A large, circular, multi-tiered metal structure, likely a neutrino detector, is shown over a body of water. The structure is composed of a complex network of steel beams and supports. The sun is low on the horizon, creating a bright glow and reflecting on the water. The background shows a hazy landscape with hills and trees.

CHIPS Neutrino Event Characterisation Using Convolutional Neural Networks

Josh Tingey

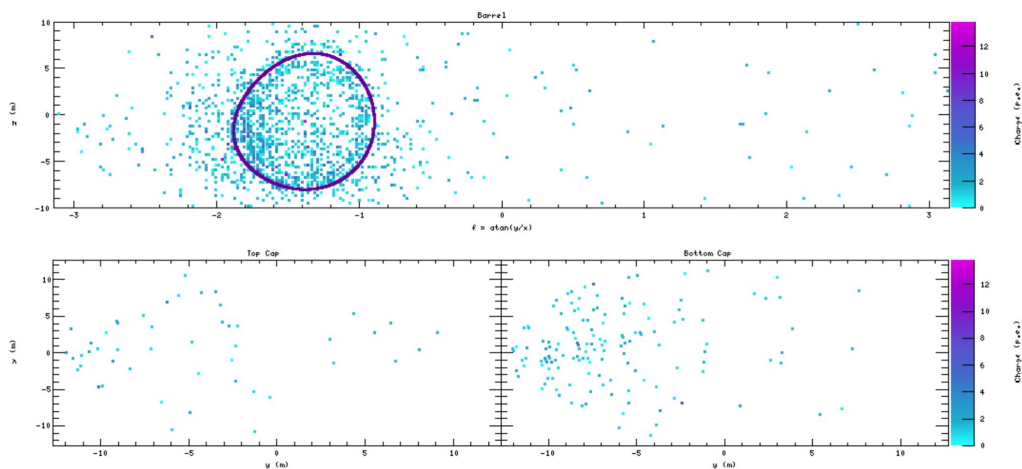
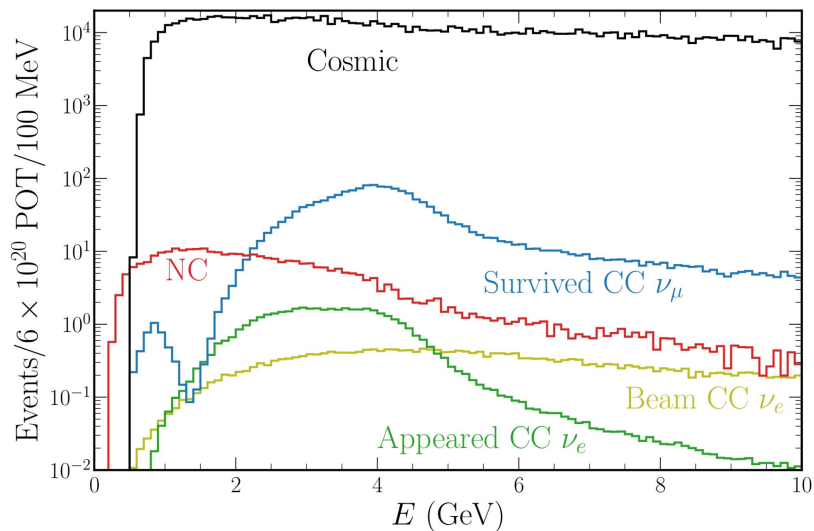
1. **The Problem:** Describe the problem we are trying to solve, how it has been solved in the past, and what the new approach entails (with a little bit of theory).
2. **The Solution:** Detail how we implement convolutional neural networks for CHIPS, highlighting essential things for other water Cherenkov detectors.
3. **The Results:** How well does it all perform, do we understand what is going on, and is it robust?

A large, circular steel structure under construction, possibly a stadium or arena, is shown at sunset. The structure is composed of a complex network of steel beams and trusses, forming a cylindrical shape. The sun is low on the horizon, creating a bright glow and casting long shadows. The structure is situated on a body of water, and the surrounding landscape is visible in the background.

The Problem

What problem are we trying to solve?

There are a wide range of neutrino interactions recorded within our detector. How do we determine useful properties about the interactions for use in neutrino oscillation analyses?



What tasks does this entail?

We want to efficiently select a pure beam CC ν_e sample from a sizeable beam CC ν_μ , beam NC, and cosmic muon background whilst accurately estimating the neutrino energy.

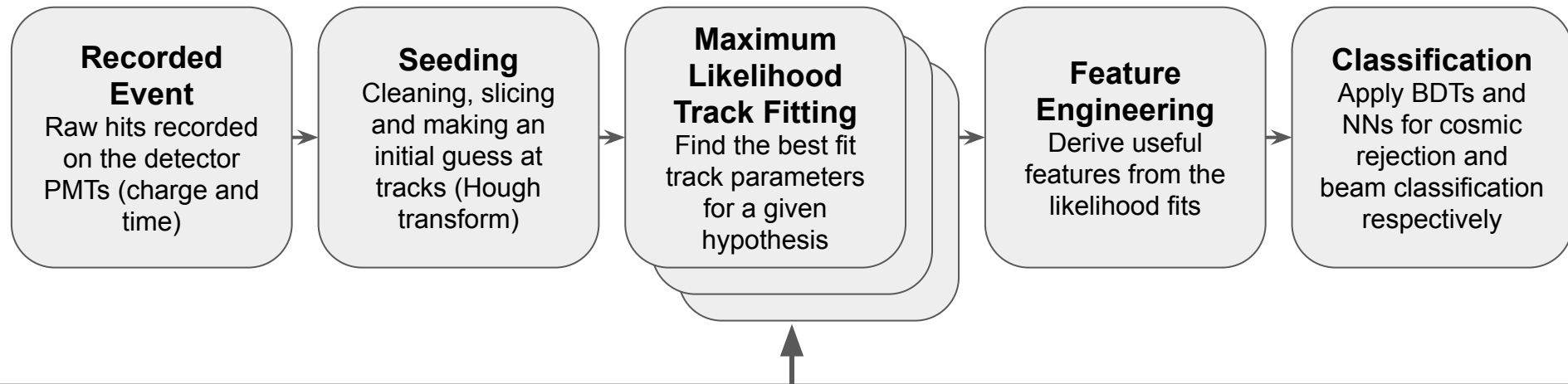
Primary Tasks:
(essential for physics)

- Cosmic event rejection
- Beam event classification
- Neutrino energy estimation

Secondary Tasks:
(nice to have)

- Event containment
- Vertex estimation
- Lepton energy estimation

How did we solve this problem?



- Dependent on a finite set of inputs that must be implemented in software.
- If a physics process is overlooked then the algorithm has access to a reduced amount of information.
- It can be a large amount of effort to implement and validate.
- Requires a predefined hypothesis, each of which can take a while to run.

Is a different approach possible?

As neutrino experiments tend to effectively record an 'image' of each event, modern computer vision algorithms can be used. Such as Convolutional Neural Networks (CNNs).



ICECUBE
SOUTH POLE NEUTRINO OBSERVATORY

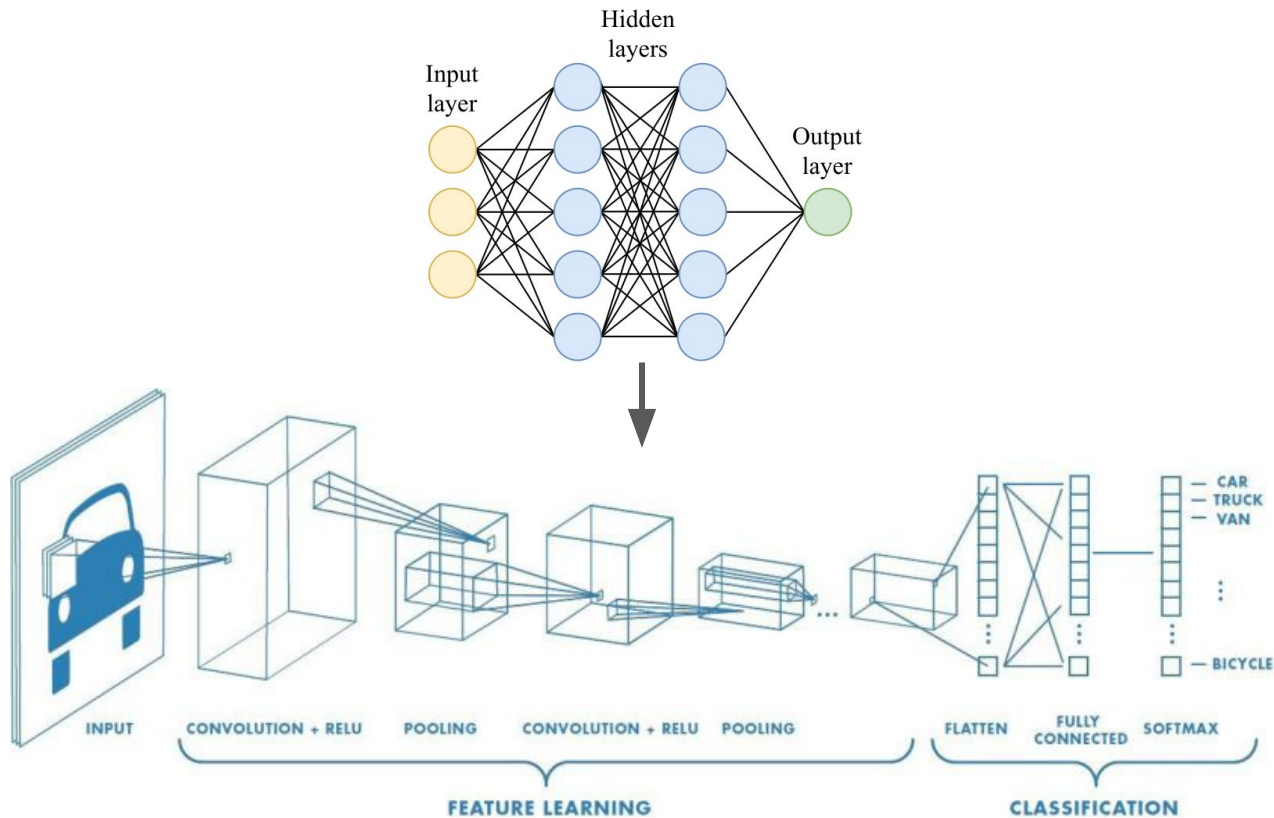


WatChMaL.org

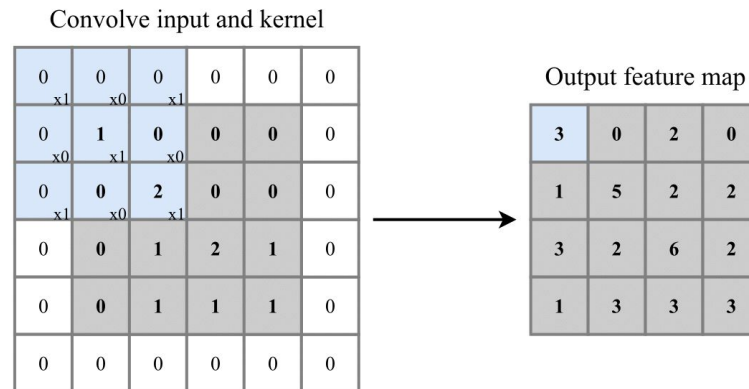
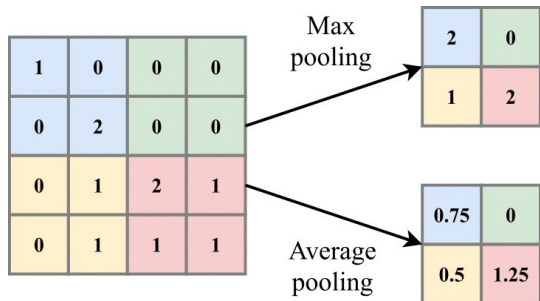
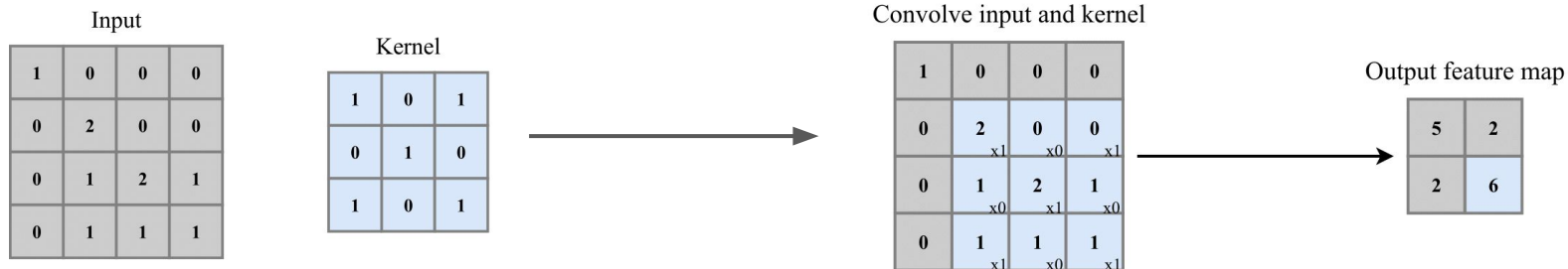


Mainly LArTPC detectors

What are convolutional neural networks?



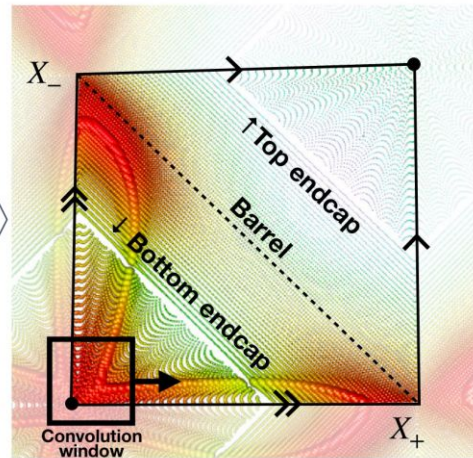
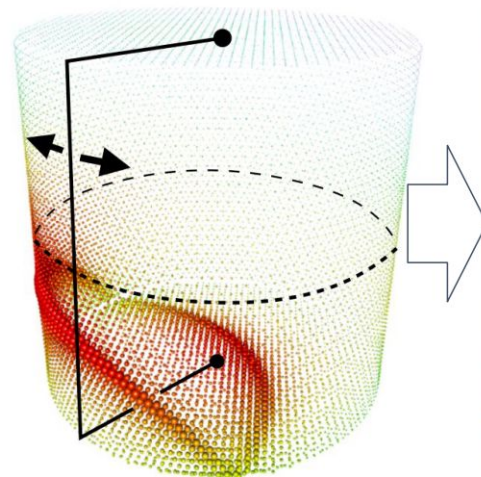
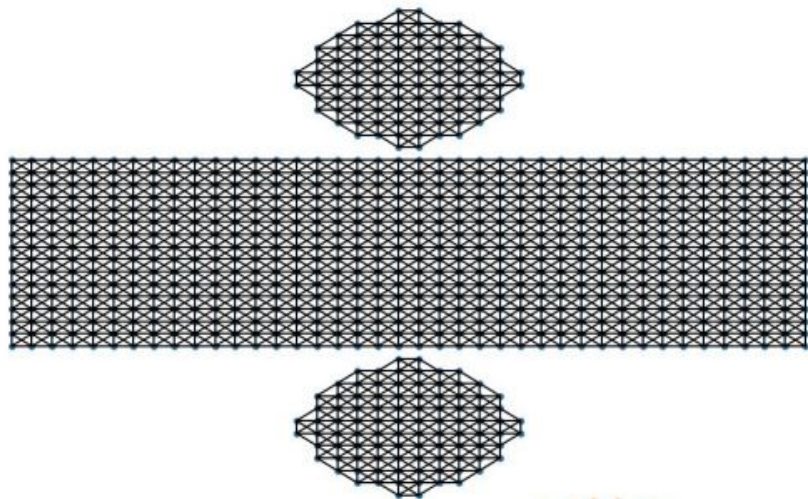
How do they work?



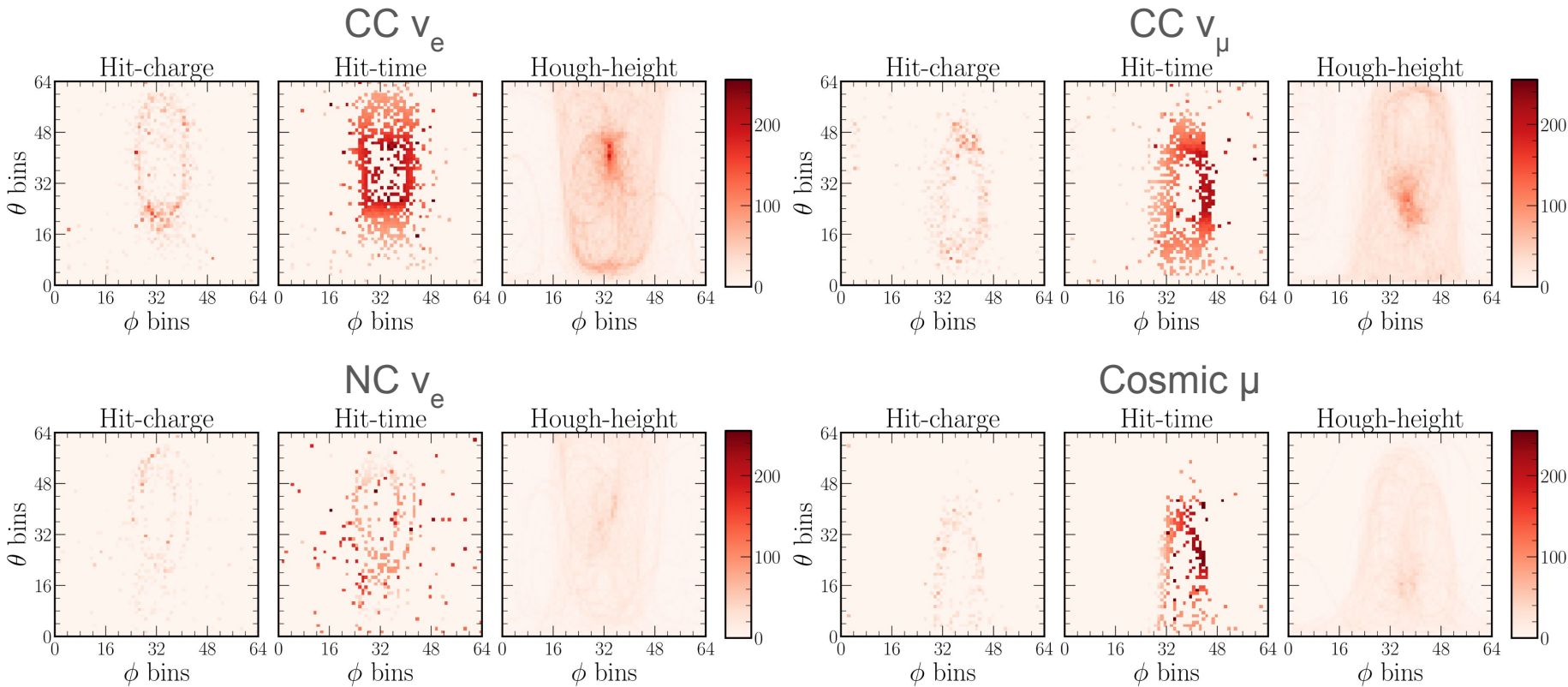
A large, circular steel structure under construction, situated on a body of water. The structure is composed of a complex network of steel beams and trusses, forming a cylindrical shape. The sun is low on the horizon, creating a bright glow and reflecting on the water. The background shows a hilly landscape with trees.

The Solution

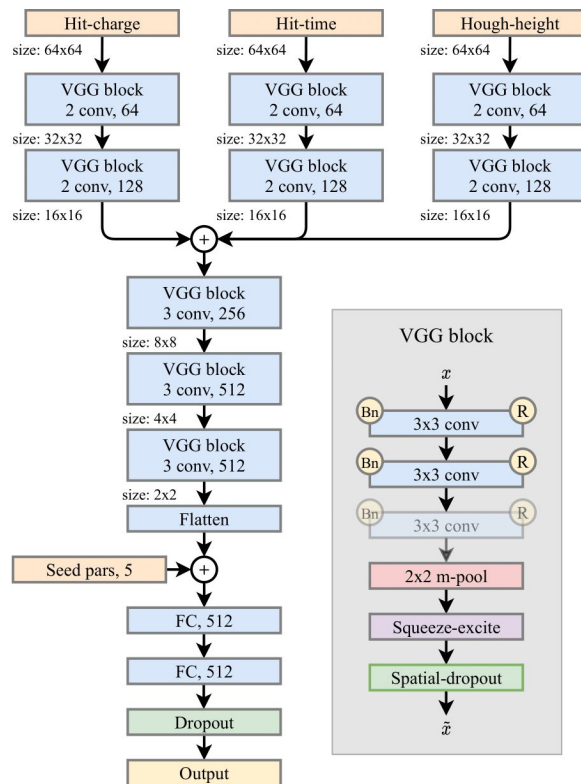
What is the main challenge?



Example inputs



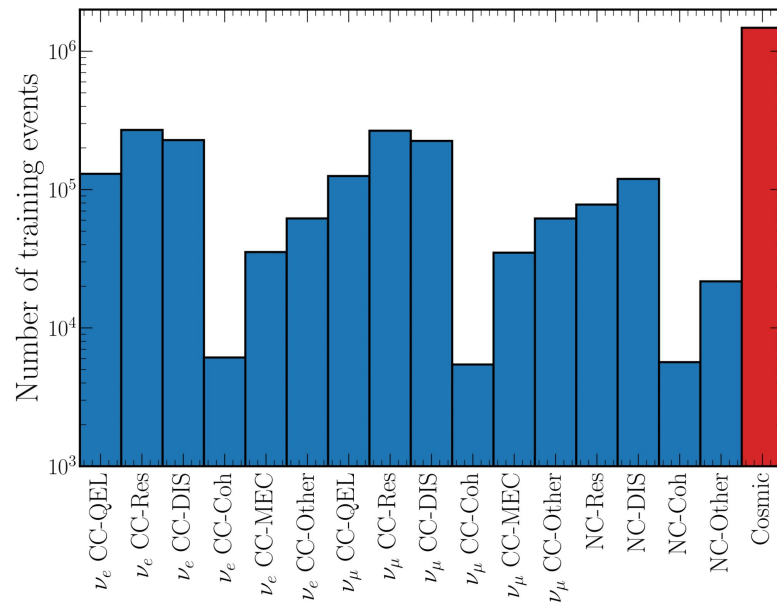
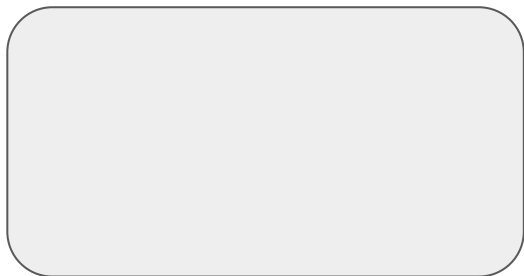
Baseline model architecture



Outputs + Training



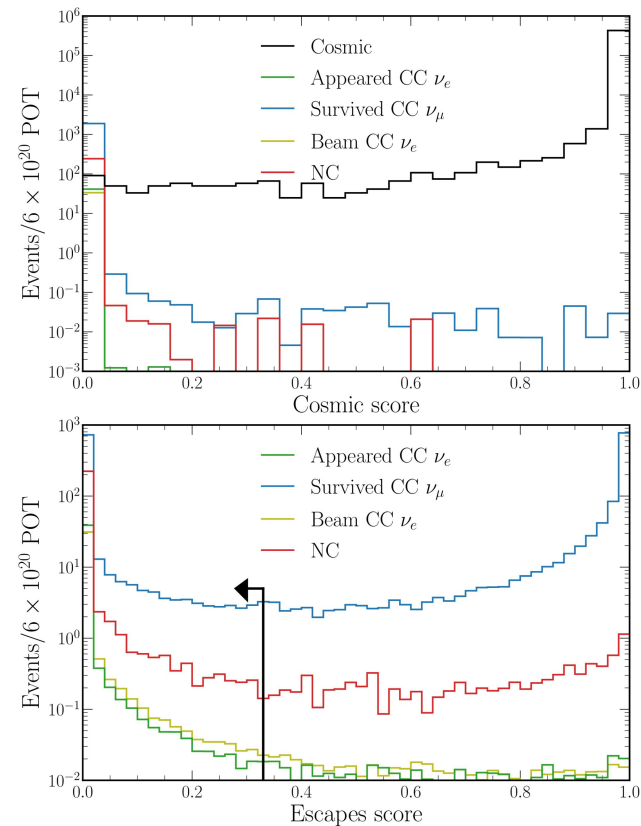
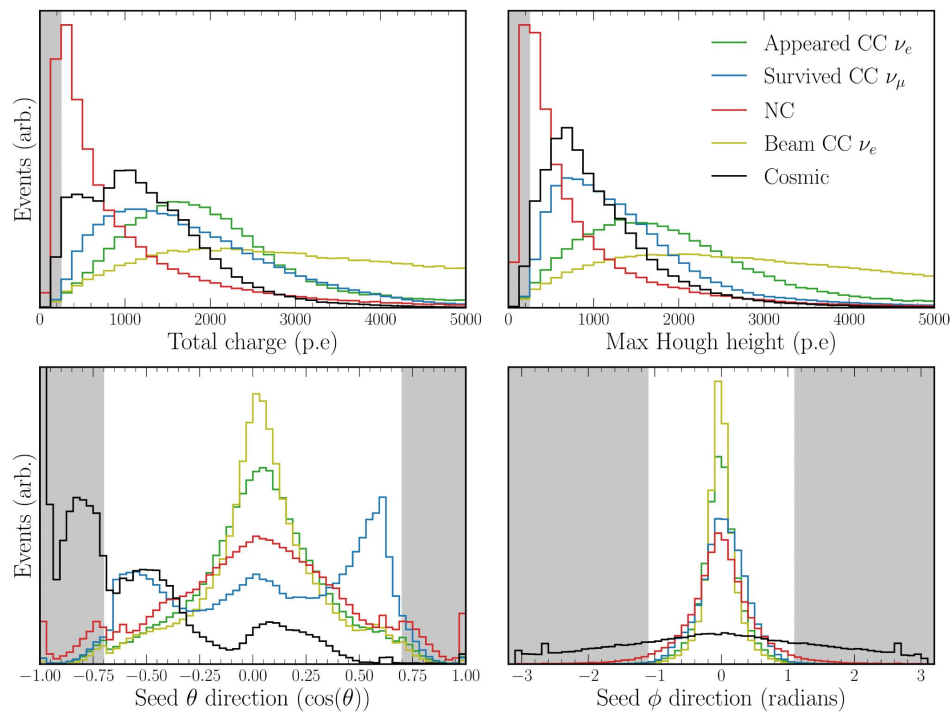
Outputs + Training



A large, circular steel structure under construction, situated on a body of water. The structure is composed of a complex network of steel beams and trusses, forming a cylindrical shape. The sun is low on the horizon, creating a bright glow and reflecting on the water. The text "The Results" is overlaid in the center of the image.

The Results

Preselection cuts



Preselection performance

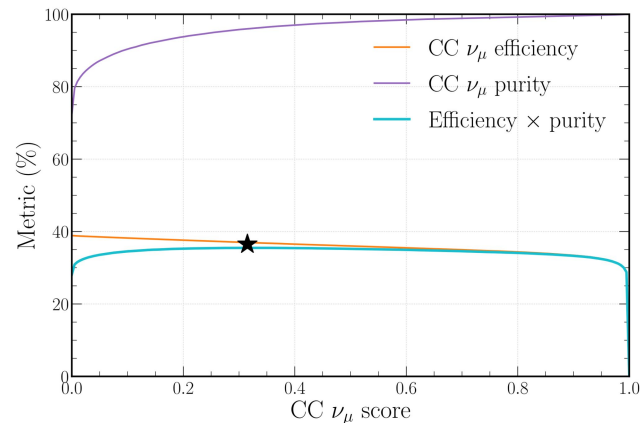
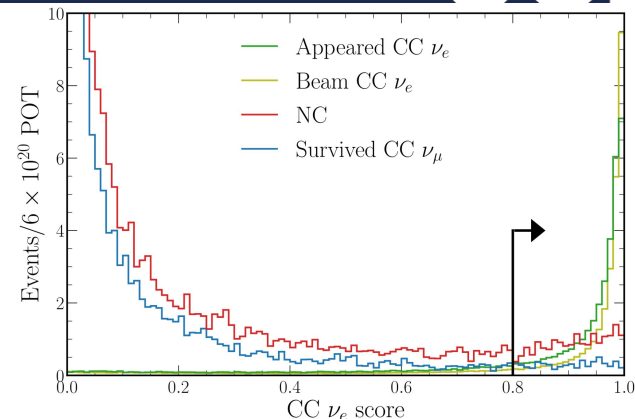
Selection	App CC ν_e	CC ν_μ	Beam CC ν_e	NC	Cosmic
Total events	44.17 ± 0.15	2045.9 ± 5.8	35.06 ± 0.08	354.7 ± 2.4	2100000 ± 4200
+ Preselection	41.21 ± 0.14	1889.5 ± 5.6	33.52 ± 0.08	243.2 ± 2.0	430000 ± 1900
+ Cosmic cut	41.10 ± 0.14	1874.4 ± 5.5	33.35 ± 0.08	241.6 ± 2.0	< 2
+ Escapes cut	40.68 ± 0.14	795.7 ± 3.5	32.86 ± 0.08	233.0 ± 2.0	< 2
Cuts Eff	$92.1 \pm 0.1\%$	$38.9 \pm 0.1\%$	$93.7 \pm 0.1\%$	$65.7 \pm 0.3\%$	$< 9.5 \times 10^{-7}$

Cosmics are not a problem, even without a veto

Results: CC ν_e selection

Selection	CC ν_e sig	CC ν_μ bkg	CC ν_e bkg	NC bkg	Purity sig	Purity CC ν_e
Total events	44.17 ± 0.15	2045.9 ± 5.8	35.06 ± 0.08	354.7 ± 2.4	$1.78 \pm 0.02\%$	$3.19 \pm 0.03\%$
+ Cuts	40.68 ± 0.14	795.7 ± 3.5	32.86 ± 0.08	233.0 ± 2.0	$3.69 \pm 0.02\%$	$6.67 \pm 0.03\%$
+ FOM- ν_e	31.27 ± 0.12	6.0 ± 0.3	26.69 ± 0.07	17.8 ± 0.6	$38.3 \pm 0.3\%$	$70.9 \pm 0.6\%$
Cuts Eff	$92.1 \pm 0.1\%$	$38.9 \pm 0.1\%$	$93.7 \pm 0.1\%$	$65.7 \pm 0.3\%$	-	-
FOM- ν_e Eff	$70.8 \pm 0.2\%$	$0.29 \pm 0.02\%$	$76.1 \pm 0.1\%$	$5.0 \pm 0.2\%$	-	-

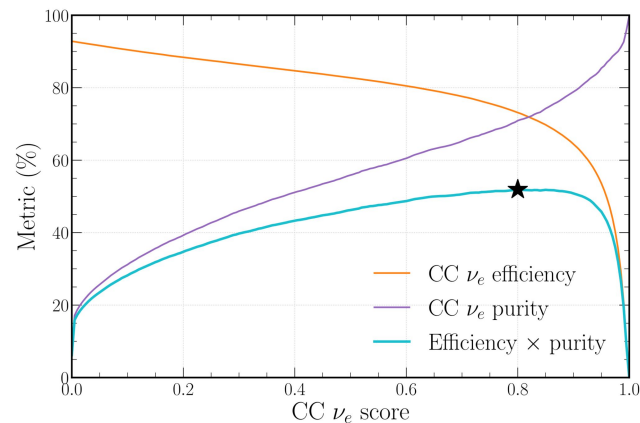
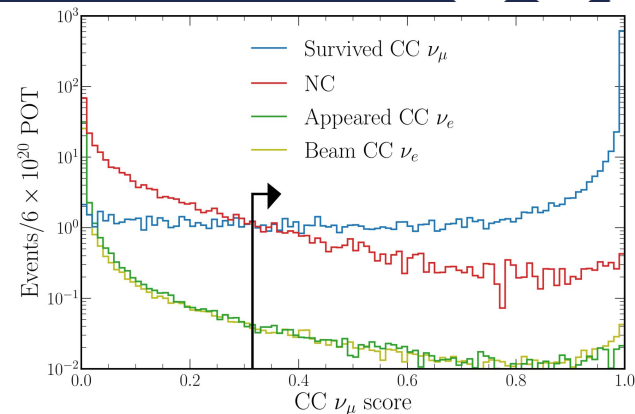
- Old methodology produced a maximum FOM (efficiency*purity) of 0.132 compared to the 0.519 now.
- The 71% signal efficiency compares well to 62% and 64% achieved by NOvA and T2k. But the purity is much lower at 38% compared to 78% and 80% reached by NOvA and T2k.



Results: CC ν_μ selection

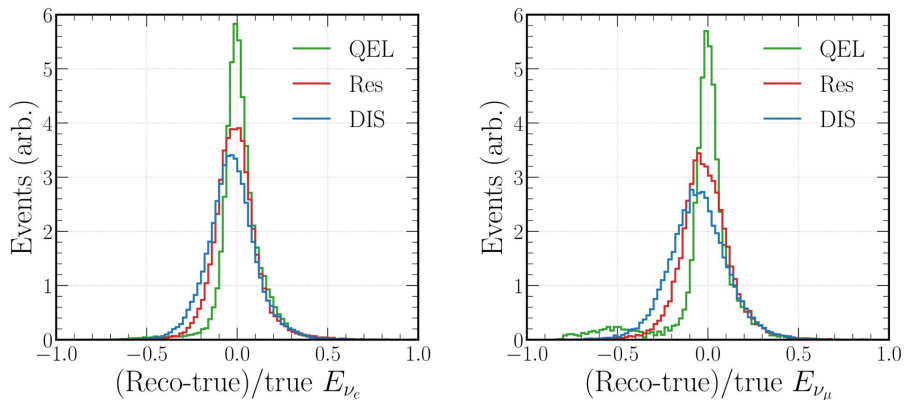
Selection	CC ν_μ sig	App CC ν_e bkg	Beam CC ν_e bkg	NC bkg	Purity sig
Total events	2045.9 ± 5.8	44.17 ± 0.15	35.06 ± 0.08	354.7 ± 2.4	$82.5 \pm 0.2\%$
+ Cuts	795.7 ± 3.5	40.68 ± 0.14	32.86 ± 0.08	233.0 ± 2.0	$72.2 \pm 0.2\%$
+ FOM- ν_μ	756.4 ± 3.4	1.293 ± 0.03	1.315 ± 0.02	29.0 ± 0.7	$96.0 \pm 0.1\%$
Cuts Eff	$38.9 \pm 0.1\%$	$92.1 \pm 0.1\%$	$93.7 \pm 0.1\%$	$65.7 \pm 0.3\%$	-
FOM- ν_μ Eff	$37.0 \pm 0.1\%$	$2.9 \pm 0.1\%$	$3.8 \pm 0.1\%$	$8.2 \pm 0.2\%$	-

- The signal efficiency of 37% compares well to the 31% and 36% achieved by NOvA and T2K
- This is also the case for the signal purity of 96% compared to the 98.6% and 94% purities of NOvA and T2K.

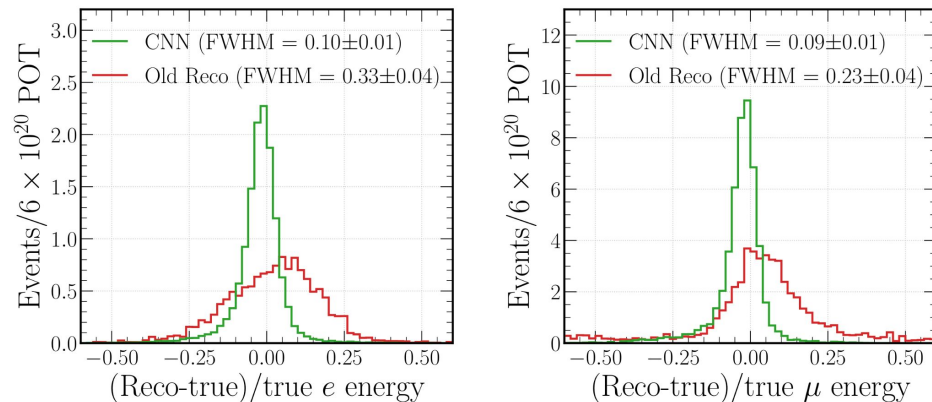


Results: Neutrino energy estimation

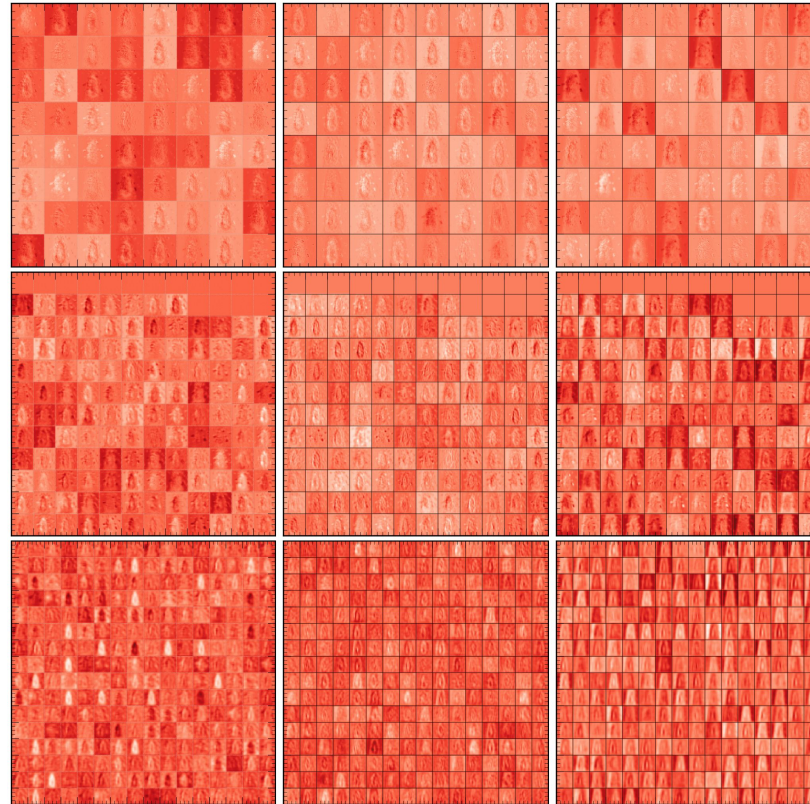
Neutrino energy resolution by component



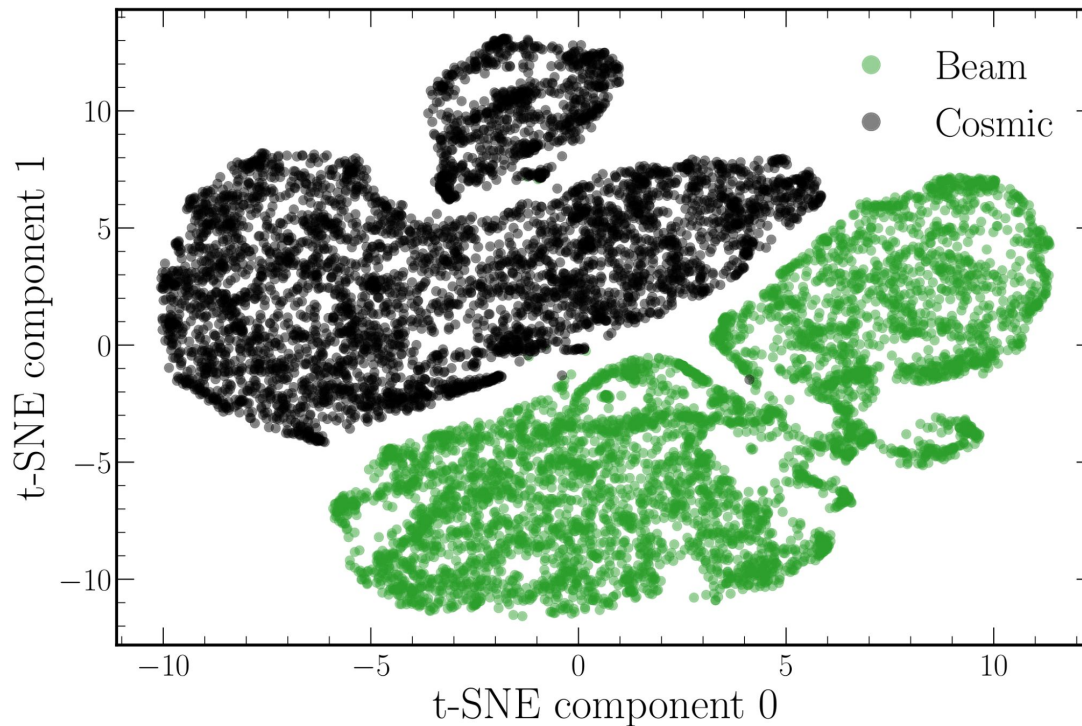
Charged lepton energy resolution comparison



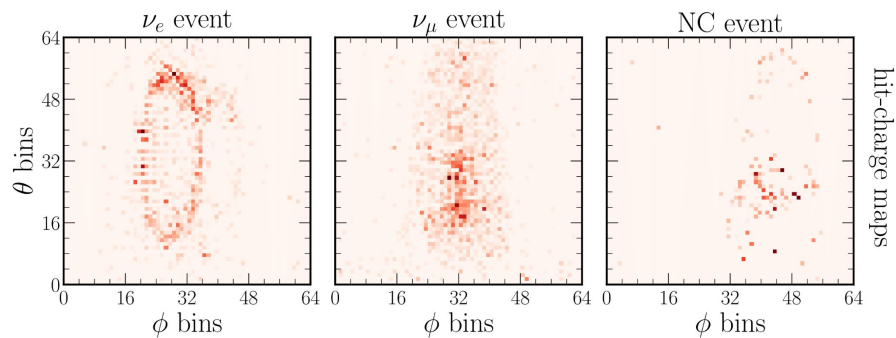
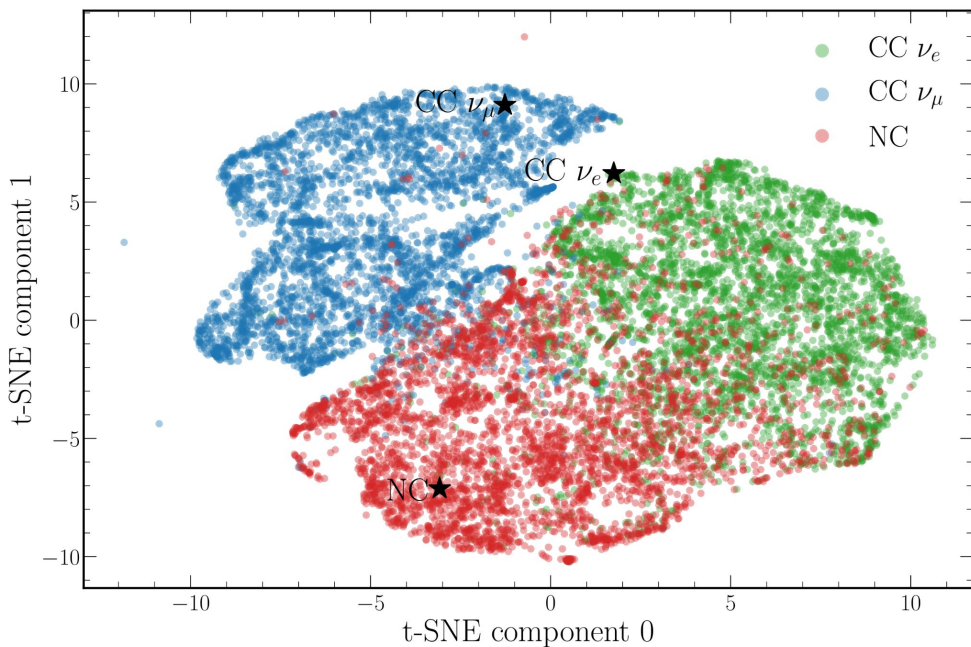
Explainability: The learnt kernels



Explainability: Cosmic rejection t-SNE



Explainability: Beam classification t-SNE



A large, circular steel structure under construction, possibly a bridge or a large-scale architectural element, is shown in a hazy, sunset or sunrise setting. The structure is composed of a complex network of steel beams and supports, forming a cylindrical shape. The sun is low on the horizon, creating a strong glow and casting long shadows. The structure is reflected in the water in the foreground. The overall scene is atmospheric and industrial.

Conclusions + Future Work

Conclusions

- Convolutional neural networks can be incredibly effective at solving neutrino event characterisation problems (classification and regression) in water Cherenkov detectors.
-

A large, circular steel structure under construction, possibly a stadium or arena, is shown at sunset. The structure is composed of a complex network of steel beams and trusses, forming a cylindrical shape. The sun is low on the horizon, creating a warm, golden glow and reflecting on the water in the foreground. The background shows a landscape with trees and hills under a clear sky.

Any Questions?

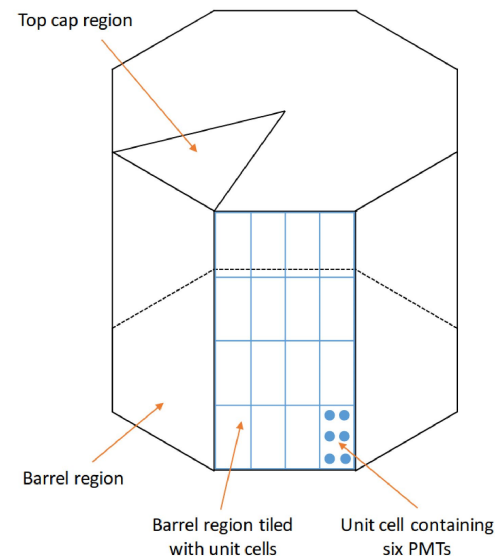
Future Work



- Blah Blah

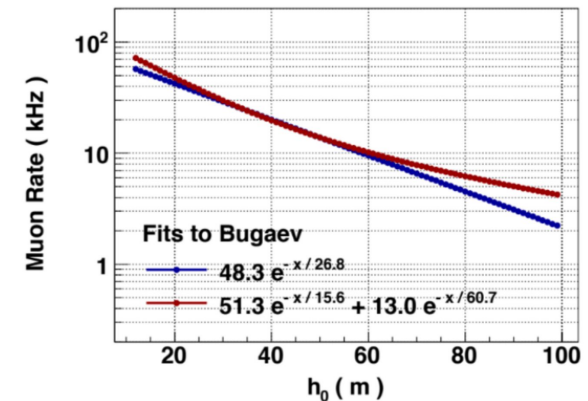
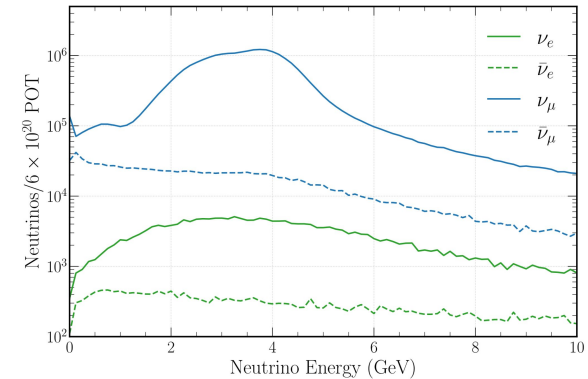
Detector simulation

- We use a modified version of **WCSSim** (Geant4) for simulating neutrino interactions within our detector.
- It builds an n-sided regular polygonal prism consisting of two endcaps and a barrel.
- Individual *unit cells* are used to tile the walls of the detector with PMTs according to the configuration.

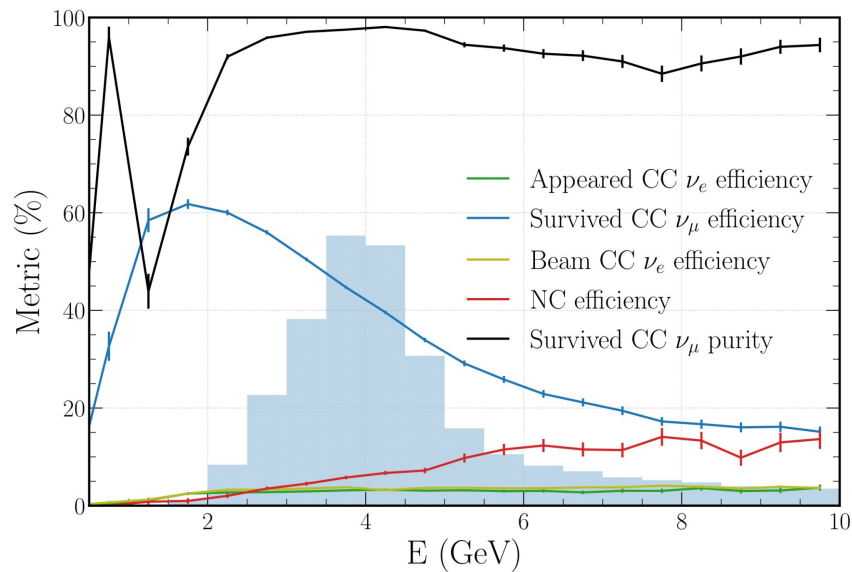
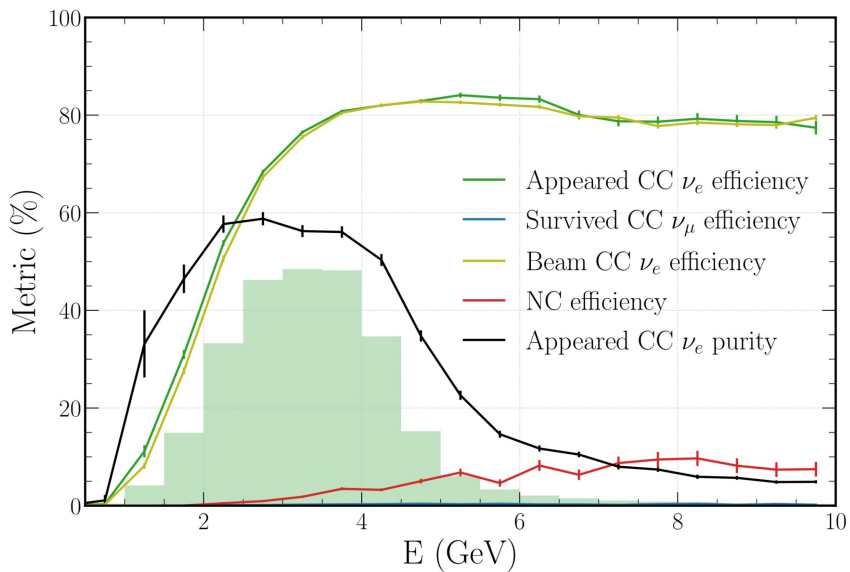


Event generation

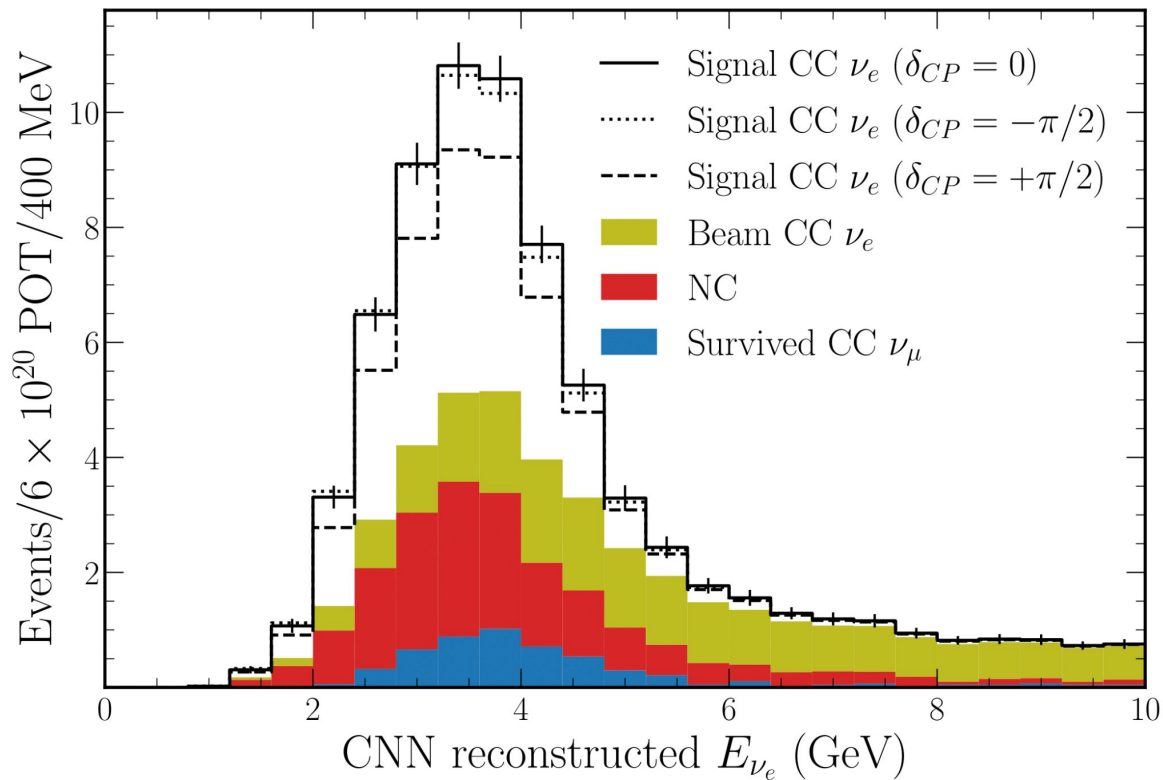
- For beam events, we use **GENIE**: The existing NuMI beam simulation is used to generate fluxes. Default cross-sections are used.
- For cosmic events, we use **CRY**: We assume an overburden of 50m of water and adjust the cosmic muon flux accordingly.
~ 11.8 KHz cosmic muon rate ~ 2.1 million/year



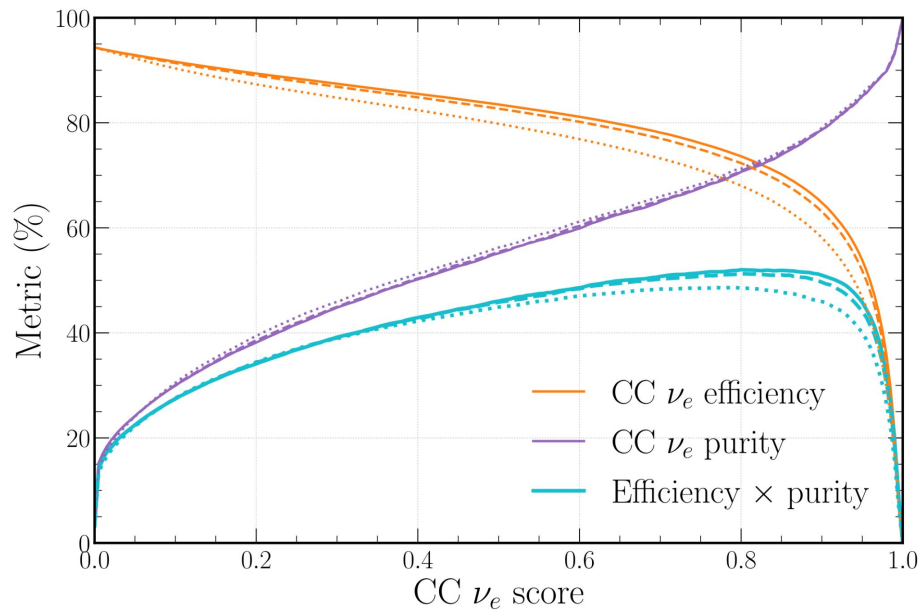
Eff/Pur vs Energy



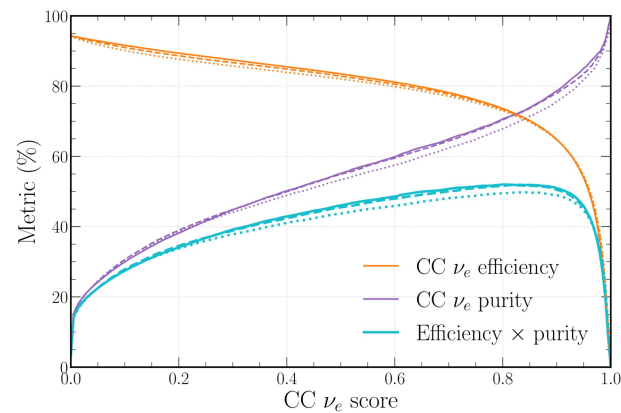
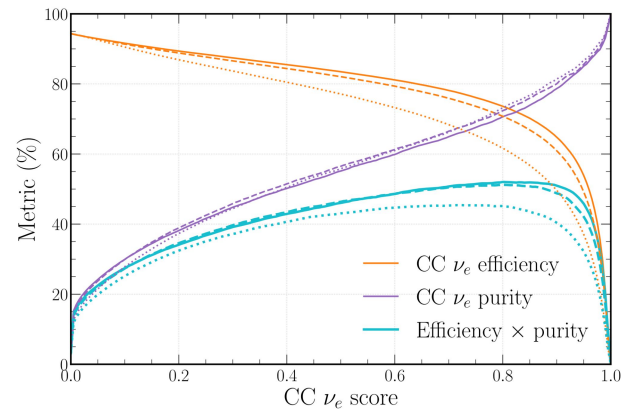
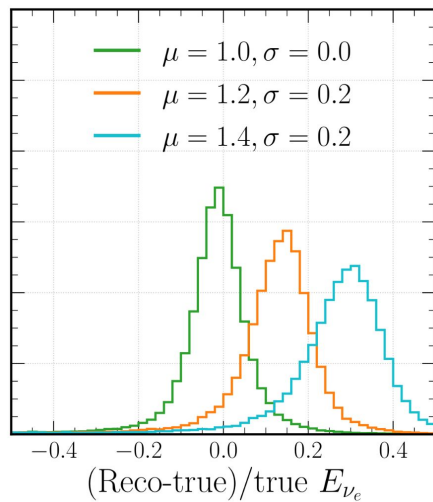
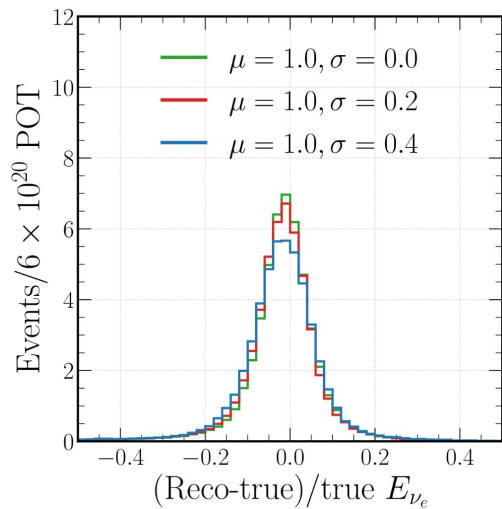
Different values of delta-cp?



Robustness: Time



Robustness: Charge



Robustness: Noise

