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PHYSTAT

Statistics meets Machine Learning

IMPERIAL

TUM

Identifying Tau Neutrinos in IceCube

PhyStat London 2024

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*** special thanks to:**

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Matthias Thiesmayer

Philipp Soldin

Philipp Fürst

Oliver Jannik

Part I: Physics Analysis

- Identifying Tau Neutrinos -

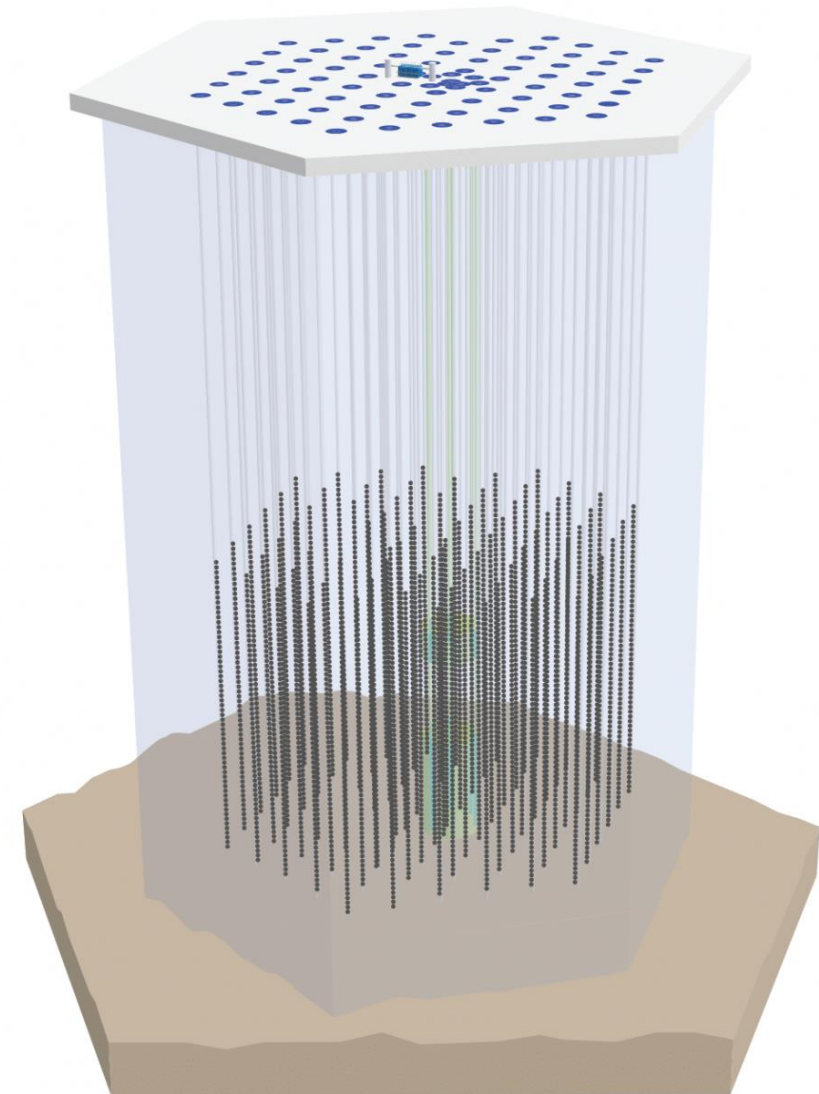
Part II: Interpretability

- Dissecting the Neural Net -



IceCube Neutrino Observatory

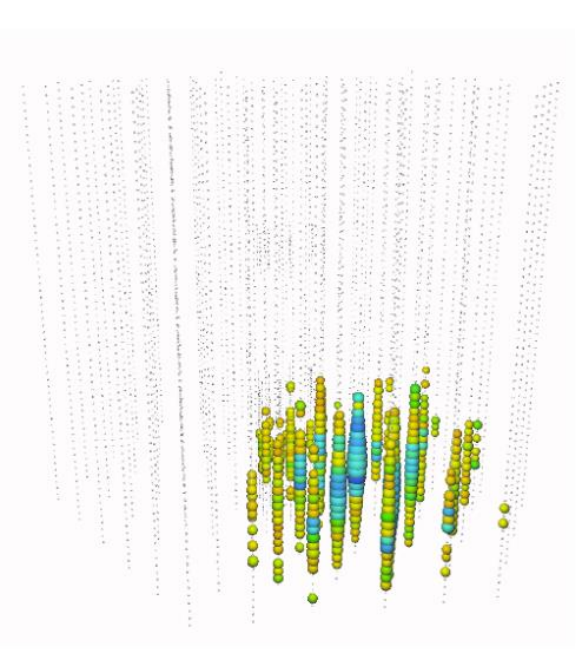
- Largest Neutrino Detector
 - Built into the South Pole ice
 - Instrumenting a volume of 1 km³
 - 5160 sensors (DOMs) arranged in 86 strings
- Capable of detecting neutrino interactions from GeV to PeV energies
- Upgrade of detector underway with 7 new strings
 - advanced sensors
 - new calibration hardware
 - to be deployed 2025/2026



Event Types in IceCube

“Casacdes”

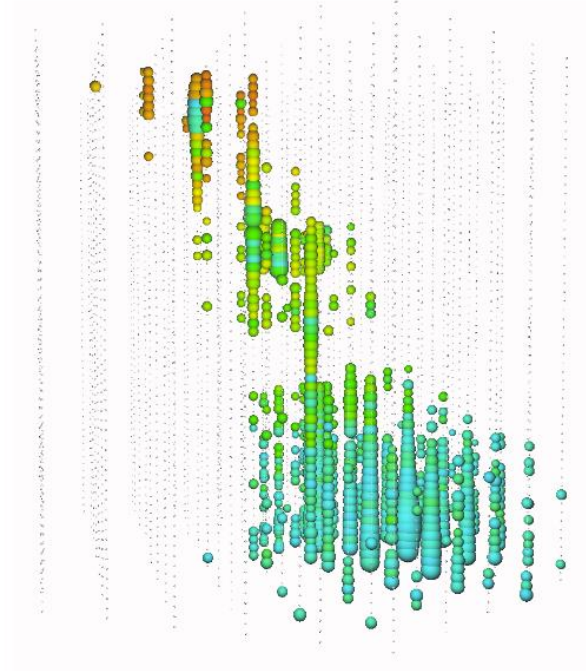
induced by ν_e or NC interactions



Common events

“Tracks”

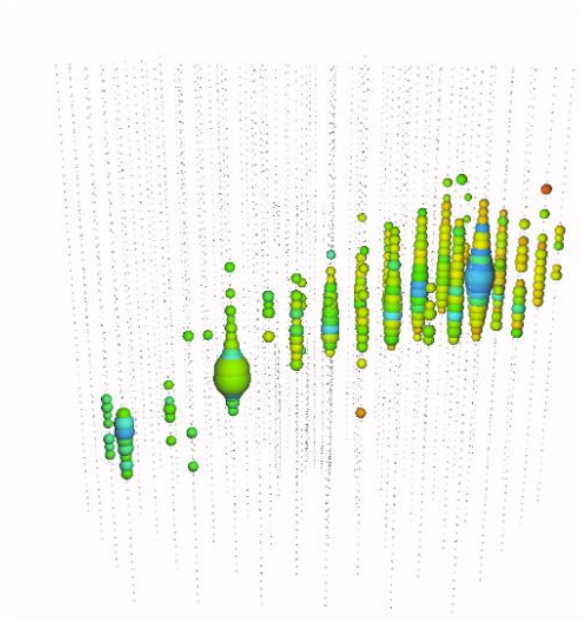
induced by Atm. μ (or ν_μ)



Ubiquitous events

“Starting Tracks”

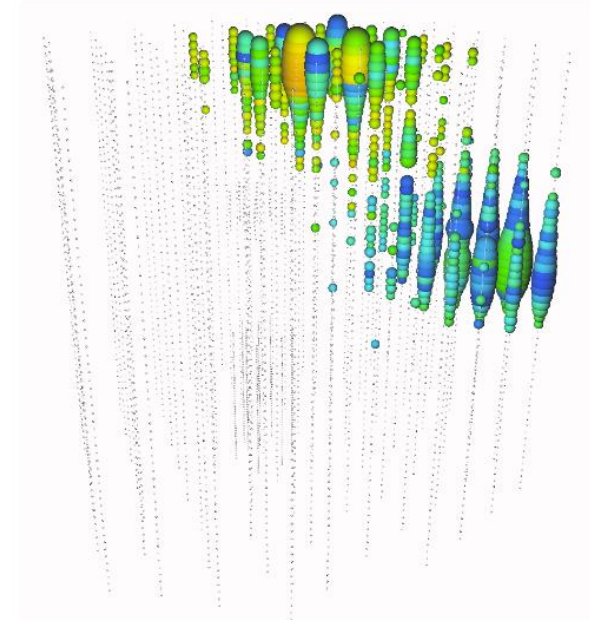
induced by ν_μ



Common events

“Double Cascades”

induced by ν_τ



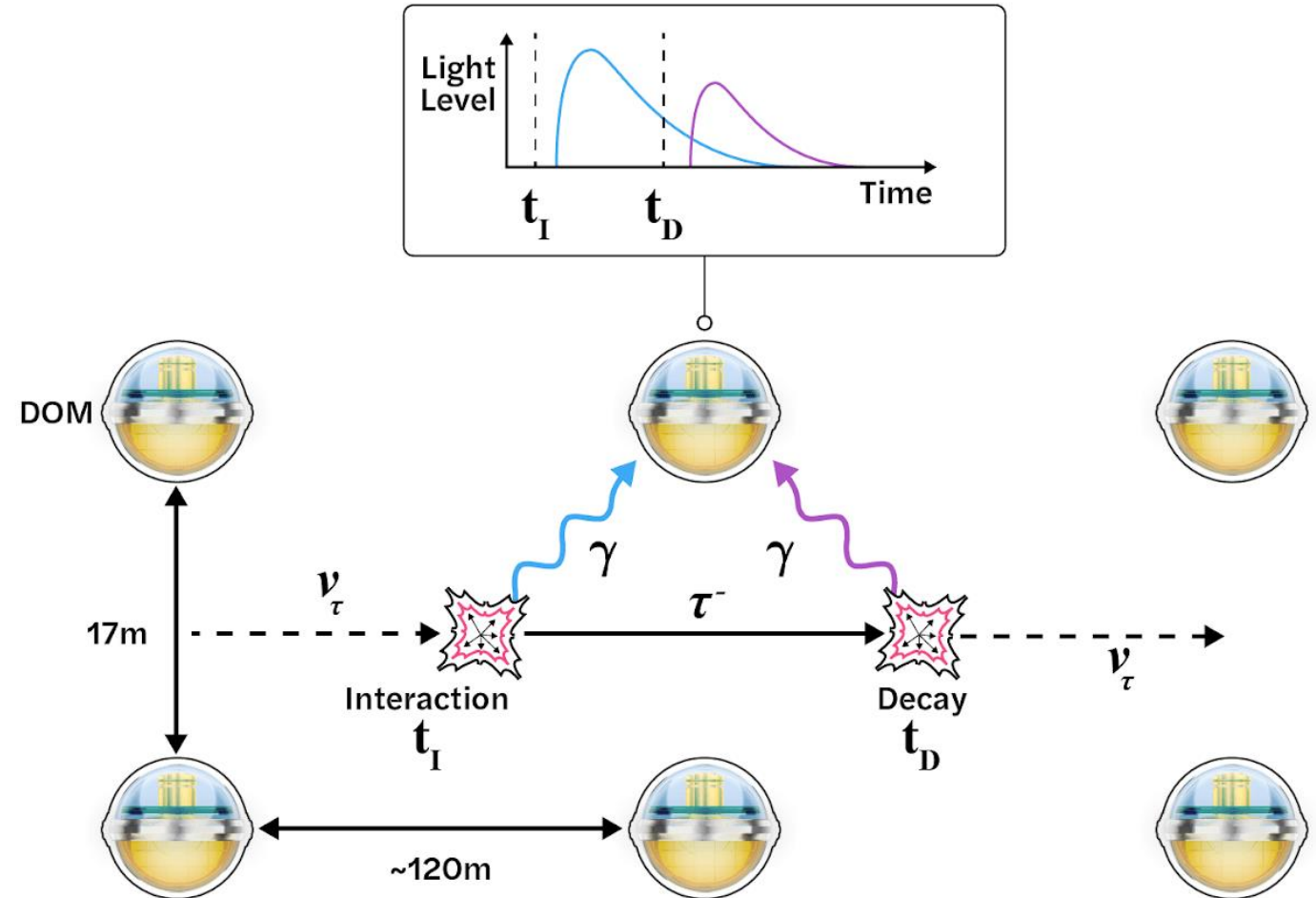
Very rare events

Tau Neutrinos

- A ν_τ **CC** interaction creates a τ lepton
 - Lifetime of 2.9×10^{-13} s
- If sufficiently energetic, it will travel several meters before decay (~ 50 m / PeV)

- Two separate vertices:
- First: Initial ν_τ **interaction**
 - Later: τ **decay**

→ **Double pulse signature**

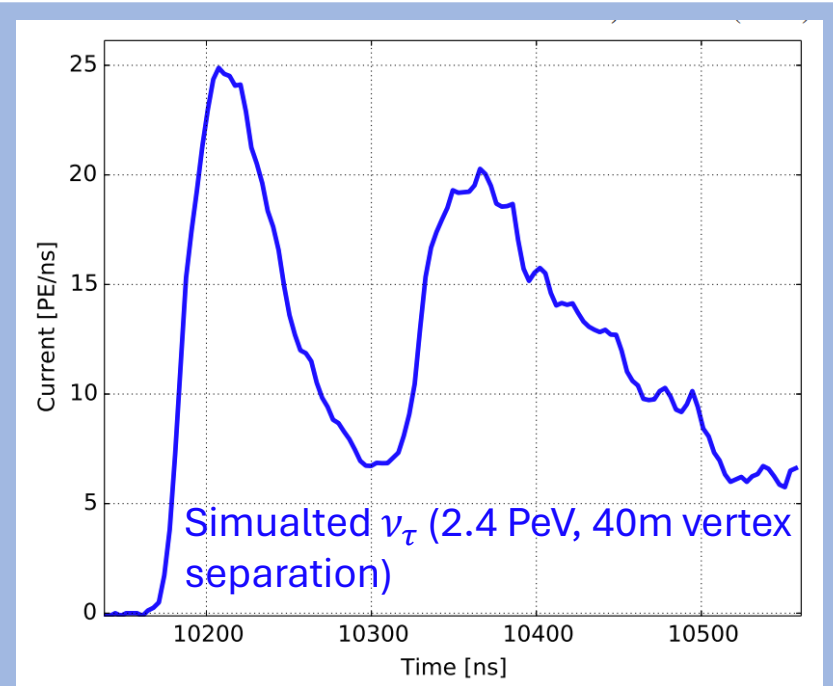


Previous Search Strategies

Double Peak in waveform

- <1 expected in 3 years of data
- 0 events observed

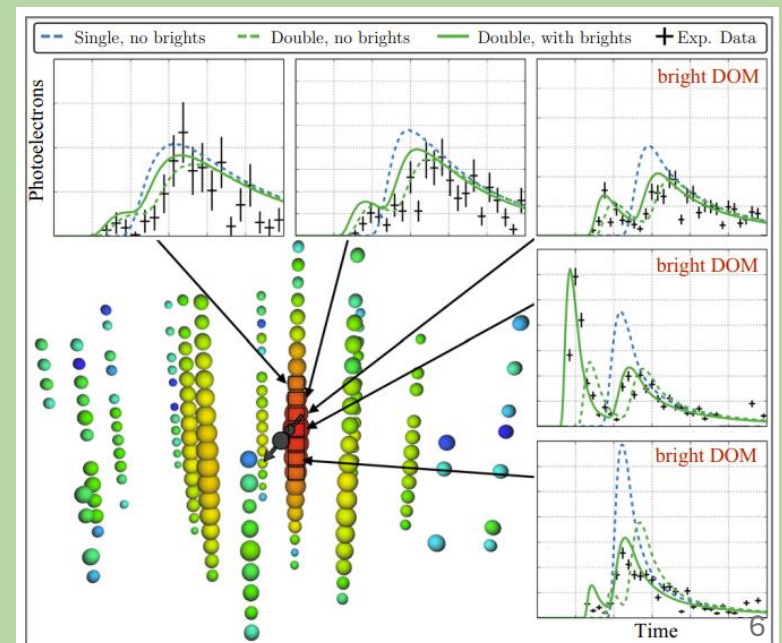
Phys. Rev. D 93, 022001



Double cascade reconstruction

- Using 60 high-energy-starting events (HESE)
- 1.5 signal and 0.8 background expected in 7.5 years of data
- 2 events observed → 2.8 sigma p-value

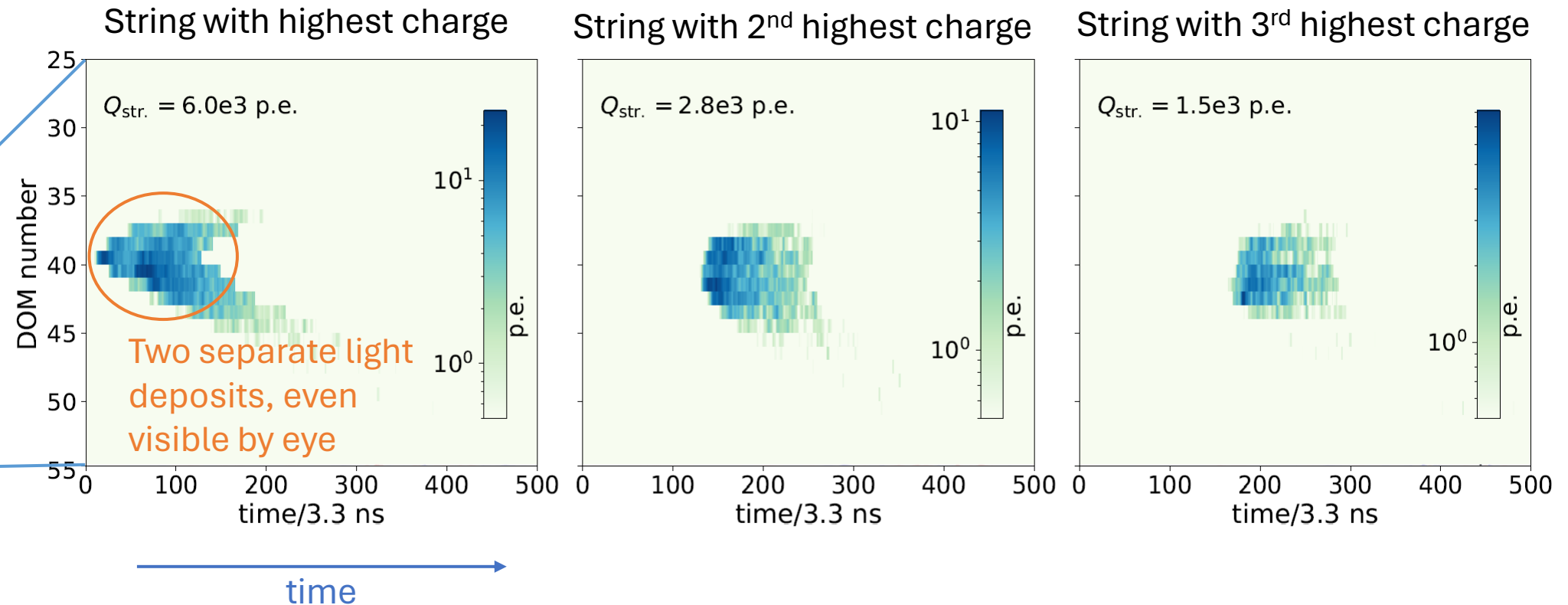
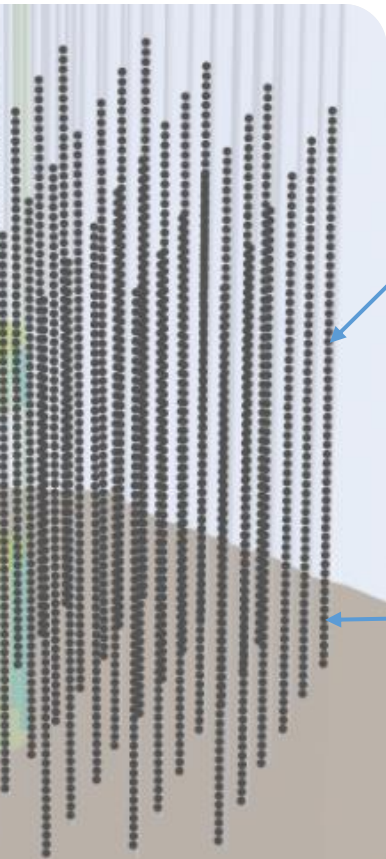
Eur. Phys. J. C 82, 1031 (2022)



New ML method

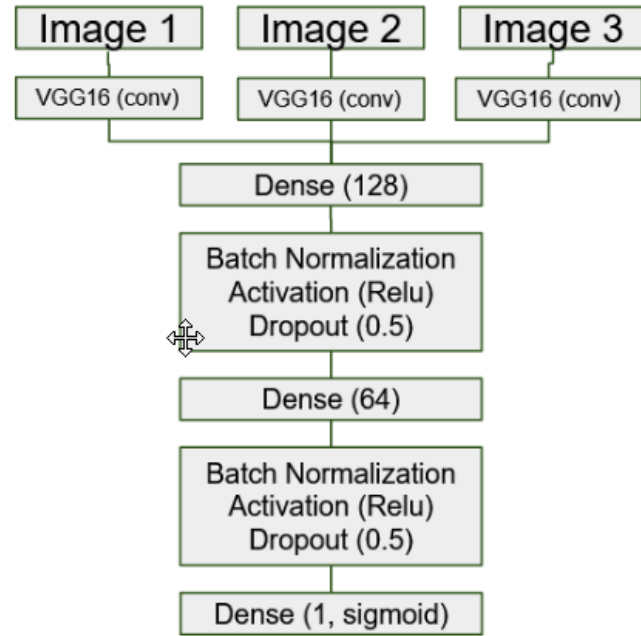
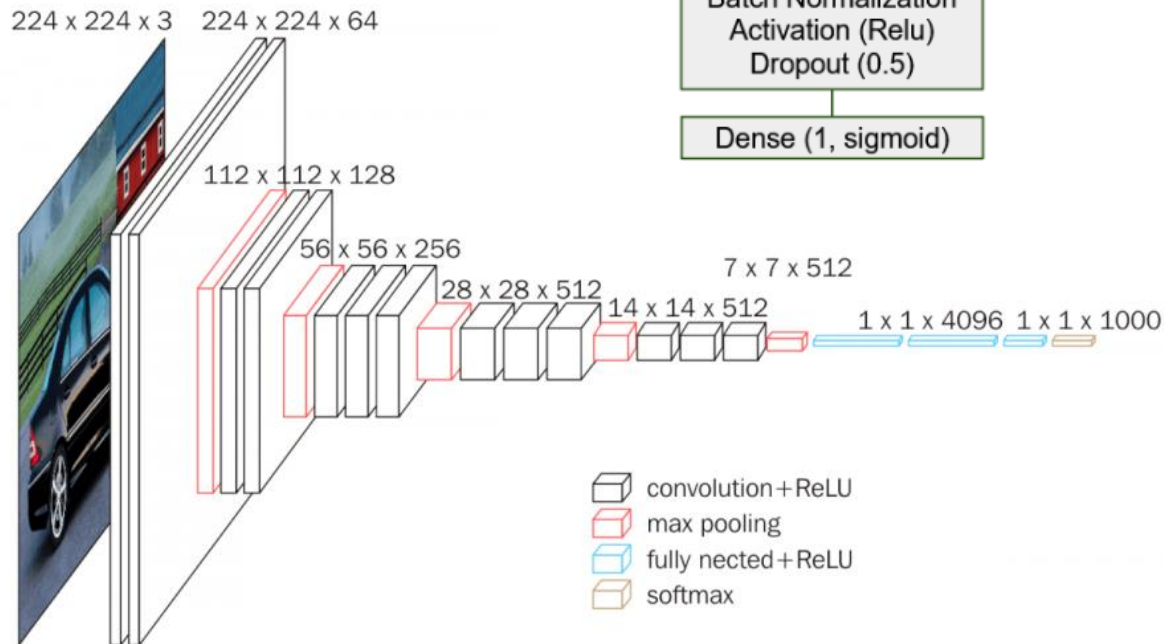
Use waveforms of the three detector strings with the highest observed charge

IceCube Array

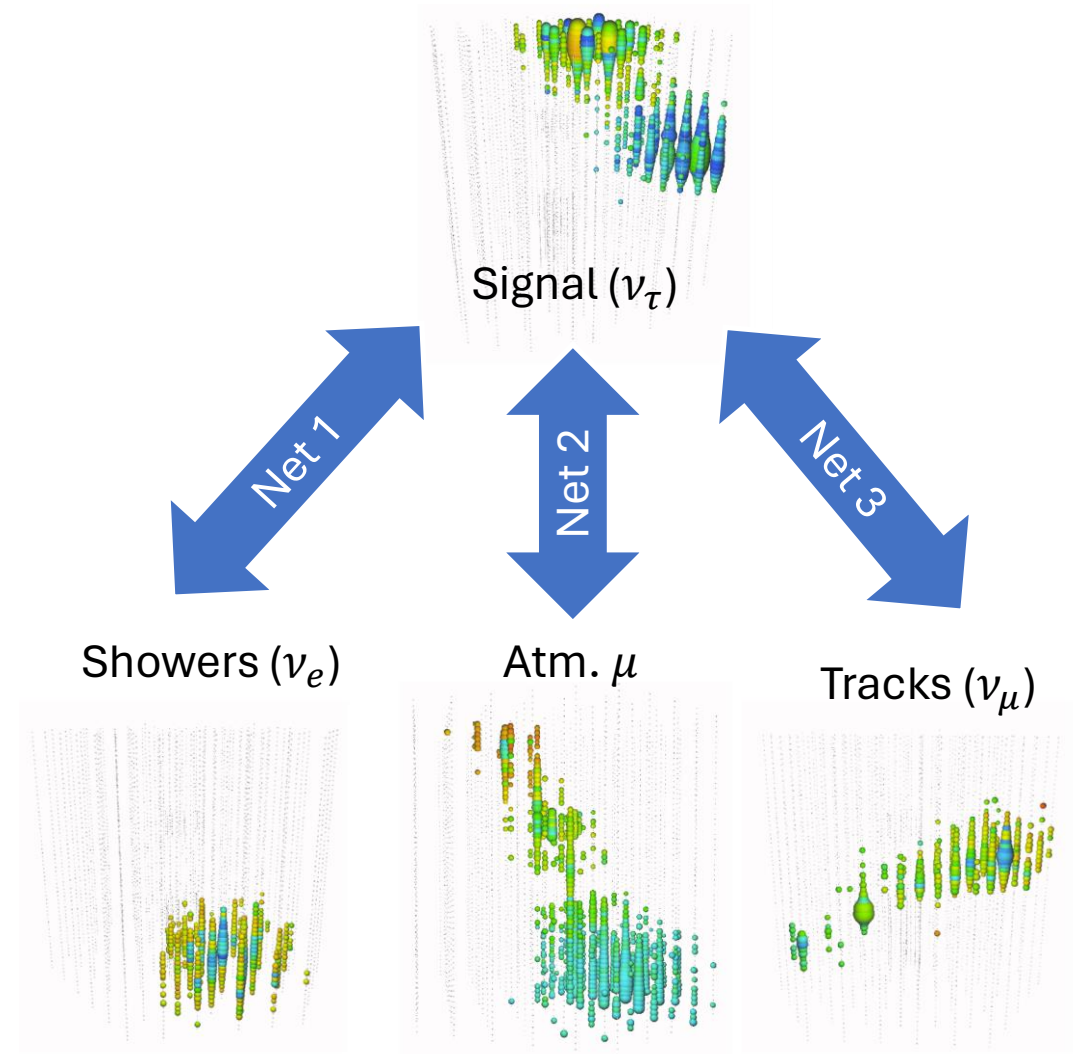


CNN

Based on
VGG16

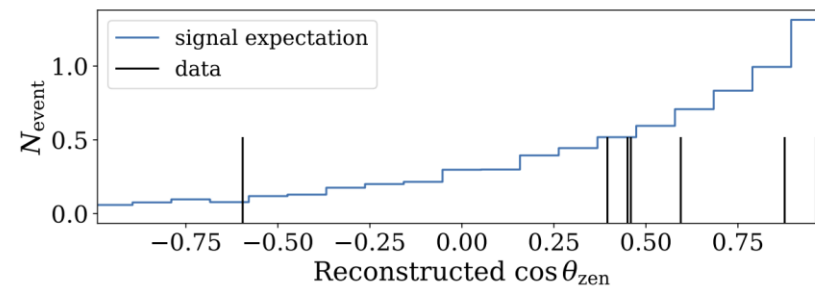
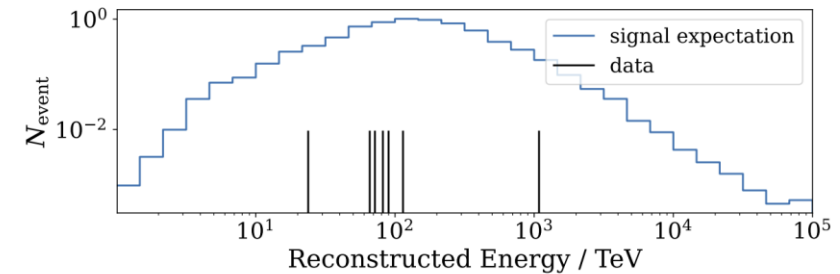
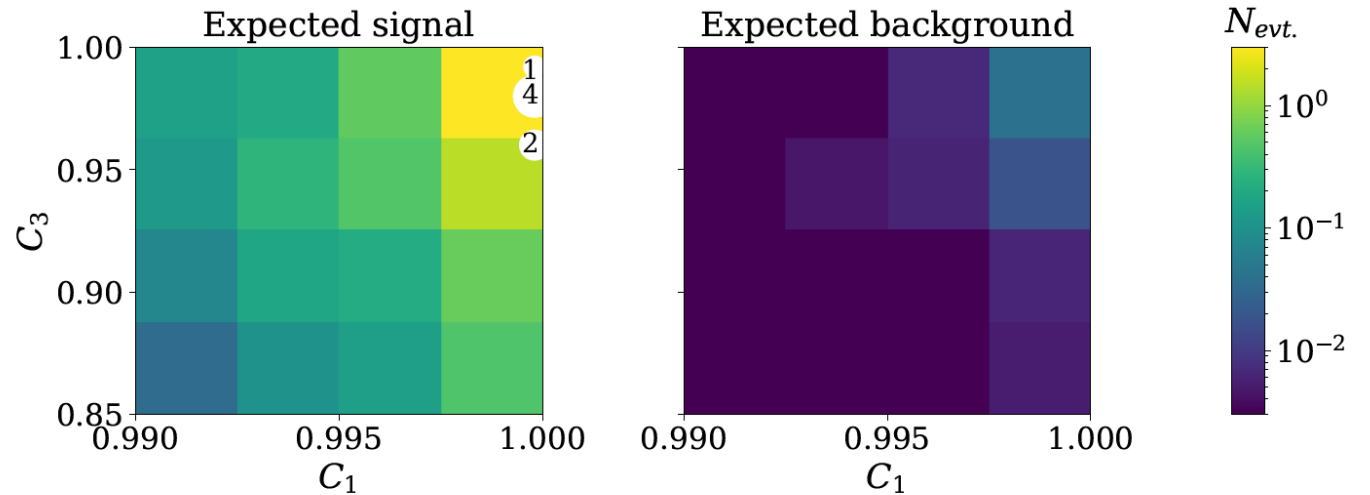


Train thrice with three
different classification tasks:



Results

- While in 9.7 years of data we expected:
 - 6.4 tau neutrinos
 - 0.5 background events
- We find 7 events in the signal region
 - One of these seven was also identified in a previous tau analysis
 - 5 sigma p-value for this being a background fluctuation



Part I: Physics Analysis

- Identifying Tau Neutrinos -

Part II: Interpretability

- Dissecting the Neural Net -



Robustness Tests I

Stress-test the neural networks, to see if they could easily be fooled:

- Uncorrelated variation of the light levels at the DOM level, to mimic mis-modeled uncertainties in DOM efficiencies
- Variation of the light levels for DOMs correlated in bands of depth to mimic unmodeled ice optical property uncertainties
- varying the light levels correlated by entire strings to mimic ice birefringence uncertainty

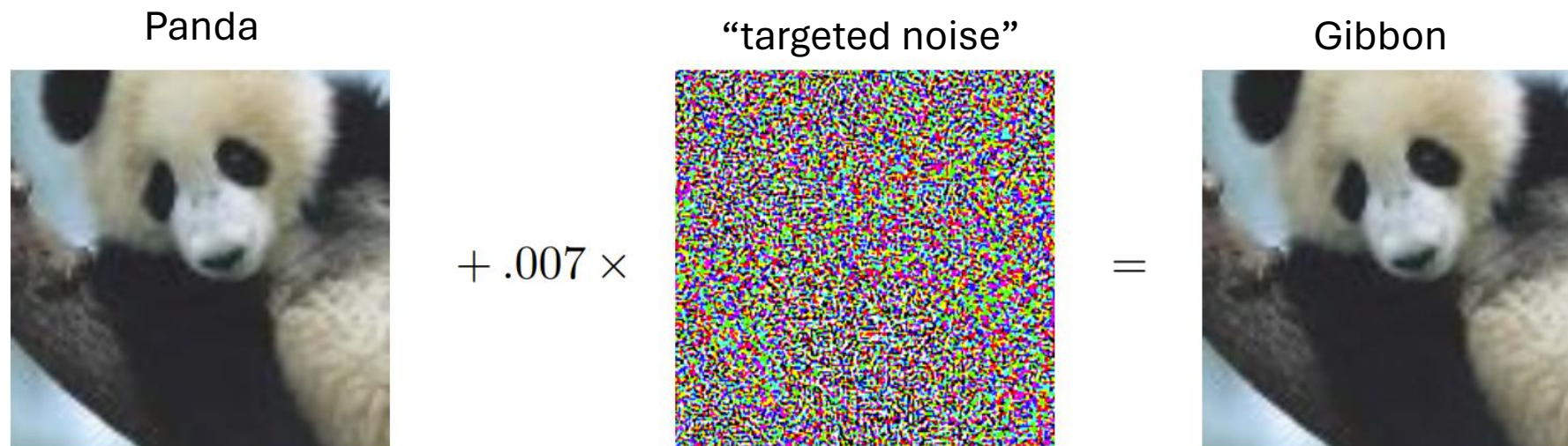
→ In all cases, the migration probabilities were small and consistent with the expectations from background simulations



Robustness Tests II: Adversarial attacks

Targeted attacks on neural nets aiming to migrate events between classes

→ Adding perturbations to the image in the direction of the gradient



arXiv: 1412.6572

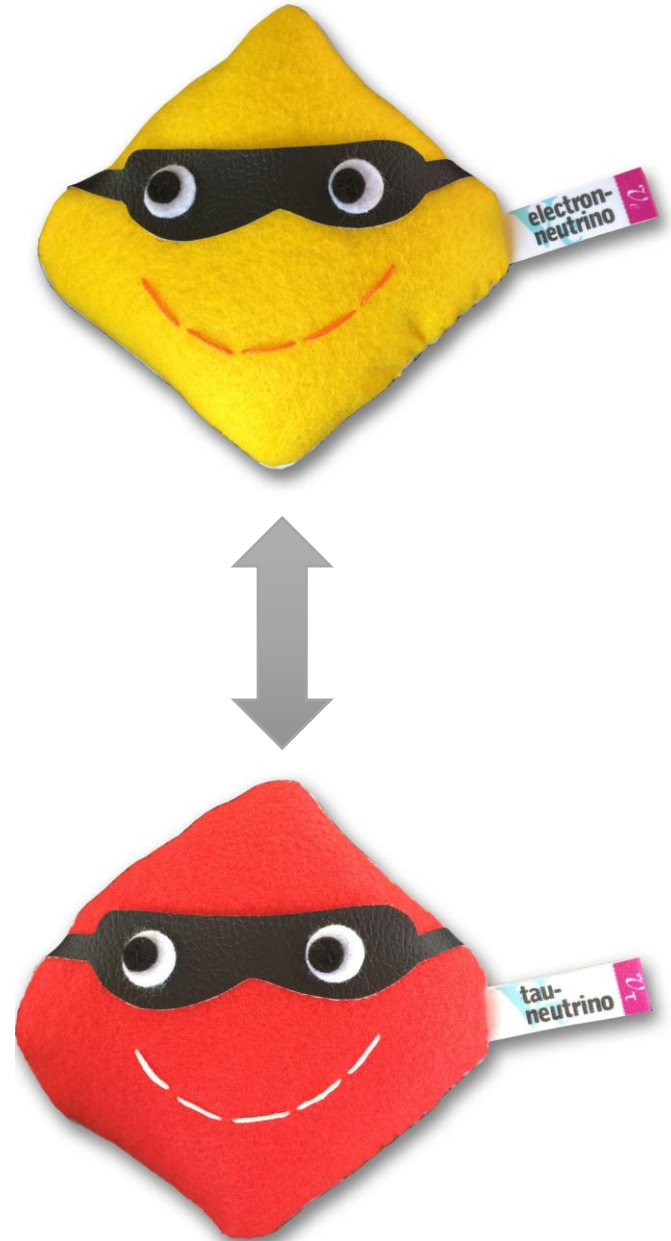
Adversarial Attacks on Tau Analysis

Turn Signal into Background

- For 6 out of the 7 signal events, pixel values would need to be changed so drastically, that it is well outside of what is considered reasonable ($\gg 10\%$)
- One event can be turned into background by introducing conceivable changes ($\sim 3\%$)
 - This is consistent with the expected 0.5 background events

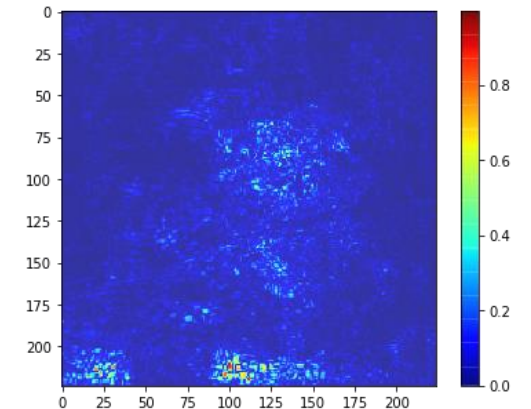
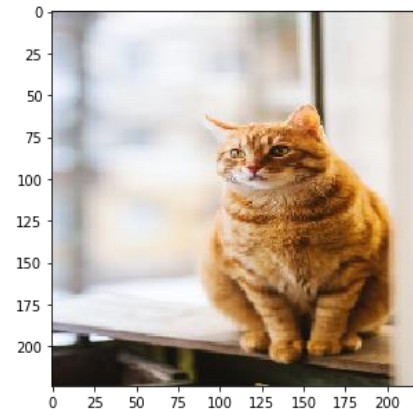
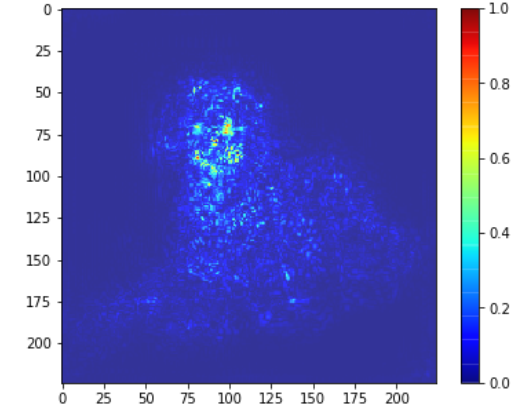
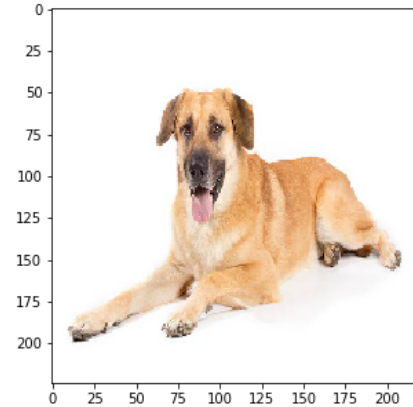
Turn Background into Signal

- Only success for one out of 634 attacked background (ν_e) events
- Needed to allow for large deviations (up to 10%)

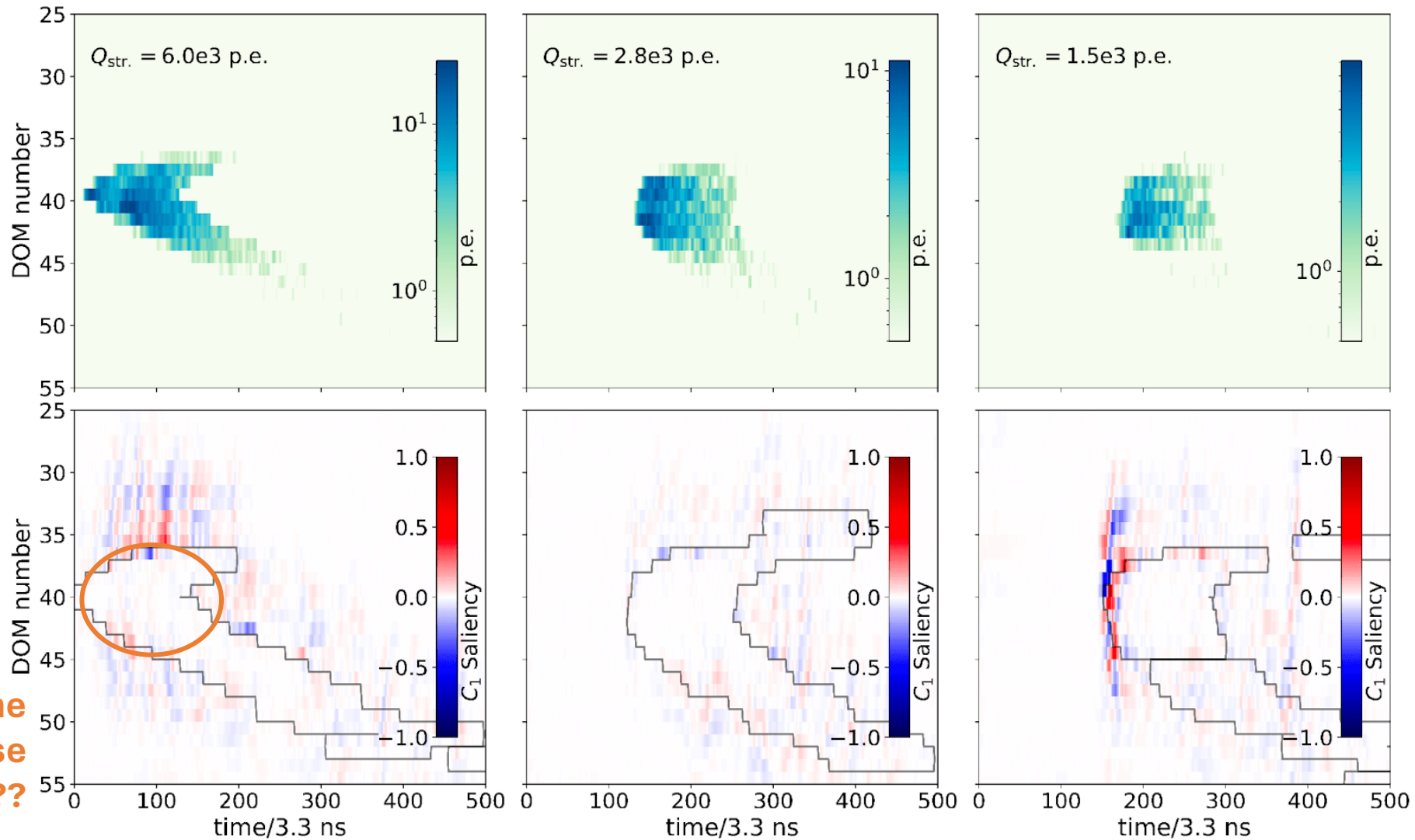


Saliency Maps

- This shows the importance of the input data on the network output
- Here it is $\partial \text{output} / \partial \text{pixel-value}$
(Normalized to 1)
- Large value \rightarrow change in pixel value will have large impact on network output

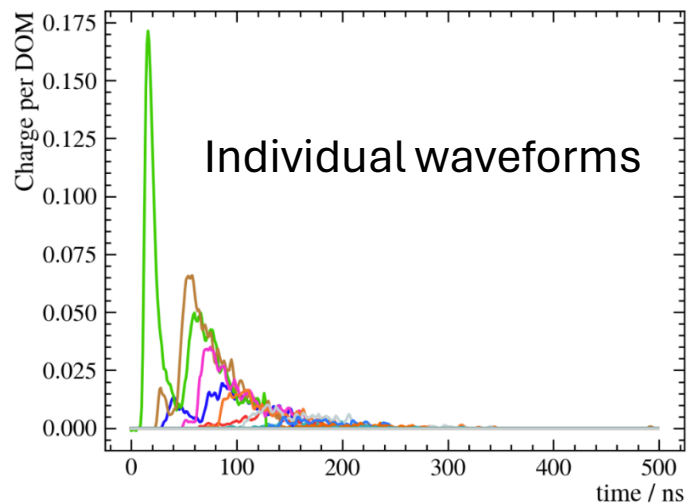
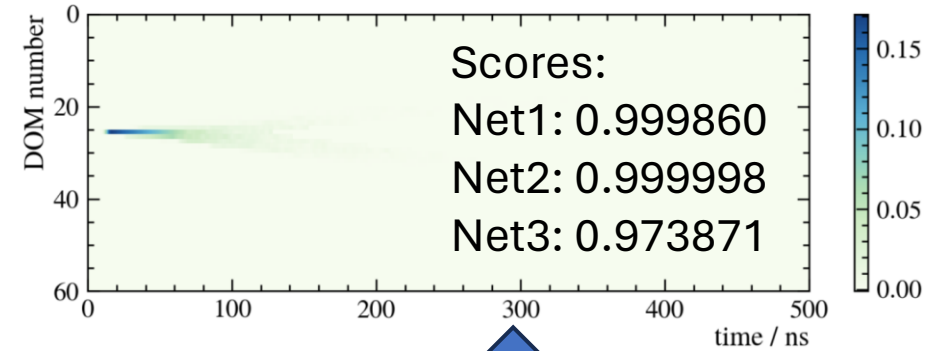
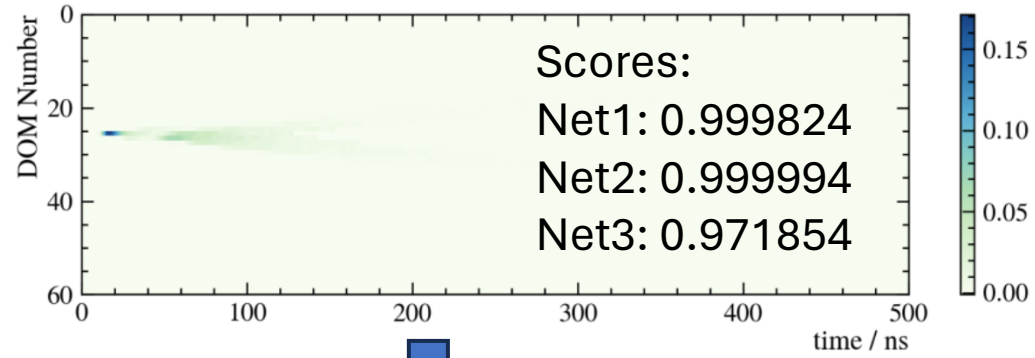


Tau Neutrino Saliency Maps

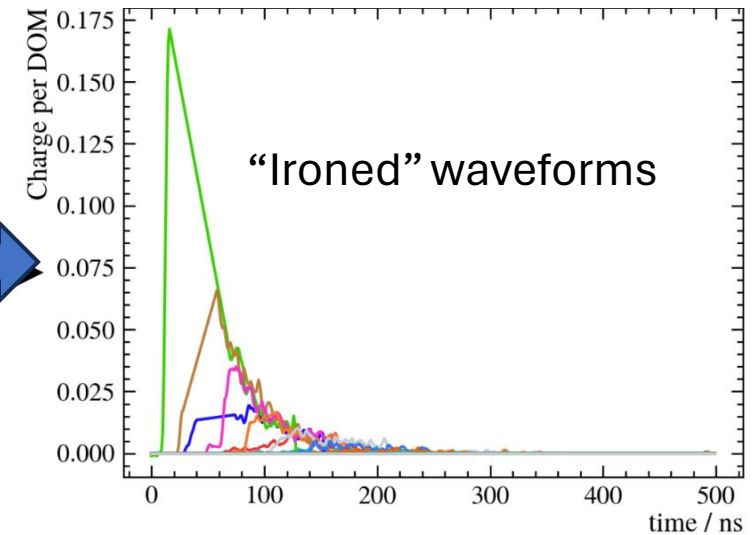


What about the double pulse waveforms??

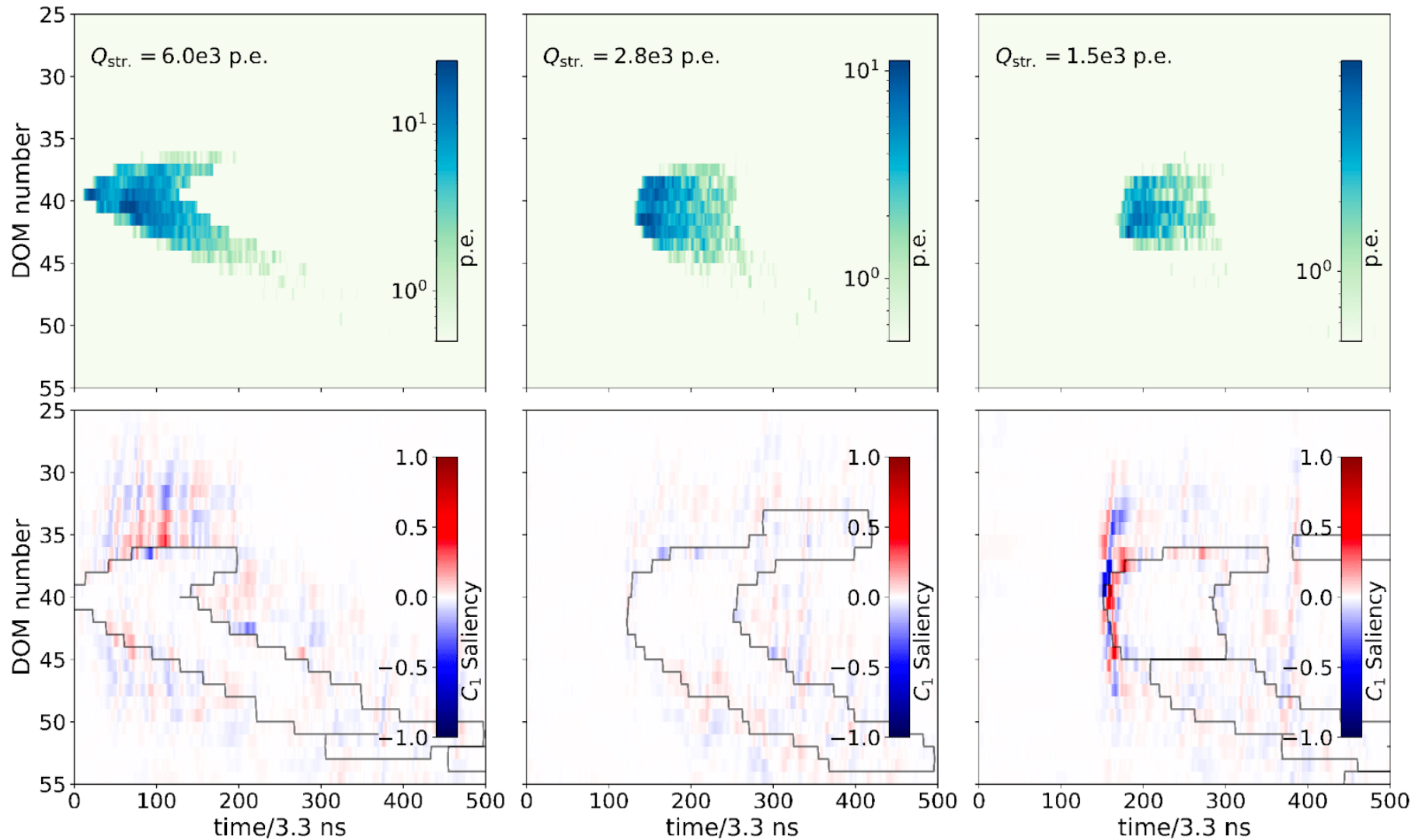
Ironing them out...



Linear interpolation
between separate
peaks



Tau Neutrino Saliency Maps



Summary / Conclusions

ML methods allowed us to **identify astrophysical tau neutrinos in IceCube for the first time!**

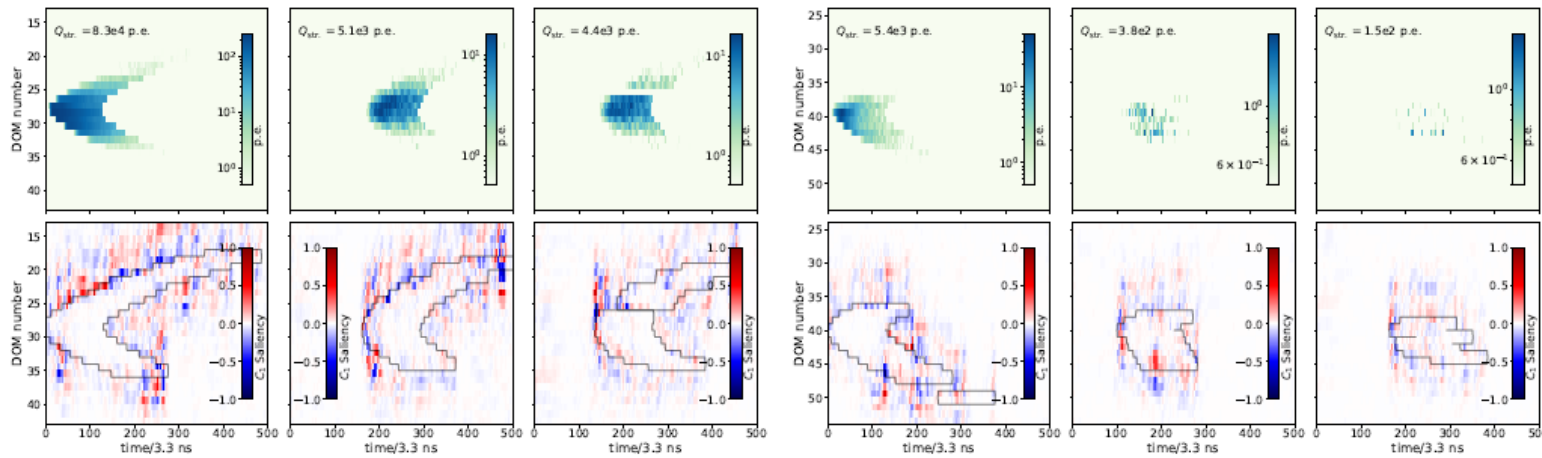
- Applying CNNs to 2d arrays of waveforms from sensors on strings
- 7 signal events observed, while expecting 6.4 tau neutrinos over 0.5 background

Large efforts were undertaken to test and understand the neural nets:

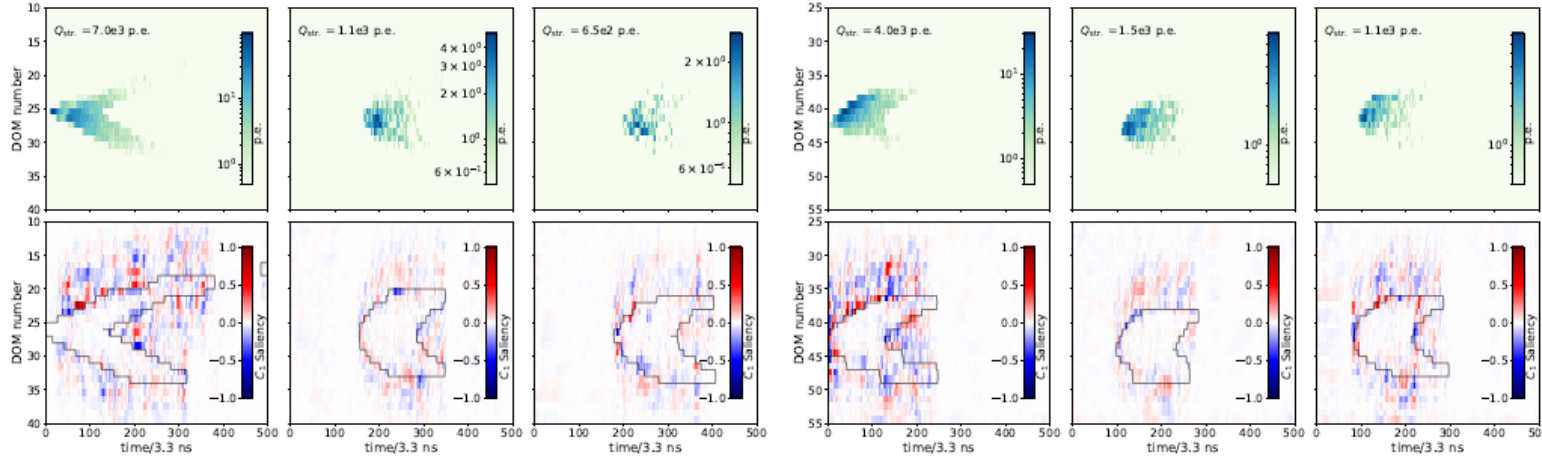
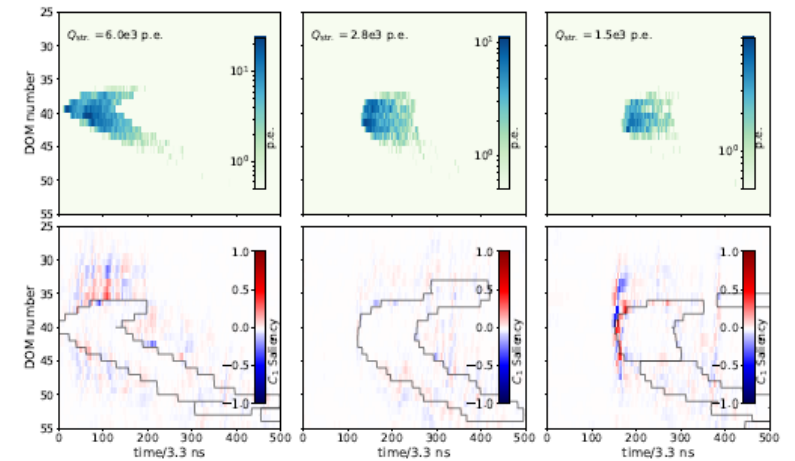
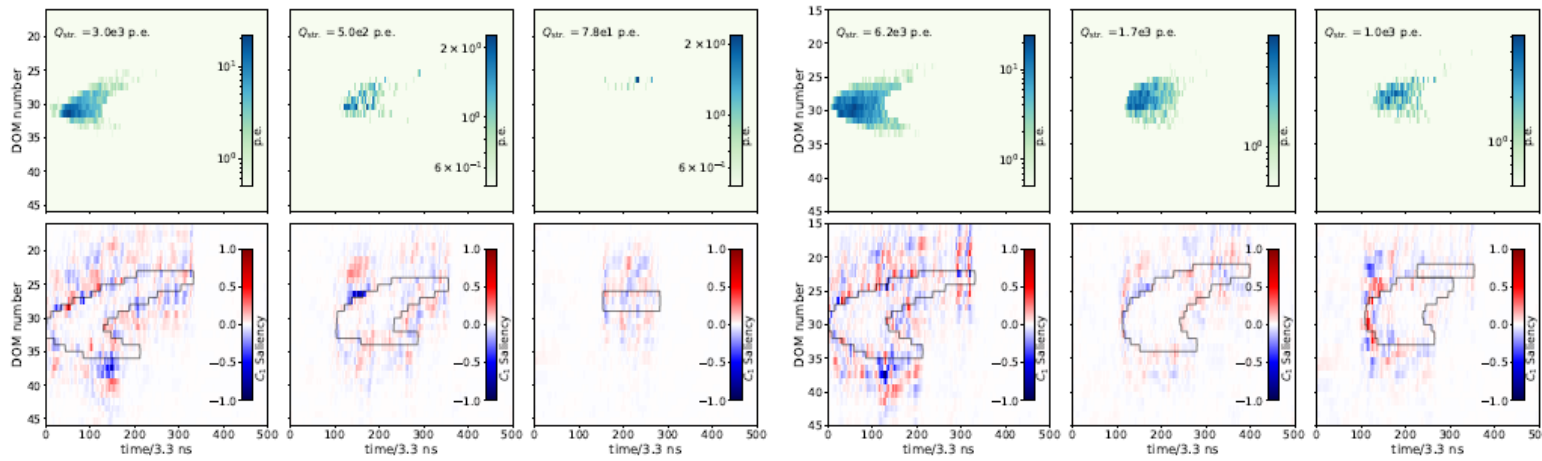
- **Robustness tests** varying input features according to uncertainties
- Targeted **adversarial attacks** to migrate events between classes
 - NNs proved robust against such attacks
- **Saliency maps** to visualize what features the network uses to form decisions
 - Revealed that double pulse waveform signature is not that useful for signal vs. background classification
 - Rather, the overall shape of the light registered in strings and the timing between strings is important, which gives us **new and better intuition** on how to improve future searches!

Additional Material





All 7 signal candidates
(observed events)

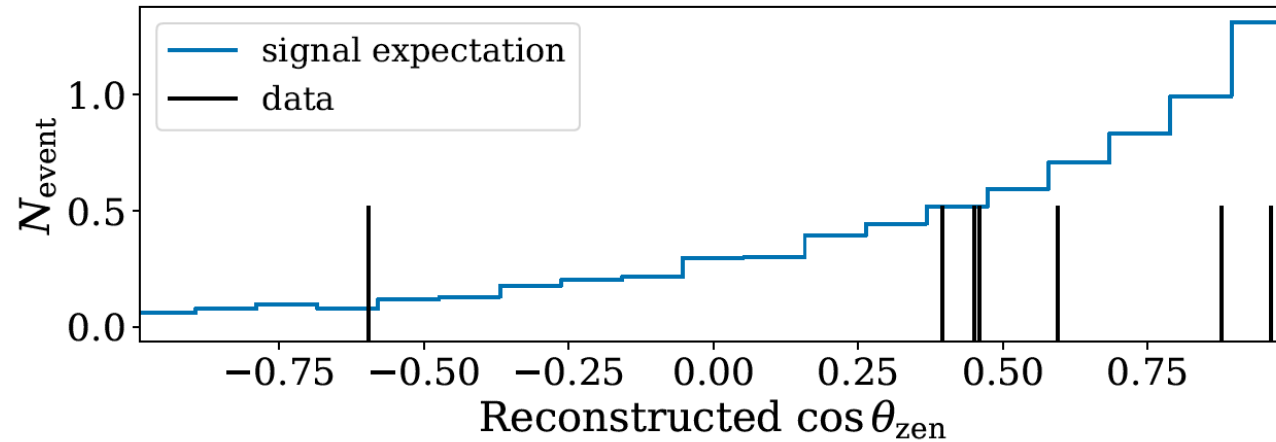
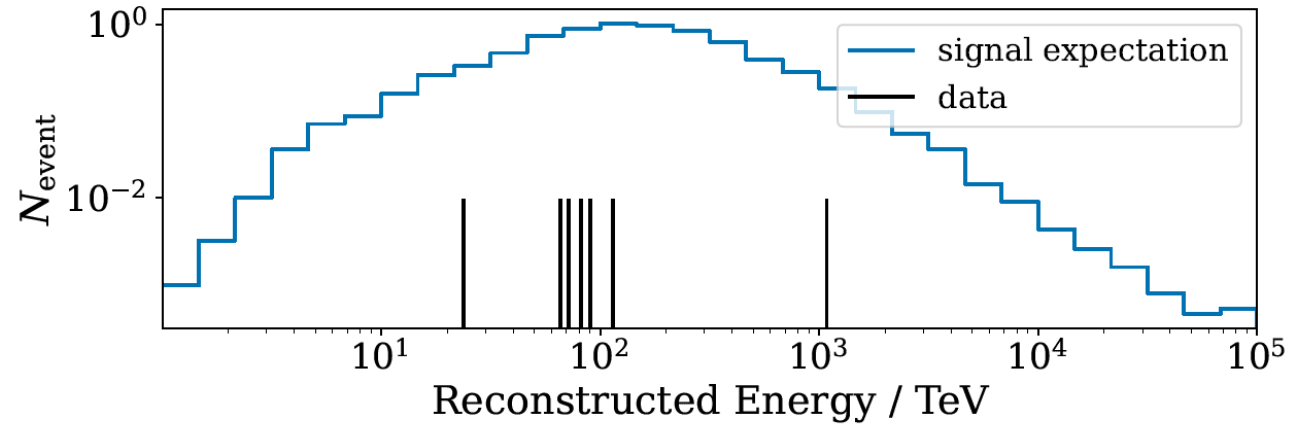


Number of Events

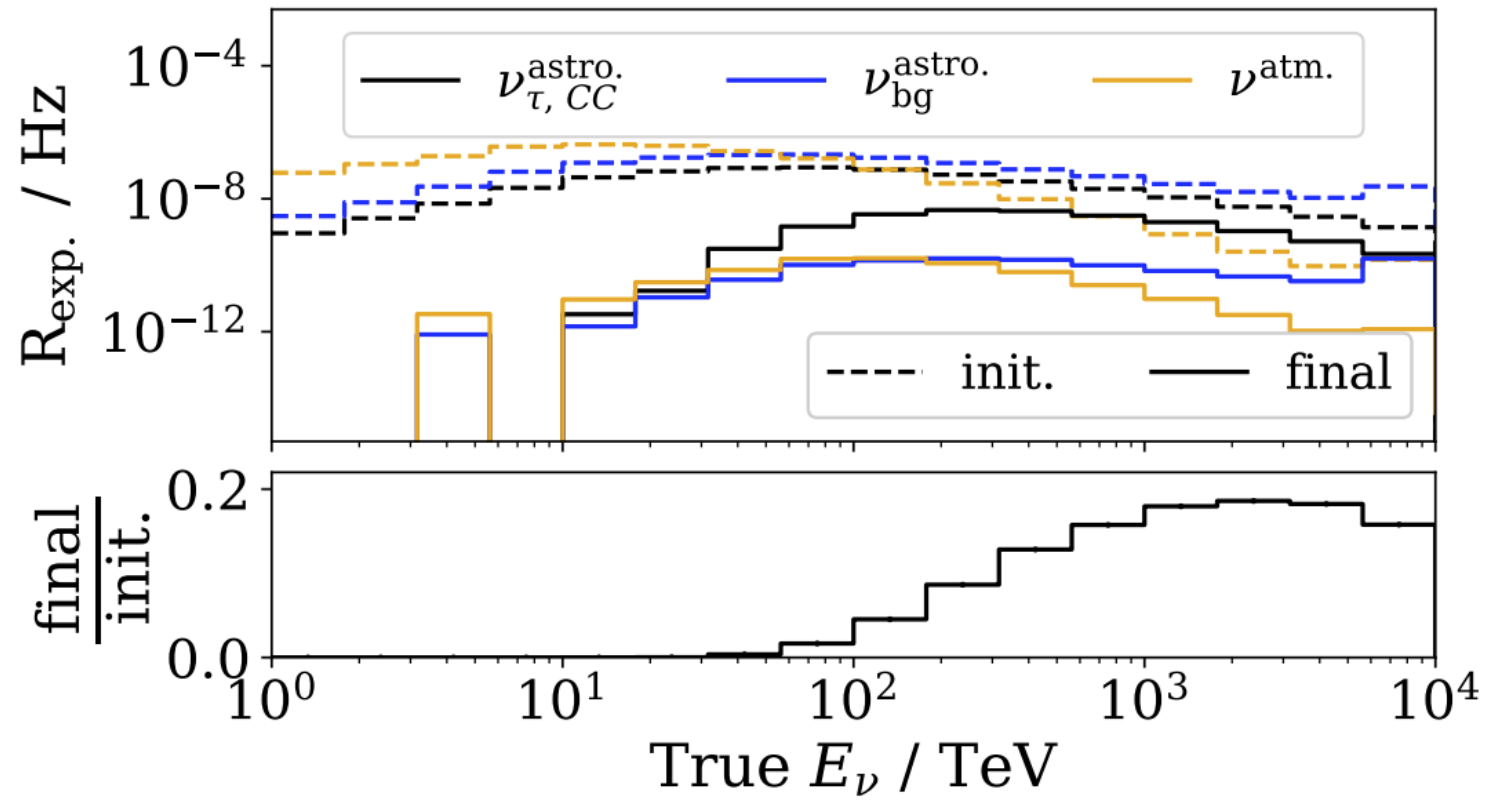
	$\nu_{\tau,CC}^{\text{astro}}$ [59]	$\nu_{\text{other}}^{\text{astro}}$ [59]	$\nu_{\text{conv.}}^{\text{atm.}}$ [60–63]	$\nu_{\text{prompt}}^{\text{atm.}}$ [56, 64–66]	$\mu_{\text{conv.}}^{\text{atm.}}$ [67–70]	all background
initial	160 ± 0.2 (190 ± 0.3)	400 ± 0.7 (490 ± 0.8)	580 ± 7	72 ± 0.1	8400 ± 110	9450 ± 110 (9540 ± 110)
final	6.4 ± 0.02 (4.0 ± 0.02)	0.3 ± 0.02 (0.2 ± 0.01)	0.1 ± 0.008	0.1 ± 0.001	0.01 ± 0.008	0.5 ± 0.02 (0.4 ± 0.02)

TABLE I. Expected number of events after initial and final set of selection criteria (including all corrections described in the text) for signal ($\nu_{\tau,CC}^{\text{astro}}$) and backgrounds, assuming IceCube’s flux from Refs. [53] and (in parentheses) [56]. About 85% of the estimated contribution from $\nu_{\text{prompt}}^{\text{atm.}}$ is from ν_{τ} . Signal and astrophysical background levels vary with the flux. The simulation did not include the self-veto effect [71] that would reduce the conventional (conv.) and prompt $\nu^{\text{atm.}}$ backgrounds. References to associated simulation packages are given; see text for details. Errors are statistical only, arising from finite simulation samples.

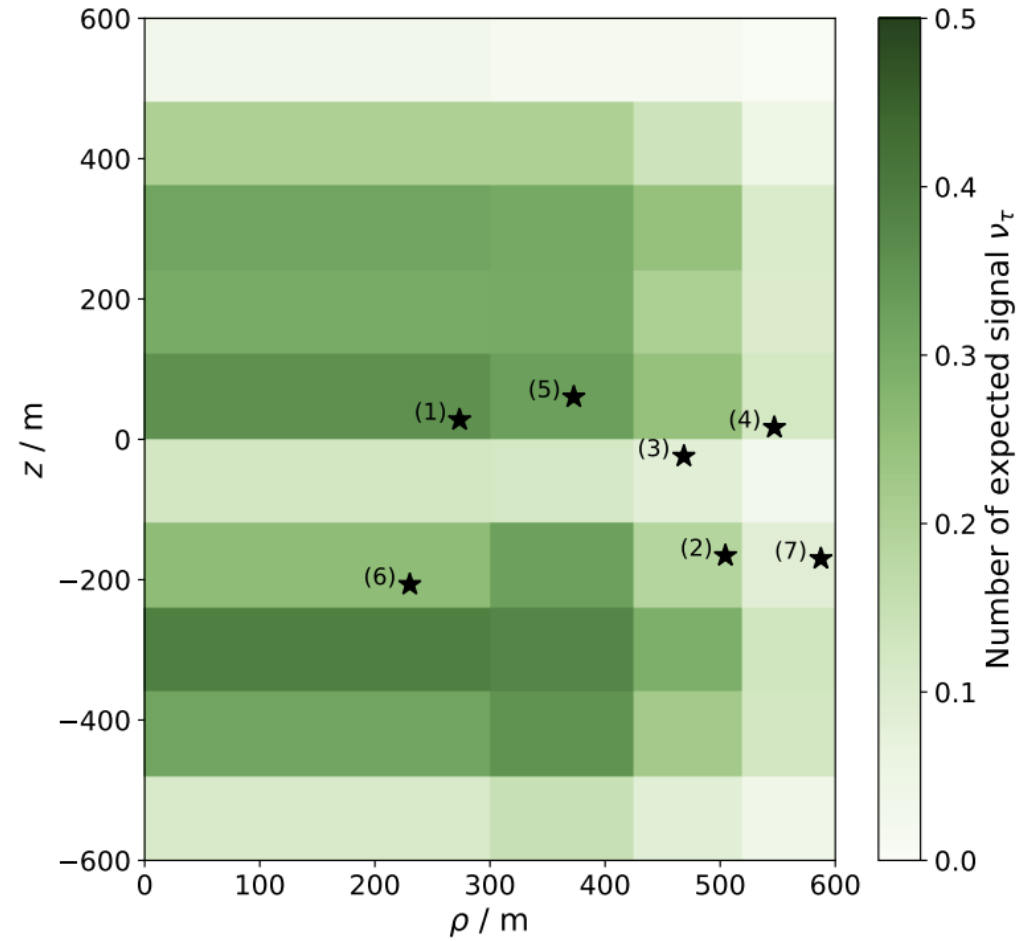
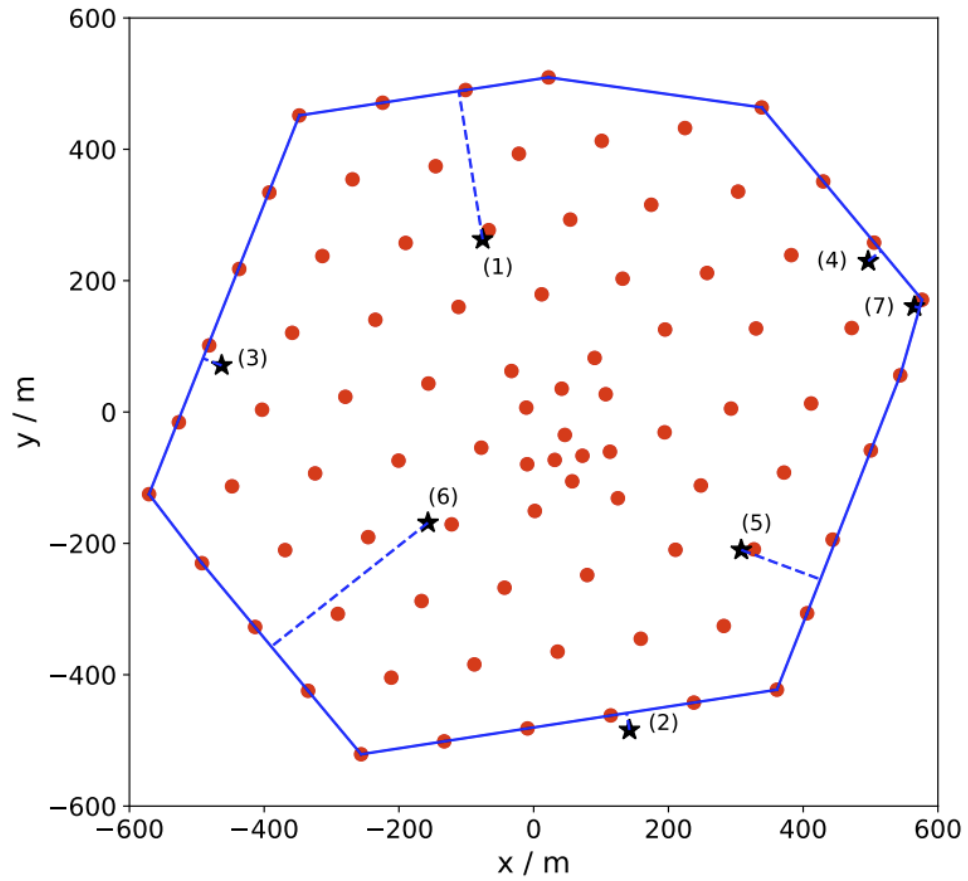
Reconstructed Signal Events



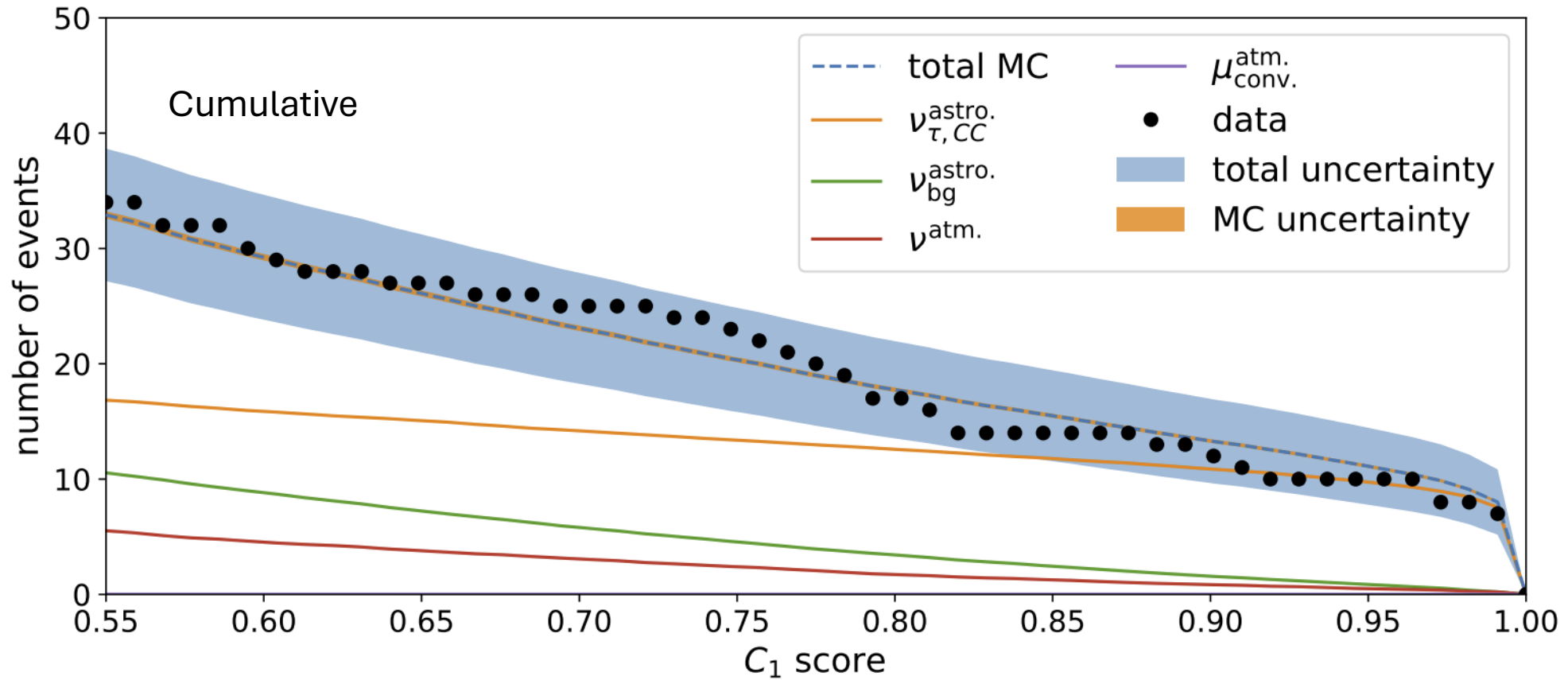
Spectrum



Spatial Distribution

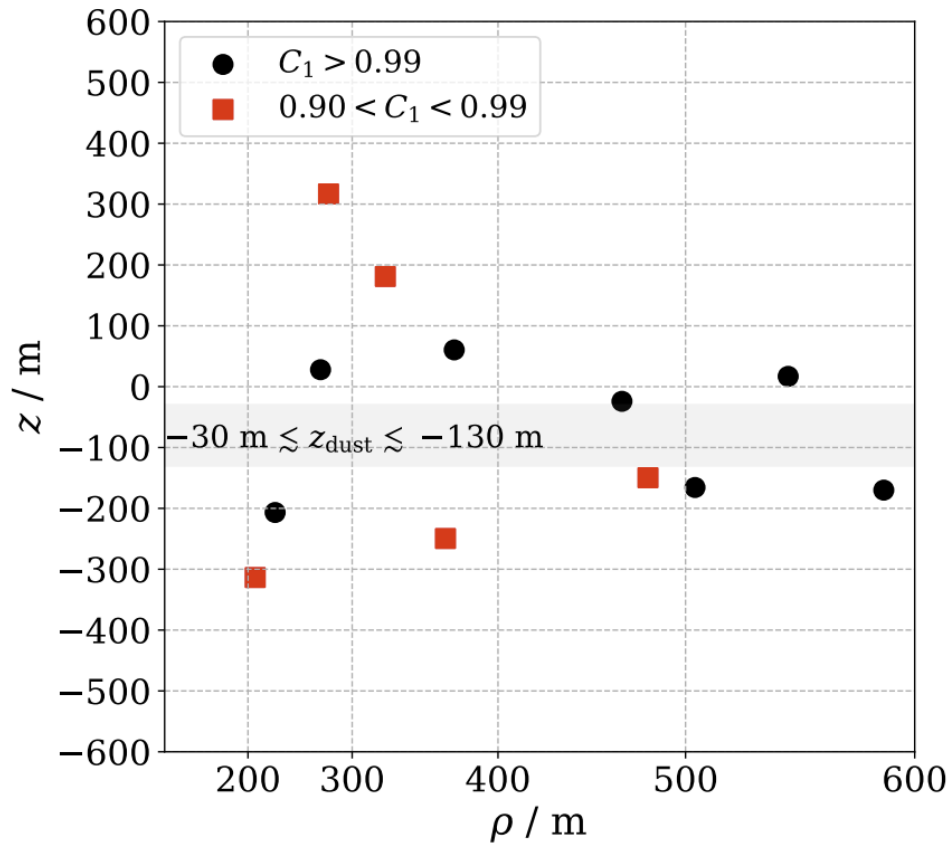


Classifier vs. data

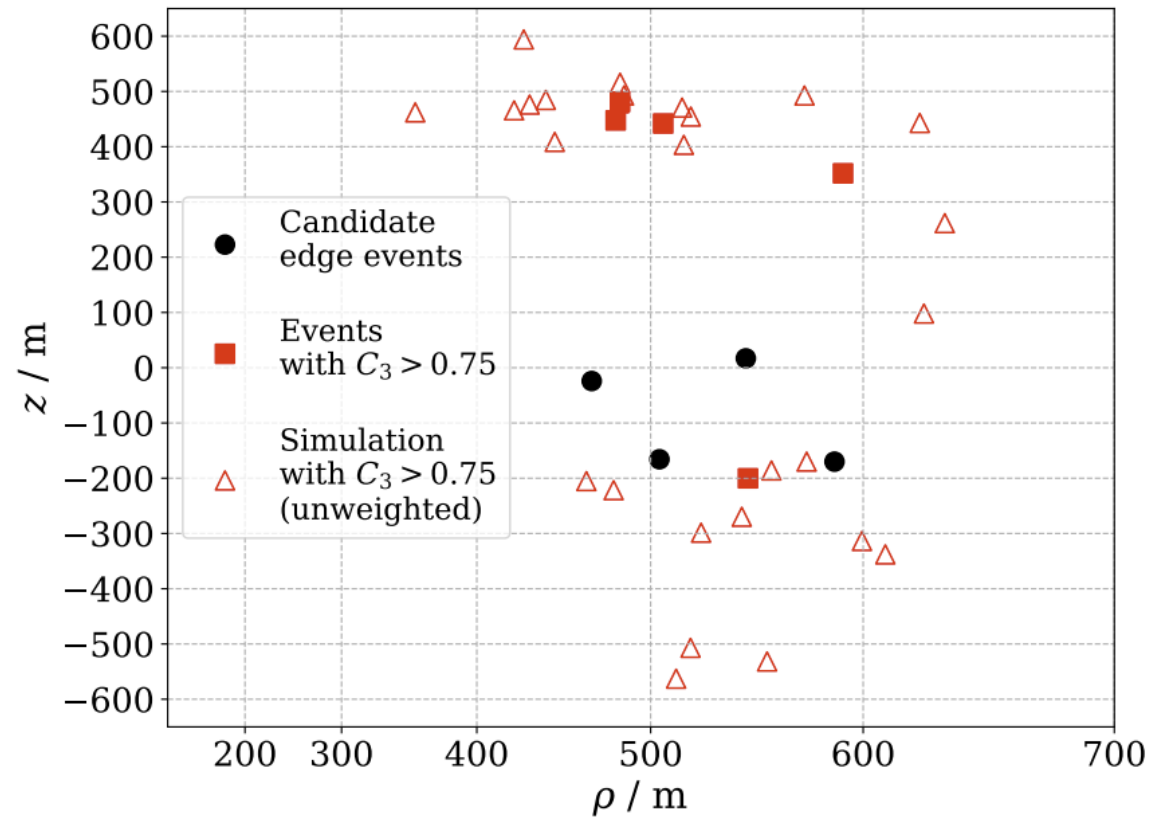


Relaxed Cuts

Net 1



Net 3



Adversarial Attack

Successful conversion of the one signal event to background

