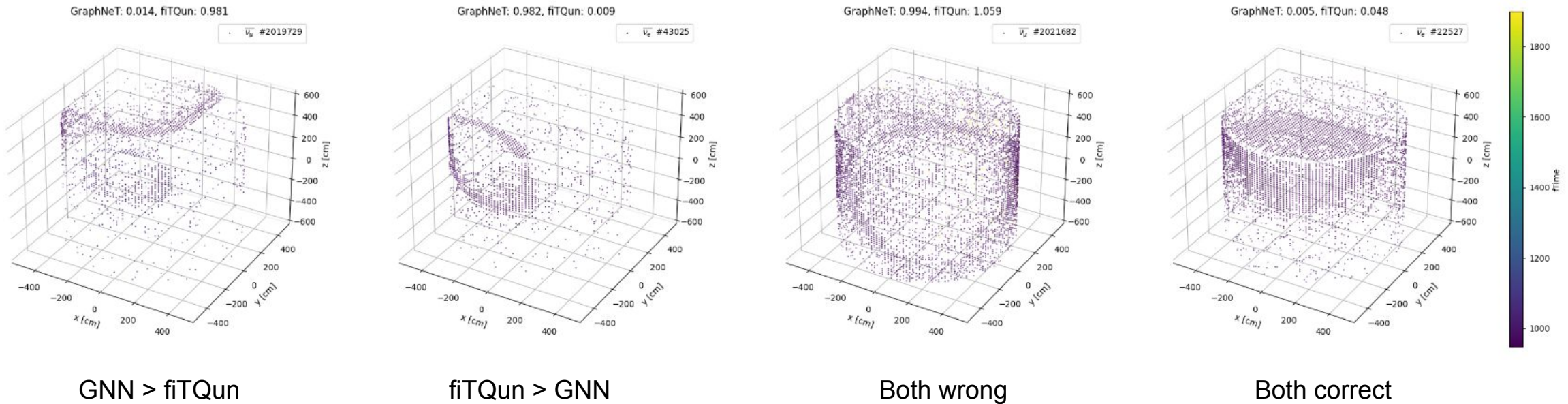
An aerial architectural rendering of a modern urban development. The scene features a large, circular, light-colored building with a green logo on its roof, situated in the center. To its left is a long, rectangular building with a flat roof. A winding blue river or canal flows through the landscape, curving around the central building and extending towards the bottom right. The surrounding area is a mix of green fields, yellow and green patches, and a dense cluster of smaller buildings in the upper left. A wide, multi-lane road or highway runs diagonally from the top right towards the bottom right. The overall style is a soft, painterly architectural visualization.

Thank you!

Backup Slides

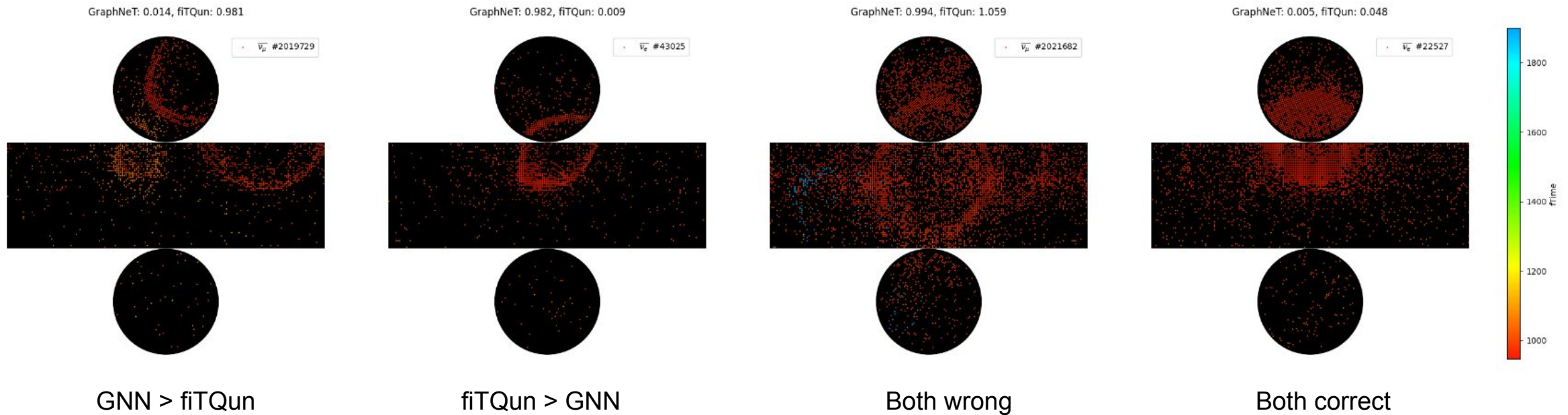
Factors impacting performance - multiple charged lepton signatures

Neutrino event simulations - **with** data cut

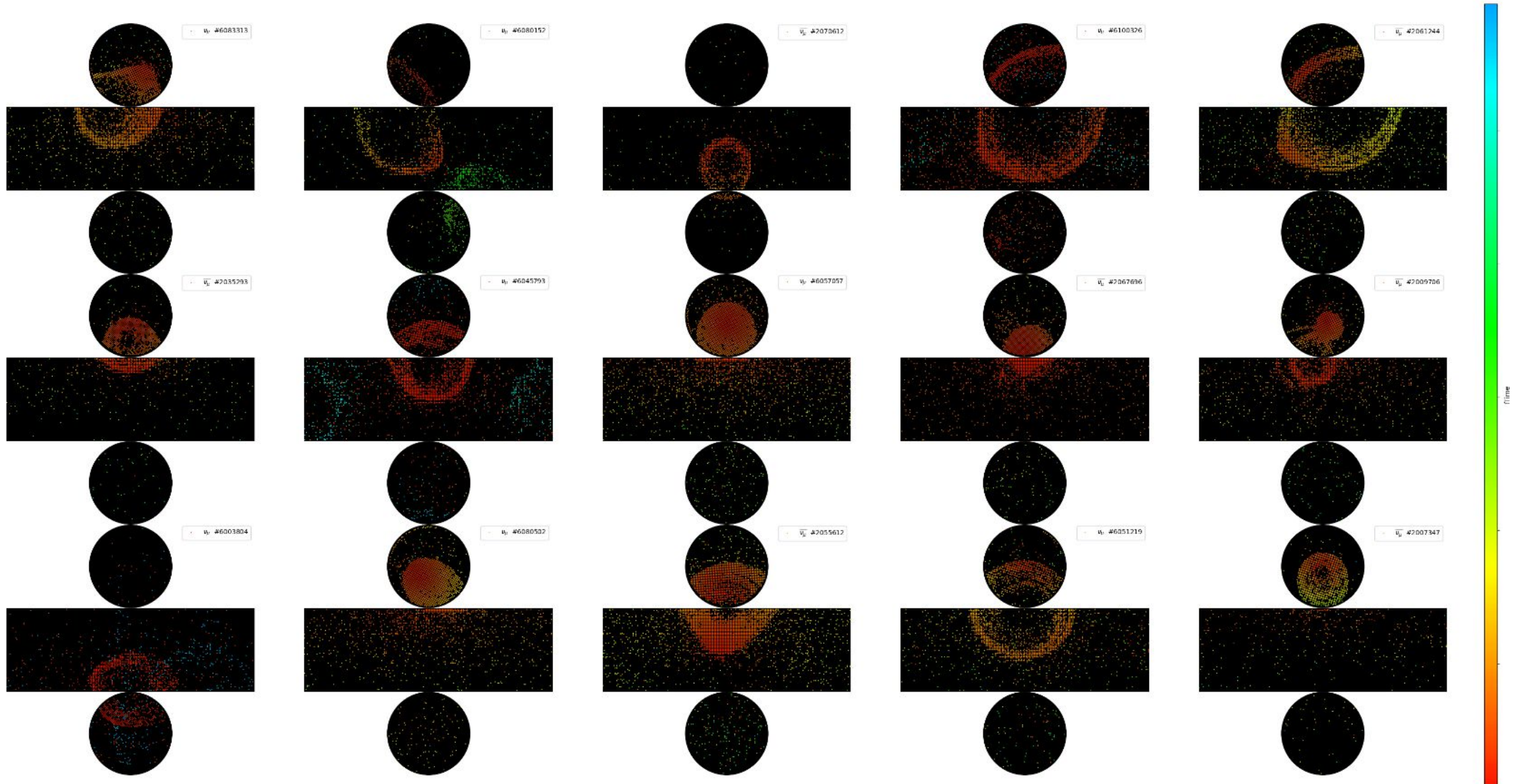


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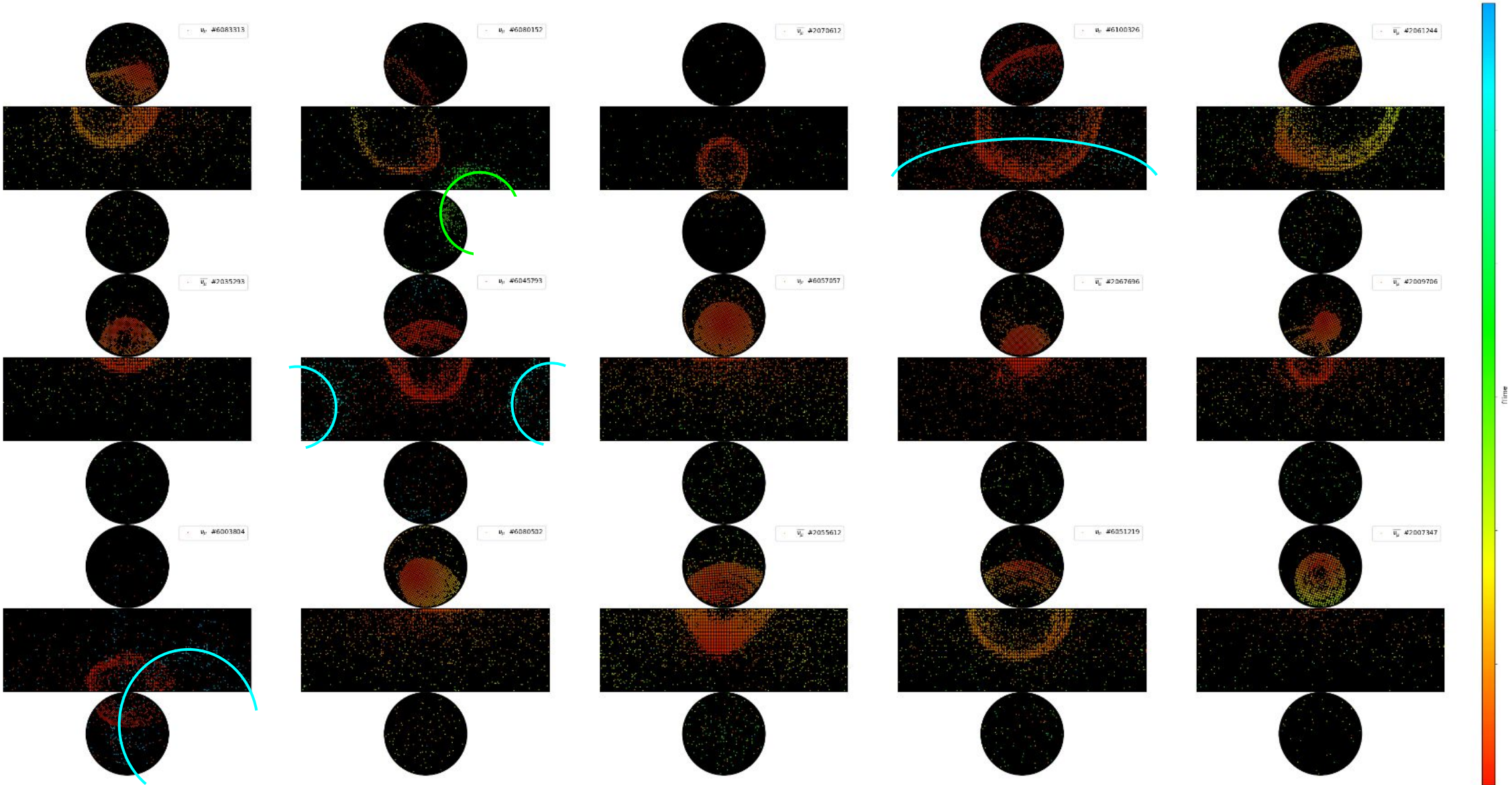
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Event for which GraphNeT performs significantly better than fiTQun



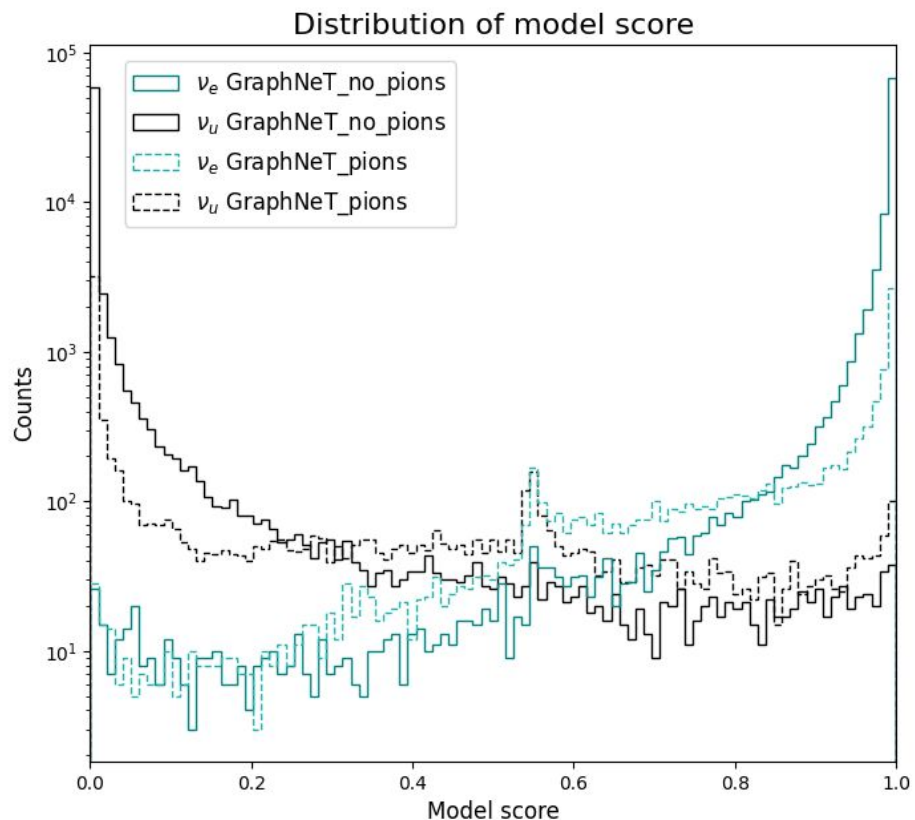
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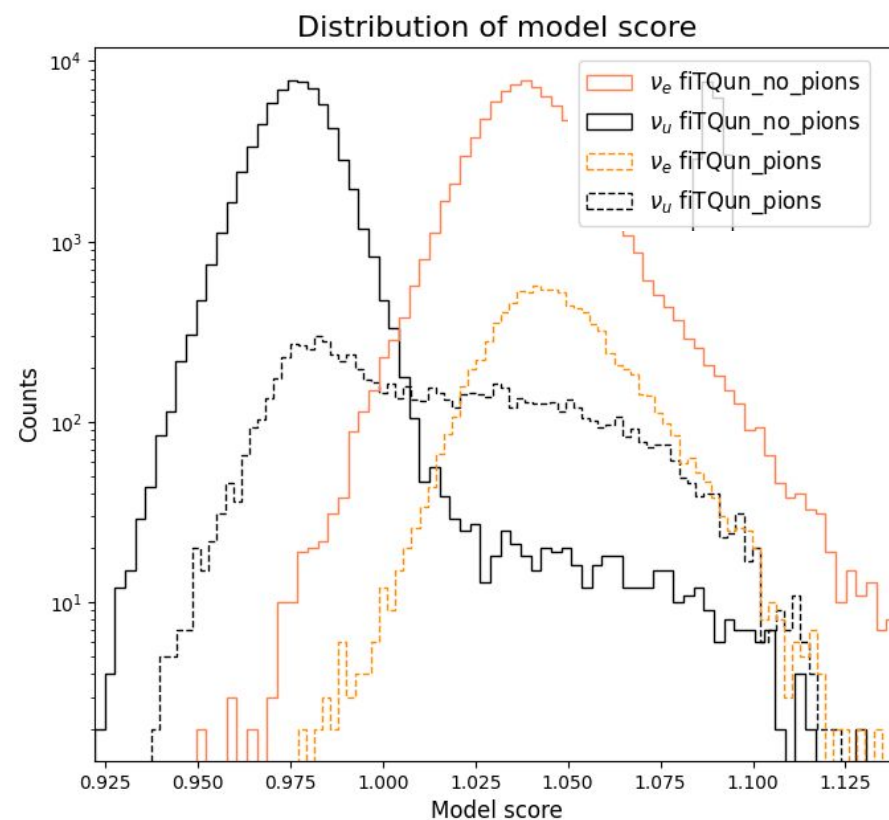
Factors impacting performance - pion creation

Neutrino event simulations - with data cut

GraphNeT performance with logarithmic x axis

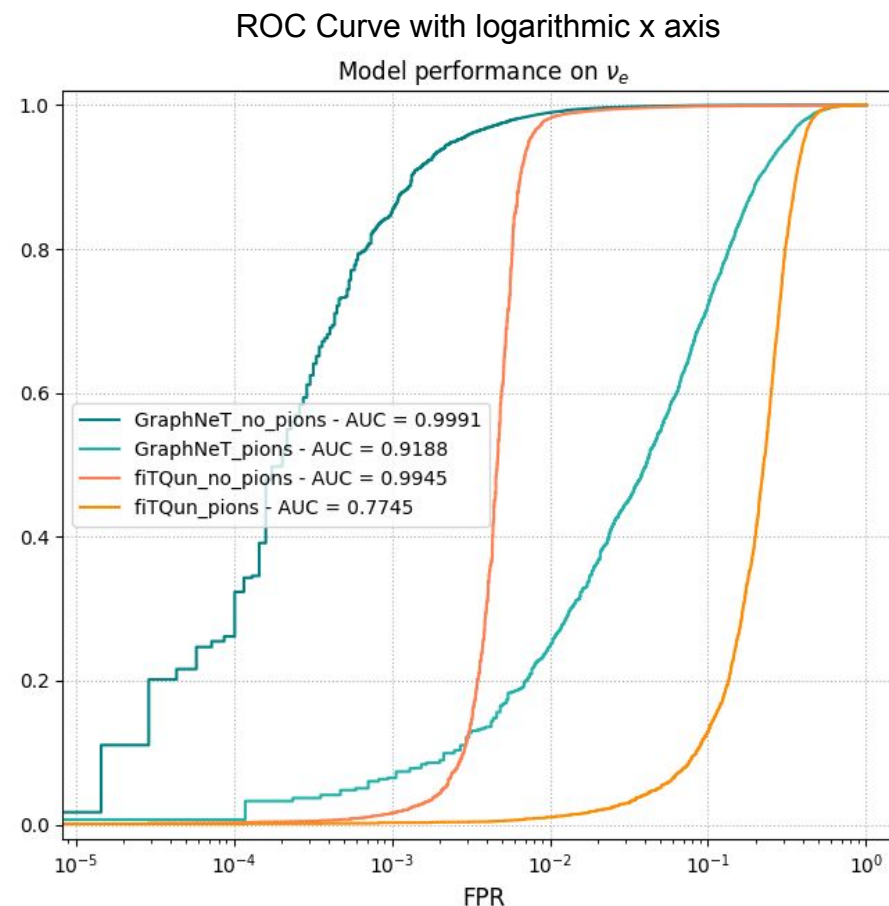
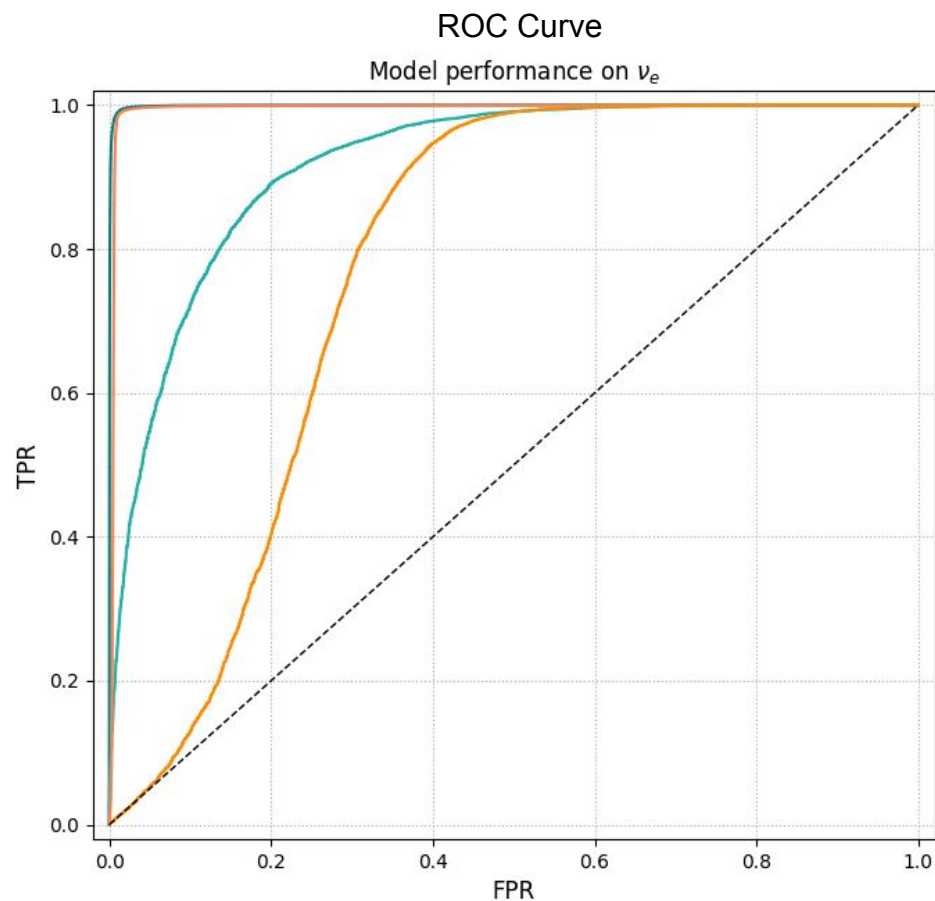


fiTQun performance with logarithmic x axis



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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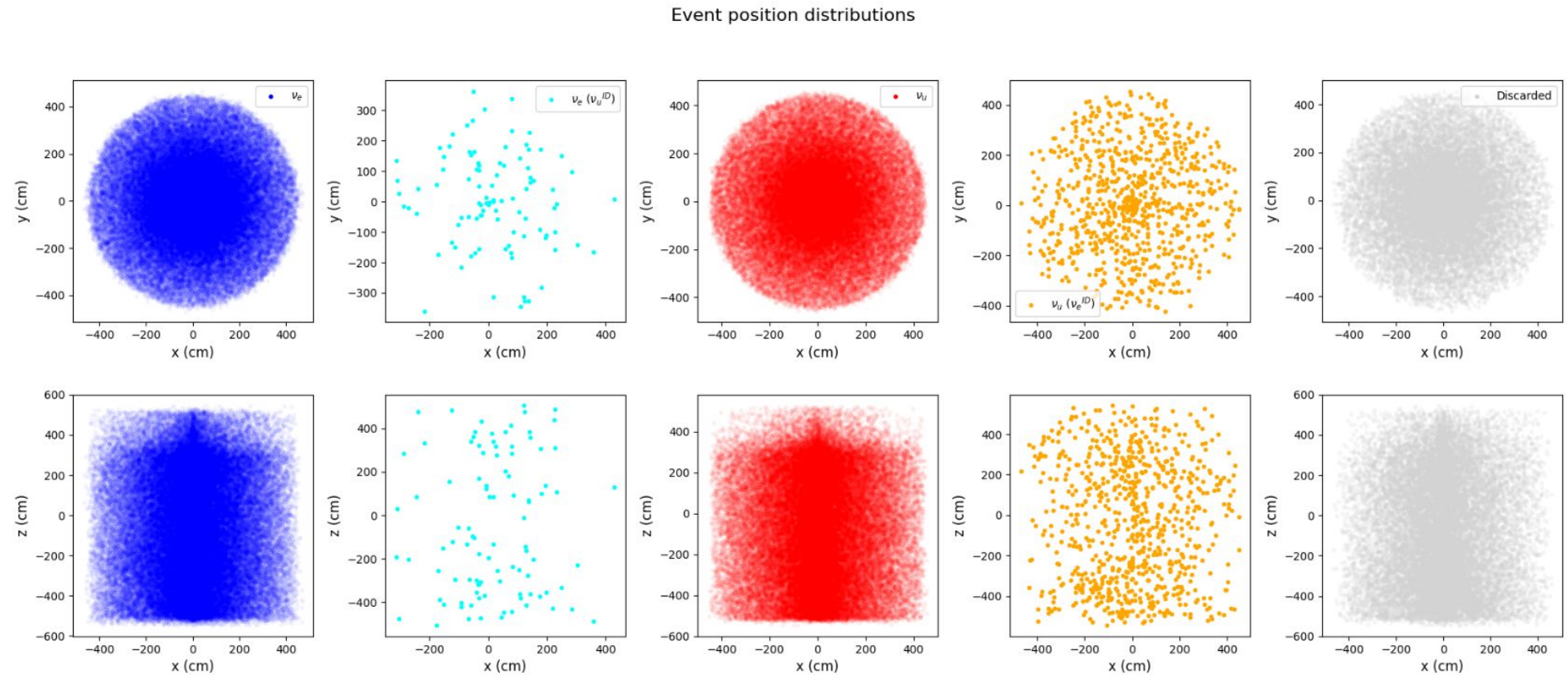
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We select samples that yield FPRs of:

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

Relation between interaction position and reconstruction performance

Neutrino event simulations - with data cut



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Neutrino event simulations - with data cut

