

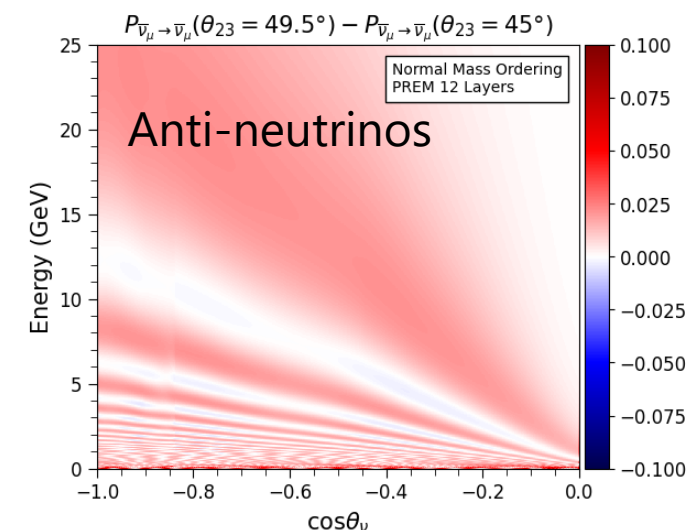
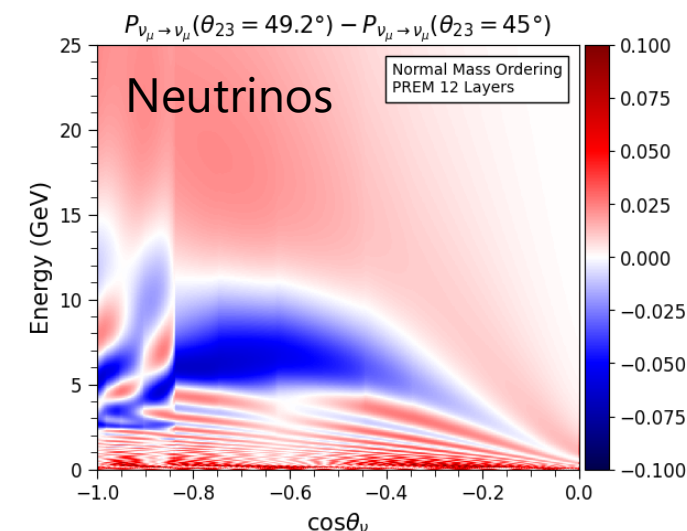
# Updates on Inelasticity and PID Reconstruction with 2D CNNs



Josh Peterson  
2023 IceCube Collaboration Meeting  
Oscillations Parallel Session

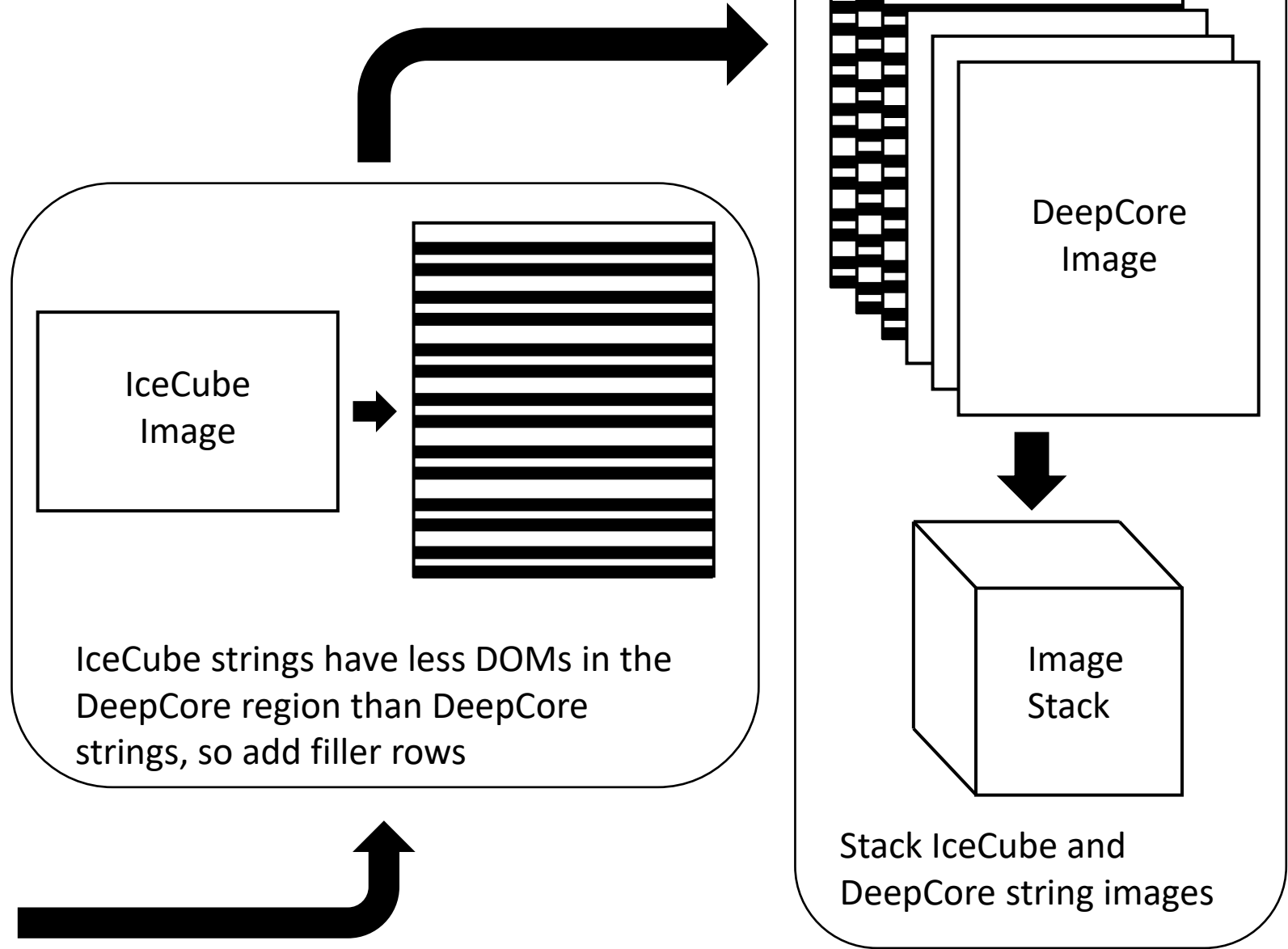
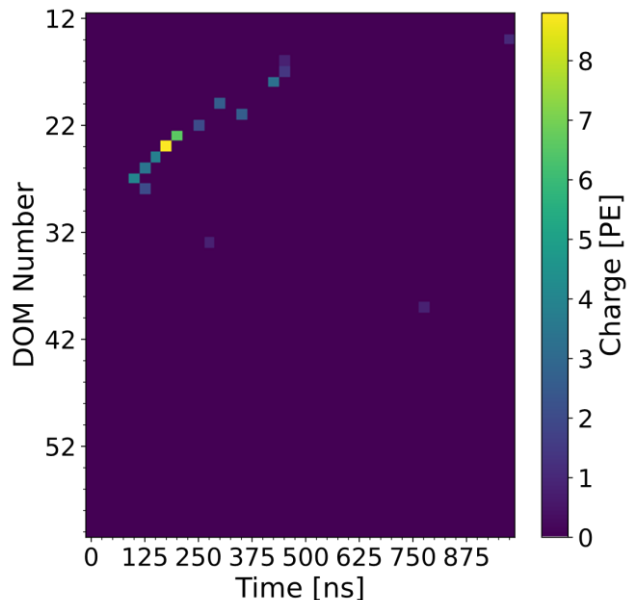
# Motivation

- **Maria and I developed a deep convolutional neural network for PID, I then used the same model for inelasticity**
  - Links to presentations on these networks: [Link](#), [Link](#)
- **PID classification is one of the main limiting factors in current oscillation studies**
- **Inelasticity could be used to statistically separate neutrinos from antineutrinos**
  - MSW resonance only occurs for neutrinos or antineutrinos depending on the NMO
  - This could help with studies about NMO, tomography, breaking the octant degeneracy of  $\theta_{23}$ , BSM studies at higher energies ([arXiv:1303.0758](#), [arXiv:1406.3689](#))



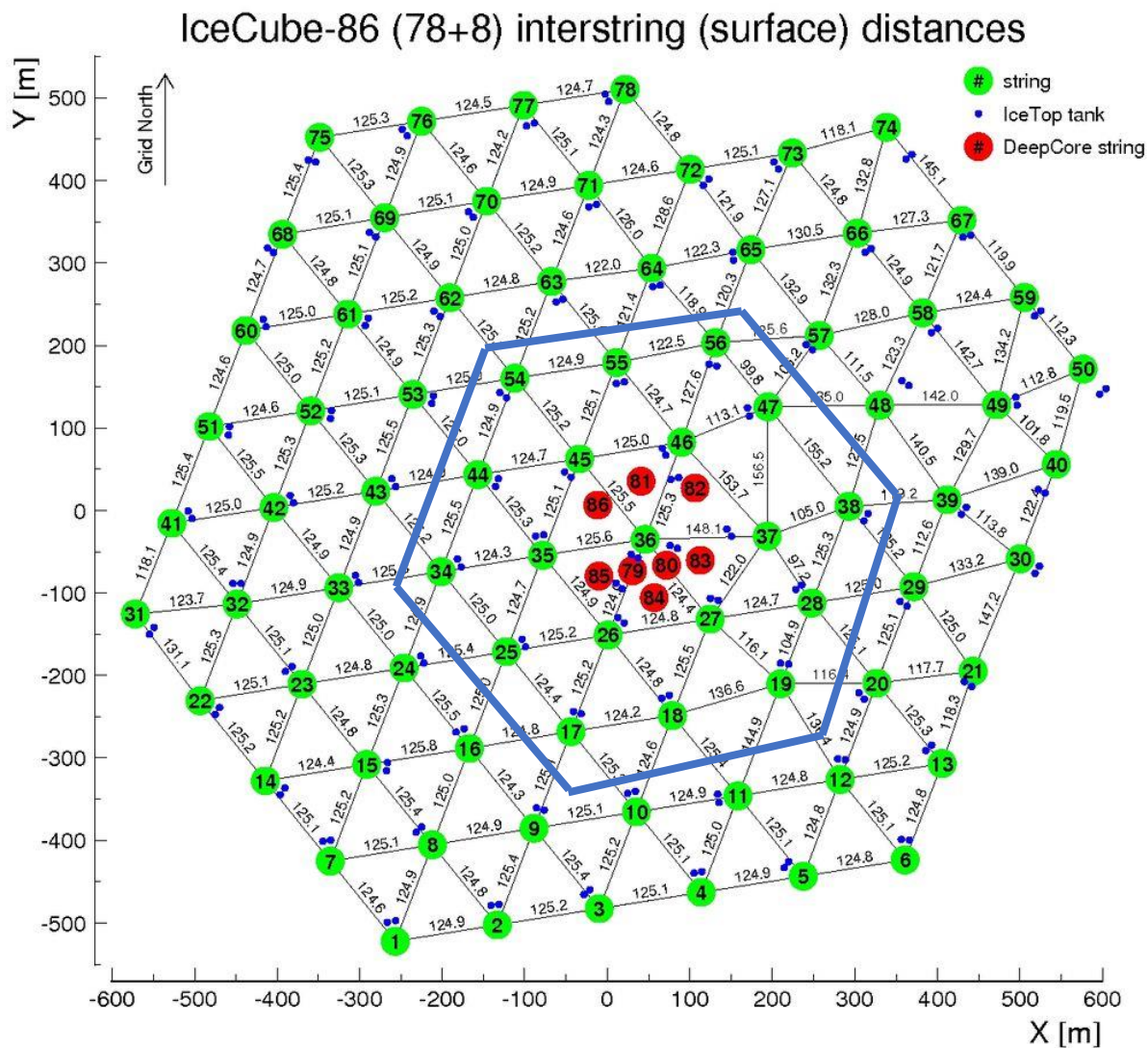
# Data Input

- Each row represents a DOM on the string
  - Only use DOMs below the dust layer
- Each column is a slice in time
- The value of each pixel represents the total charge detected



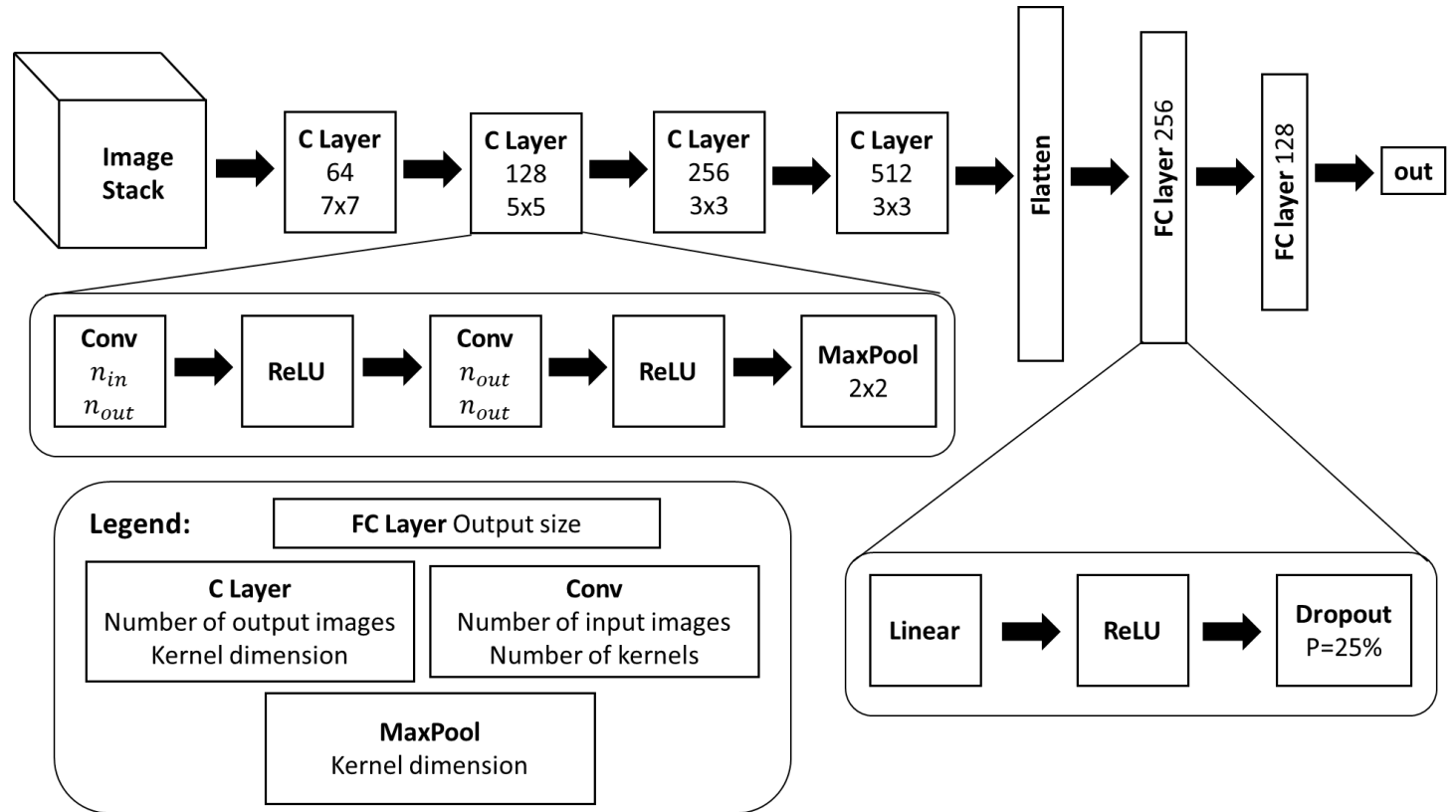
# New Geometry

- I use DeepCore and the IceCube strings that surround DeepCore (shown in the blue hexagon)
- Only DOMs below the dust layer are used
- In later slides, I will refer to this portion of the detector as “DeepCore Plus”

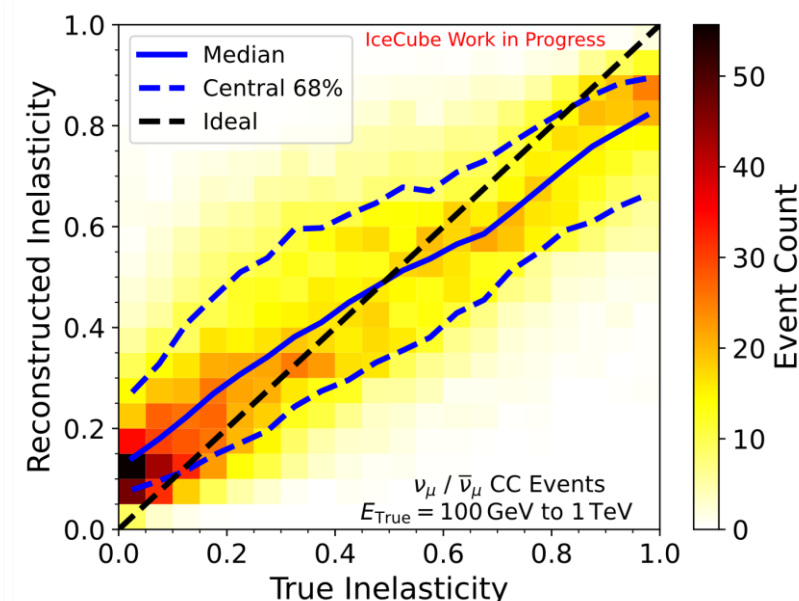
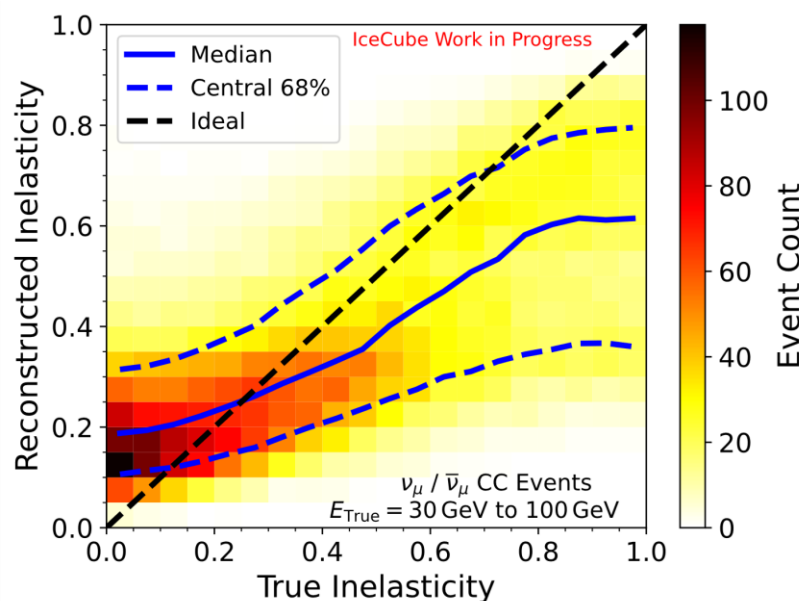
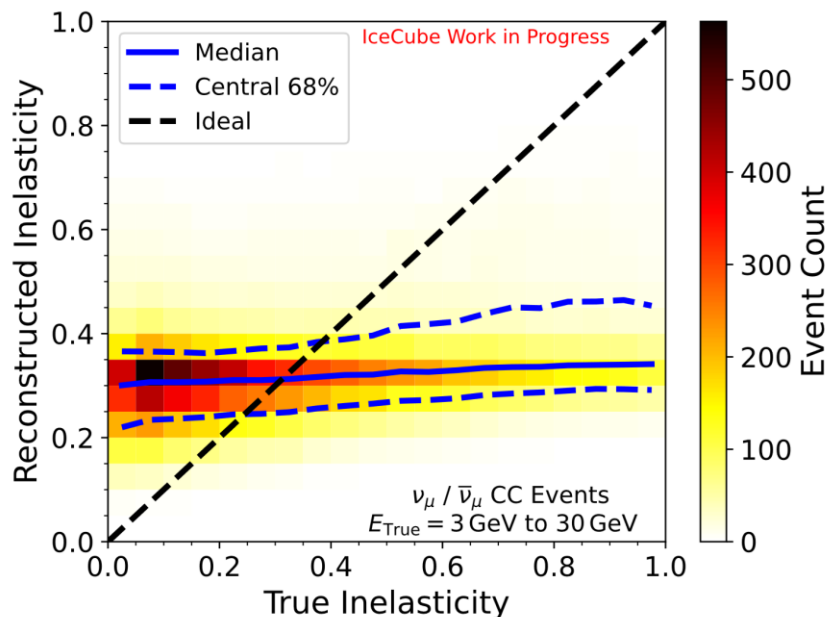


# CNN Information

- **10 layer convolutional neural network**
  - 8 convolution layers
  - 2 fully connected layers
- **Trained on level 3 data with some simple charge and containment cuts**
  - Low level data chosen for larger training set
  - Can apply network to higher levels
- **For inelasticity:**
  - Train with  $\nu_\mu$  events
  - Utilize the L1 loss
- **For PID:**
  - Train with  $\nu_\mu$  and  $\nu_e$  events
  - Utilize a modified binary cross entropy loss



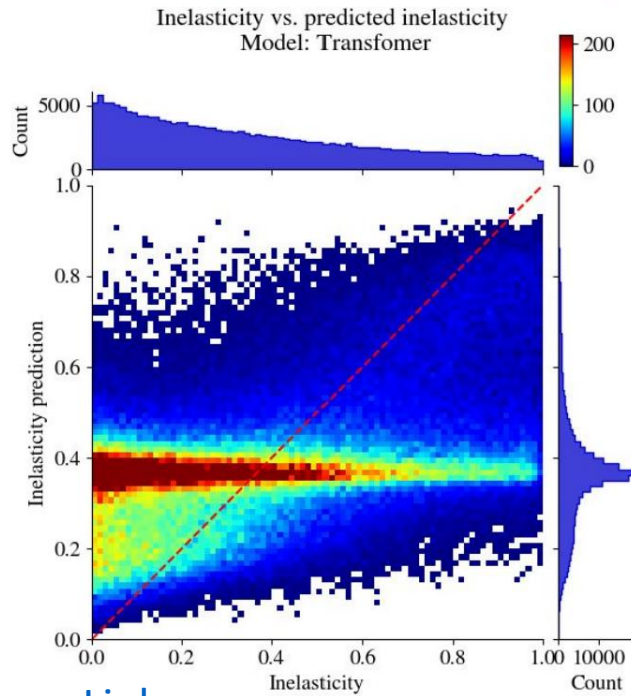
# y Results at Different Energies



- **Failure to reconstruct y for events below 30 GeV**
  - Matter effects with standard three flavor oscillations are below 20 GeV
- **Reconstruction performance improves with larger energies (expected)**

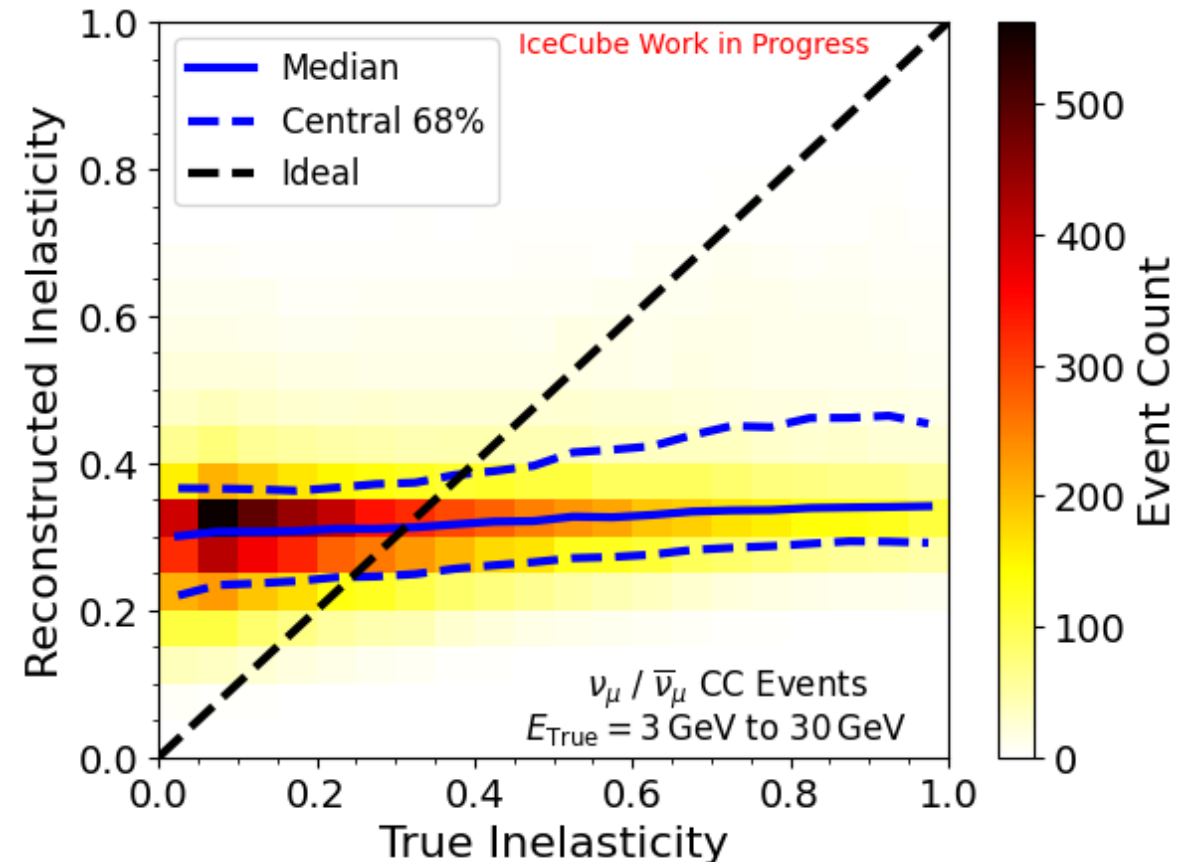


# Why a Flat Distribution?



[Link](#)

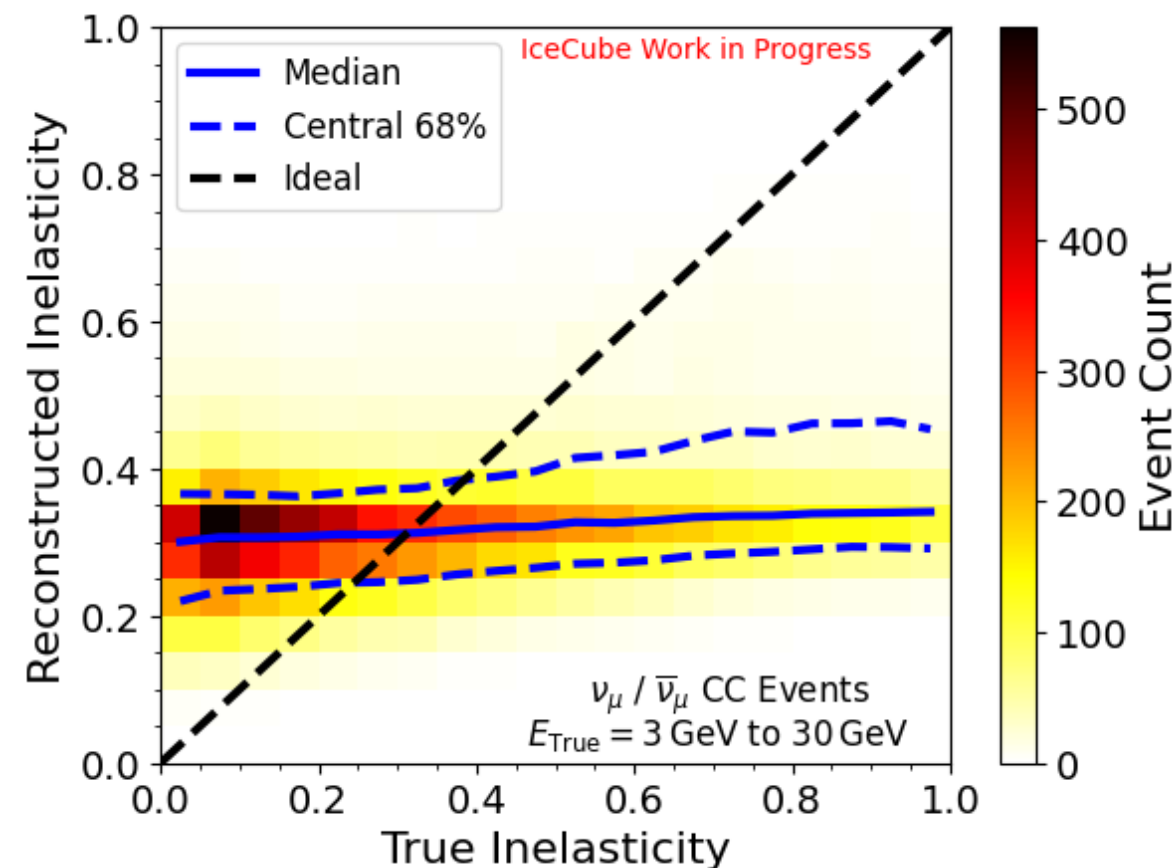
- **Below 30 GeV I get a mostly flat distribution**
  - Seeing similar behavior with transformers
- **Average inelasticity of the training set: 0.36**
- **Likely, there just isn't enough information for effective reconstruction and so the optimal thing to do is guess the mean**



# $\nu_\mu / \bar{\nu}_\mu$ Separation

- **The reconstructed inelasticity distribution isn't completely flat**
  - May have some neutrino/antineutrino separation power with just two reconstructed inelasticity bins
- **Bin the track events with true energies from 3 GeV to 30 GeV into two bins, above and below the mean reconstructed inelasticity**

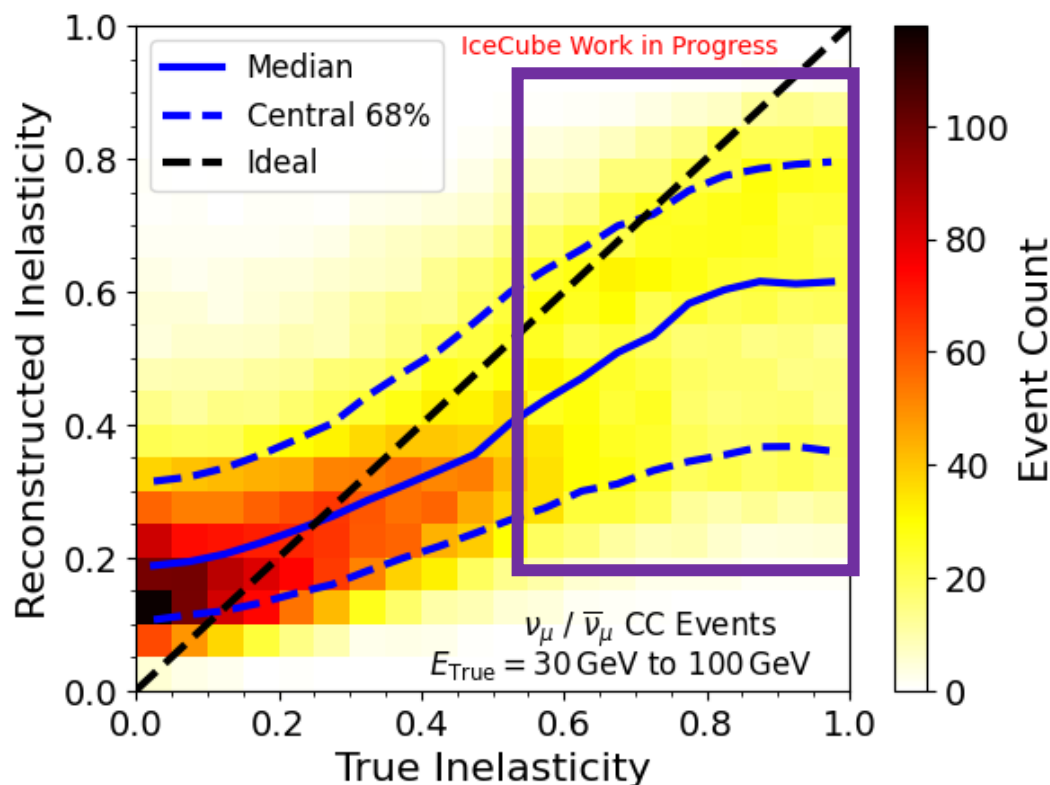
Bin	$\bar{\nu}_\mu / (\bar{\nu}_\mu + \nu_\mu)$
$> y_{reco} = 0.33$	0.30
$< y_{reco} = 0.33$	0.33



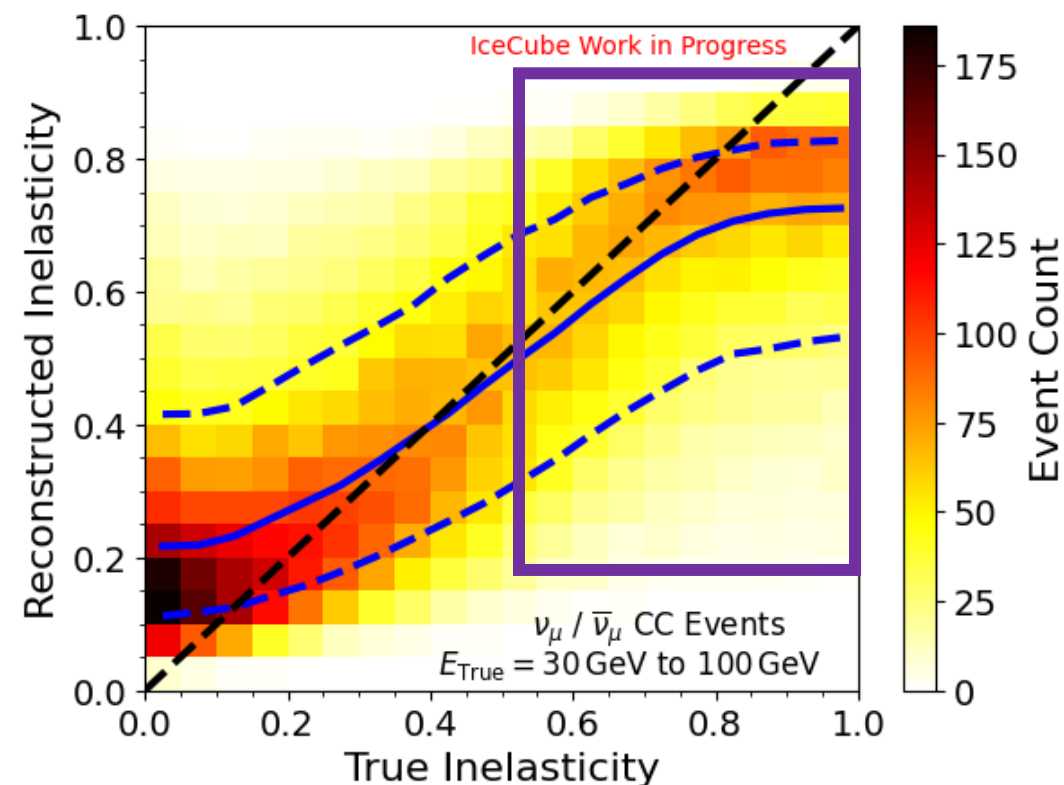


# Improving Higher Energy Reconstruction

Training set energy: 3 GeV to 1 TeV

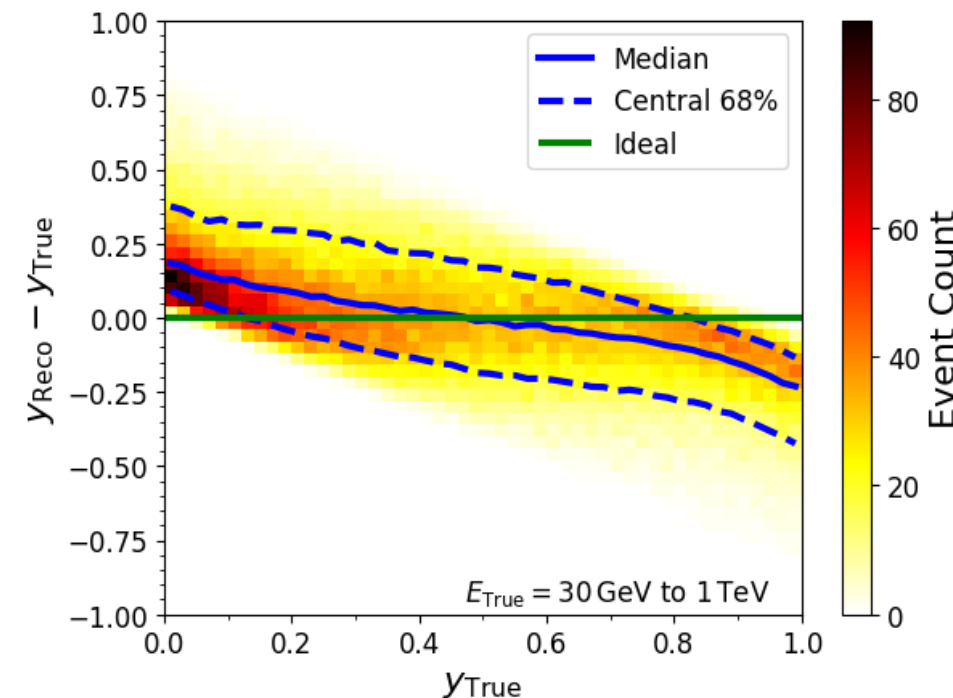
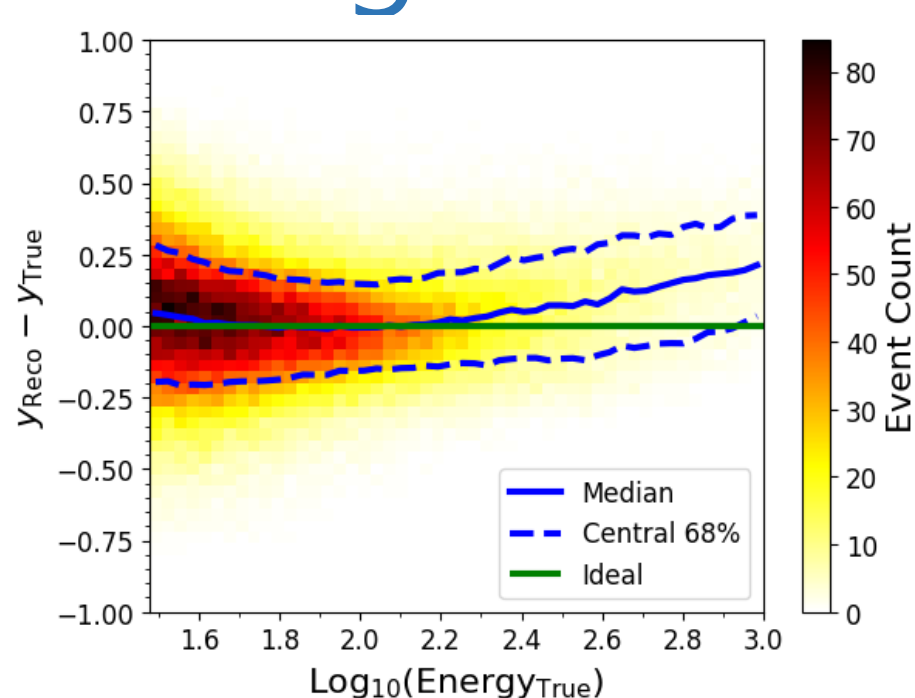


Training set energy: 30 GeV to 1 TeV



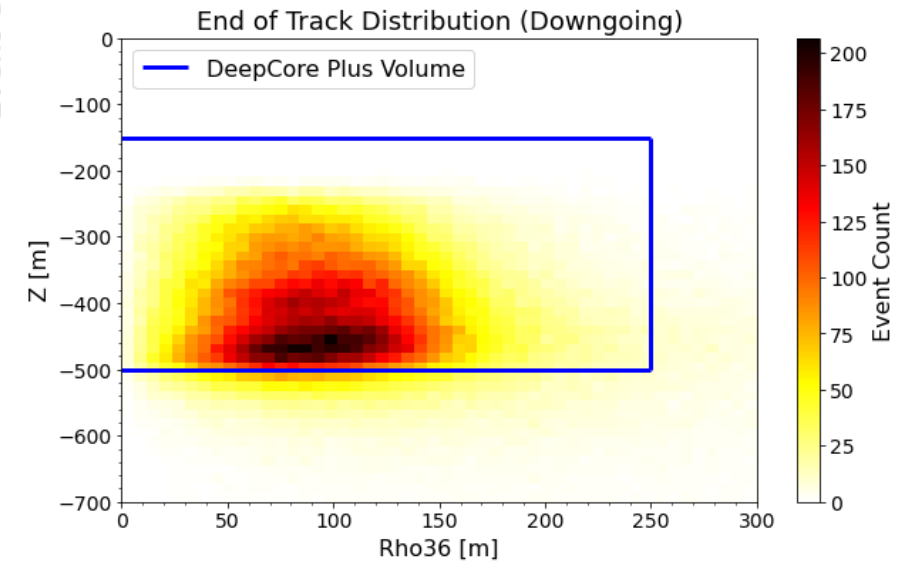
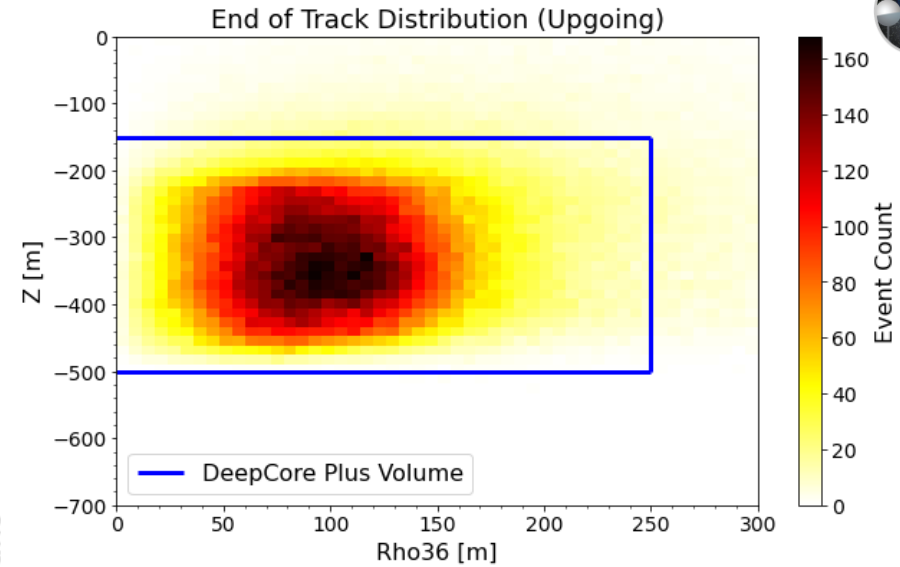
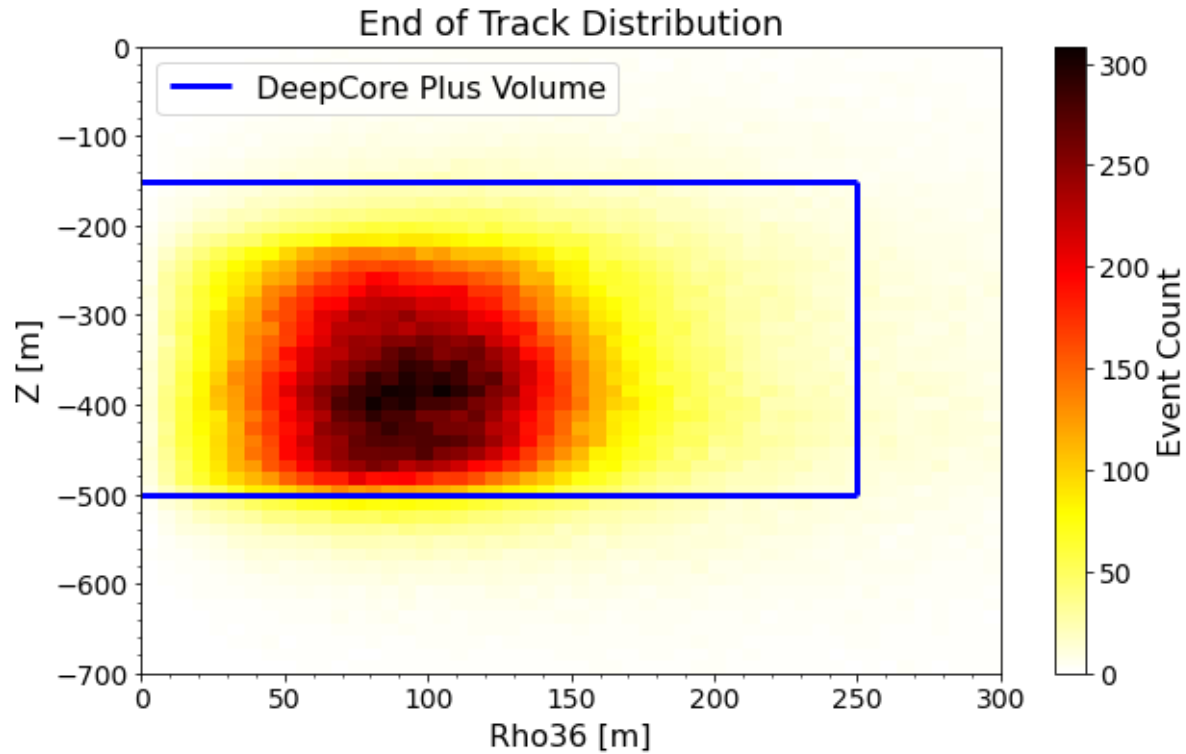
- Definite improvement in reconstruction of larger inelasticities when we remove events with energies below 30 GeV from the training set

# Checking Difference (Above 30 GeV)



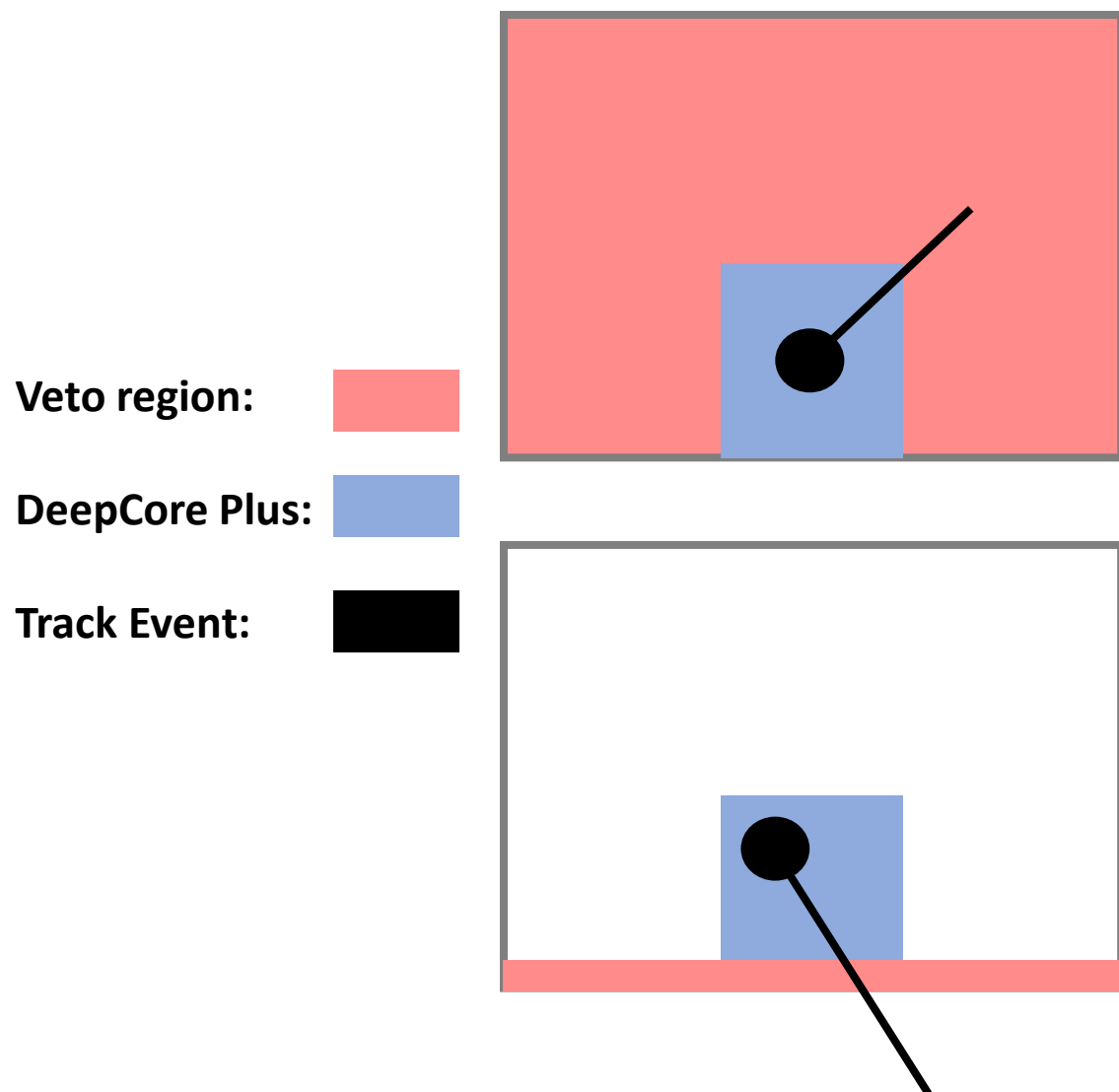
- **The difference vs. energy plot (left) shows a skewing of reconstruction above 300 GeV**
  - Skewing is towards higher inelasticity values, suggesting that the outgoing muon is escaping, which means we see less of the track
  - This gives an upper-end energy limit on what we can do with the "DeepCore Plus" volume
- **The difference vs. true inelasticity plot shows a "squeezing" of the CNN output**

# Escaping Muons



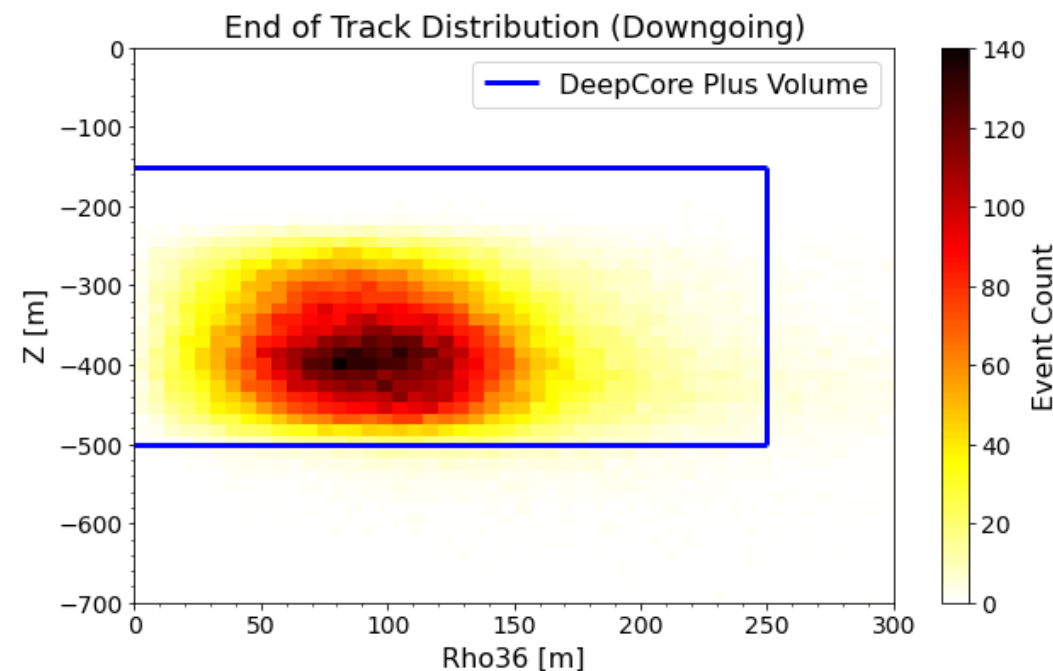
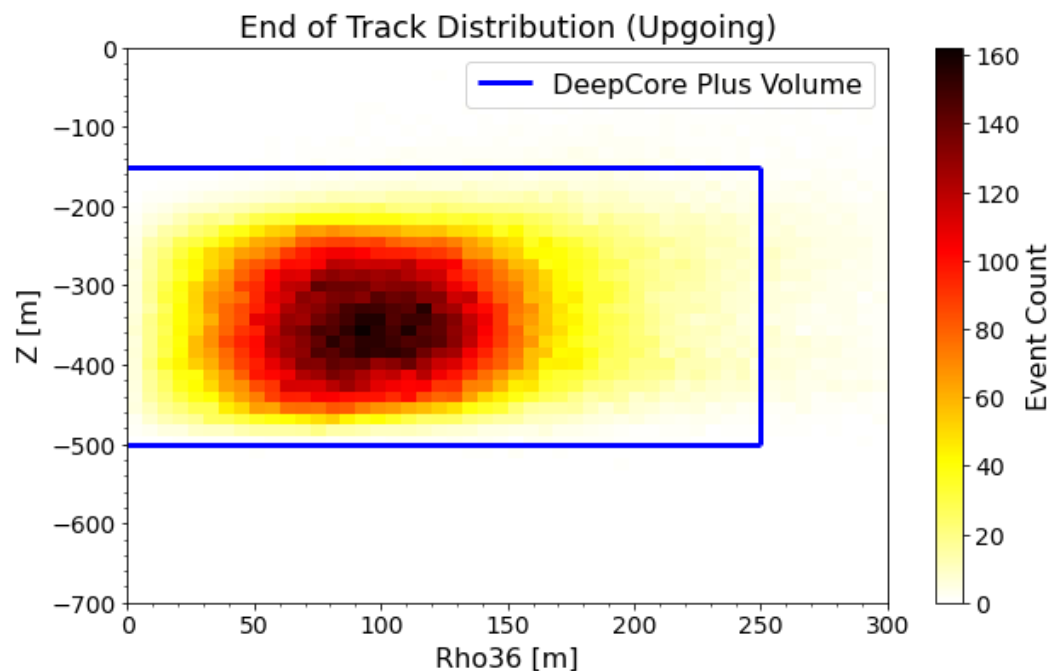
- When we plot the end of the muon tracks, we see tracks escaping the detector, especially down going events

# Containment Cuts



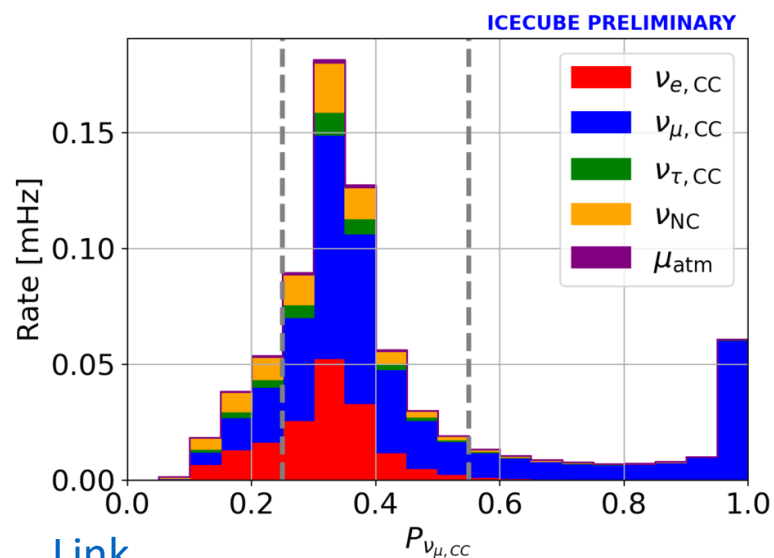
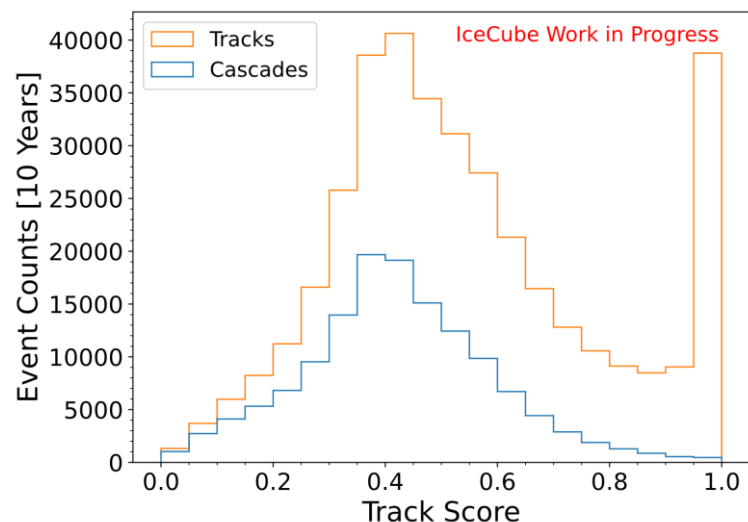
- **Since we are only using part of the detector, we can use the rest of the detector as a veto region**
  - Calculate the ratio of charge in DeepCore Plus to total charge, cut harshly on that ratio (I used 0.99)
- **Use the bottom layer of DOMs as a veto region for down going events. If any charge is seen on the bottom layer, we cut the down going event**
  - Cannot make charge cuts on muons that escape out the bottom of the detector, since there aren't any DOMs to pick up charge

# Cut Results



- **With these new cuts we see significantly less events escaping the DeepCore Plus volume**
- **Currently training CNNs for PID and  $y$  with FLERCNN Level 6 & 8 MC sets using these new cuts**
  - Also limiting the energy to 300 GeV

# Many Reconstruction Algorithms

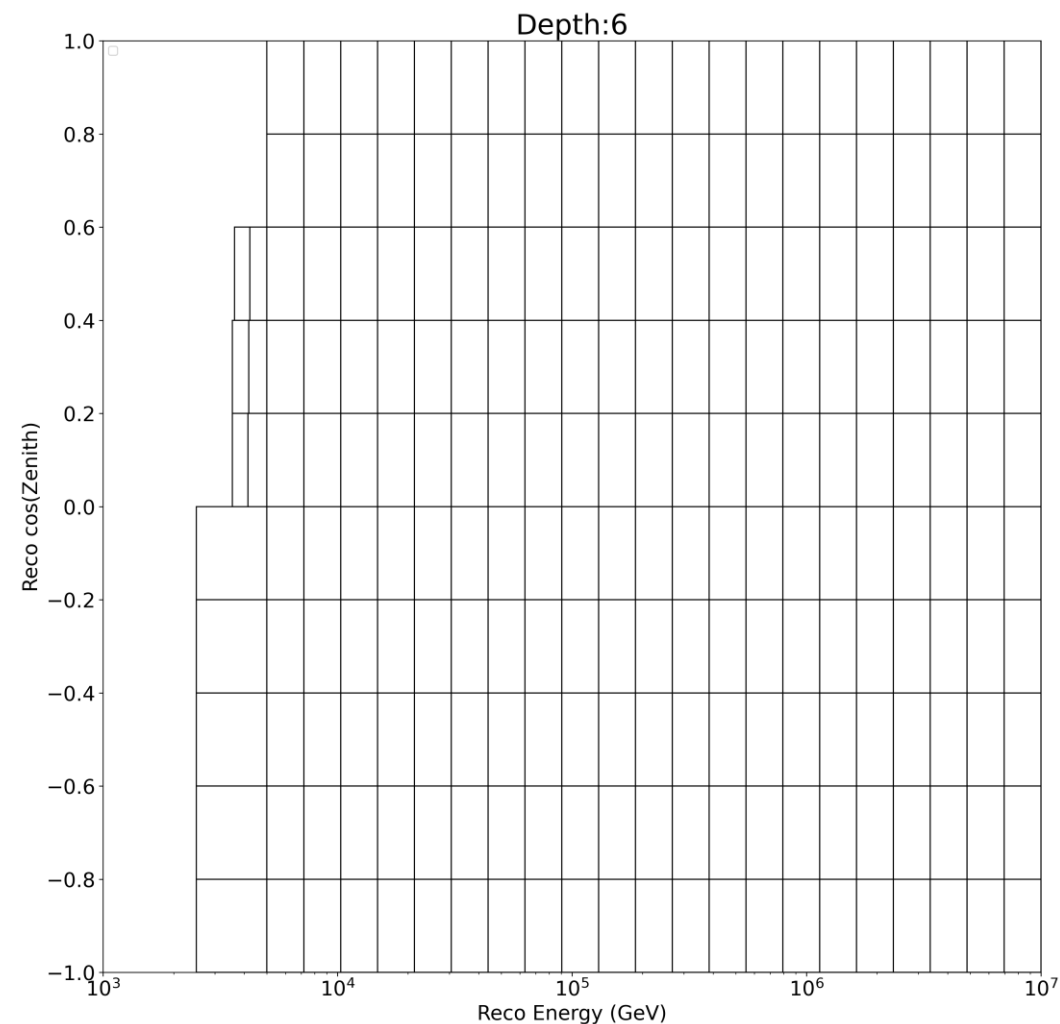


[Link](#)

- **At this point in time, we have multiple reconstruction algorithms for PID, Energy, etc.**
- **Rather than sticking with one, it might be worth using multiple**
  - Stacked Generalization – Use outputs of many classifiers / regressors as input for another classifier / regressor
  - Voting – select classification based on majority vote
  - Averaging
- **Going to investigate combining FLERCNN and my 2D CNNs for improved PID classification**

# Binning w/ Decision Trees

- **Can use a decision tree to generate binning schemes**
- **Bin sizes correspond to the amount of information available in that region of parameter space**
  - Can enforce minimum bin counts, good for hypersurfaces!
- **Vedant Basu and I are working on figuring out how to encode physics information into the generation of the decision trees**
- **Maybe applicable to oscillation analyses as well?**

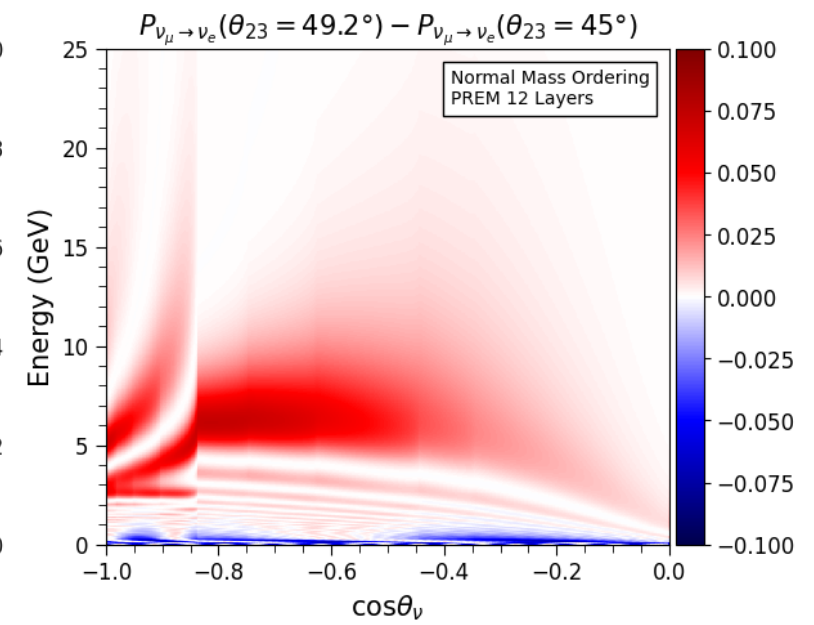
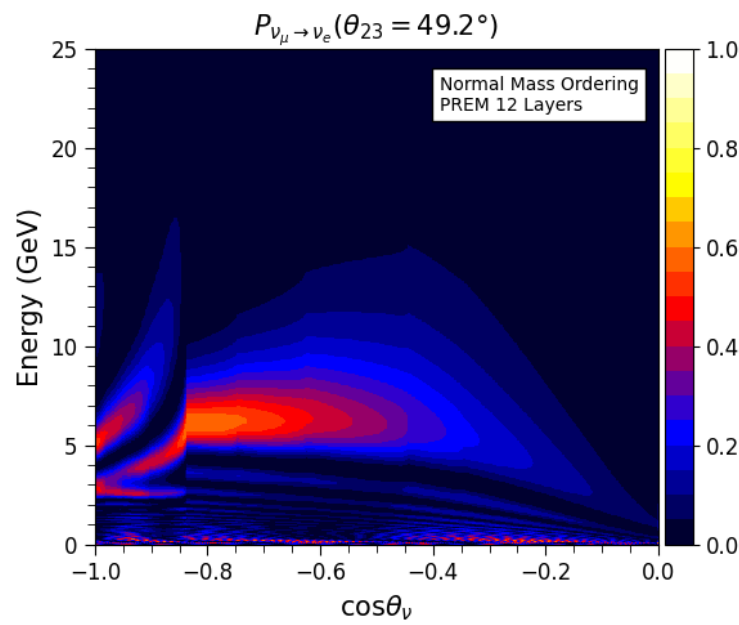
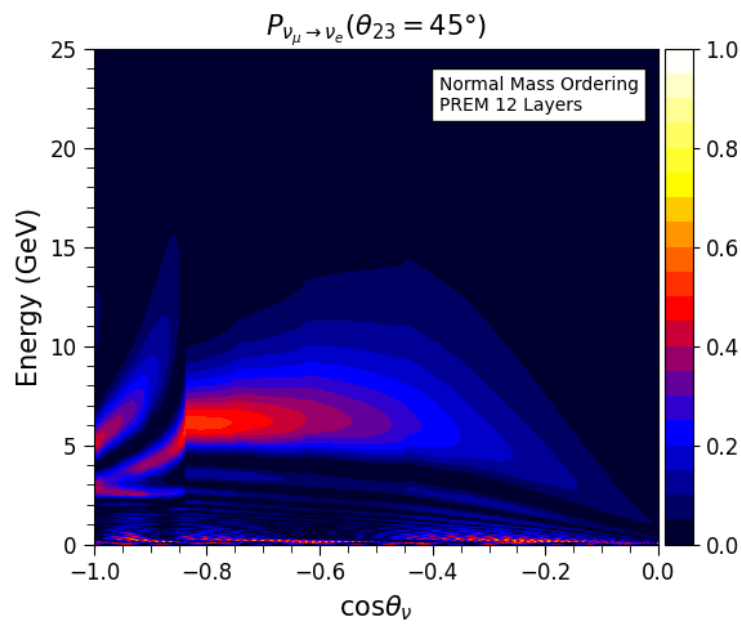




# Summary

- **Have trained 2D CNNs for inelasticity and PID reconstruction**
  - Can reconstruct inelasticity above 30 GeV
  - Minimal neutrino / antineutrino separation below 30 GeV
- **Determined better energy cuts and developed new containment cuts for training**
- **Coming in the near future:**
  - CNNs trained with new energy and containment cuts
  - Combining FLERCNN and my 2D CNNs for PID
  - Binning with decision trees?

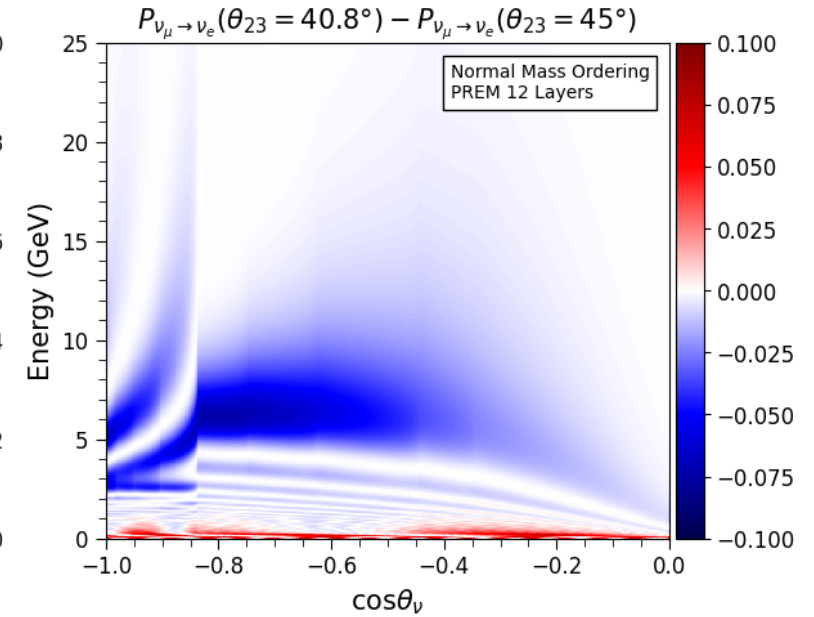
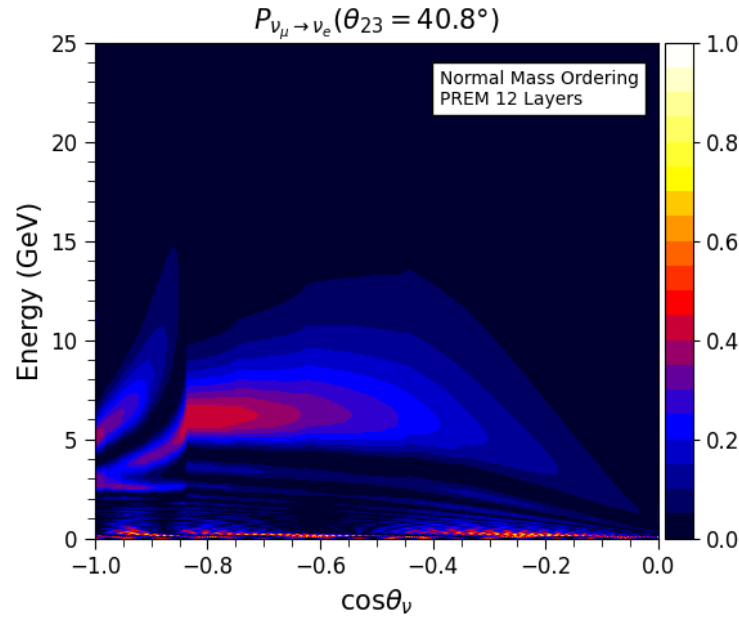
# Backup Slides



- $\nu_\mu \rightarrow \nu_e$  probability increases when  $\theta_{23} > 45^\circ$  and decreases when  $\theta_{23} < 45^\circ$   
 $\Rightarrow$  suppression of track events for  $\theta_{23} > 45^\circ$   
 $\Rightarrow$  enhancement of track events for  $\theta_{23} < 45^\circ$

- $P_{\mu e} = 4s_{13}^2 s_{23}^2 \frac{\sin^2(A-1)\Delta}{(A-1)^2} + H.O.T.$

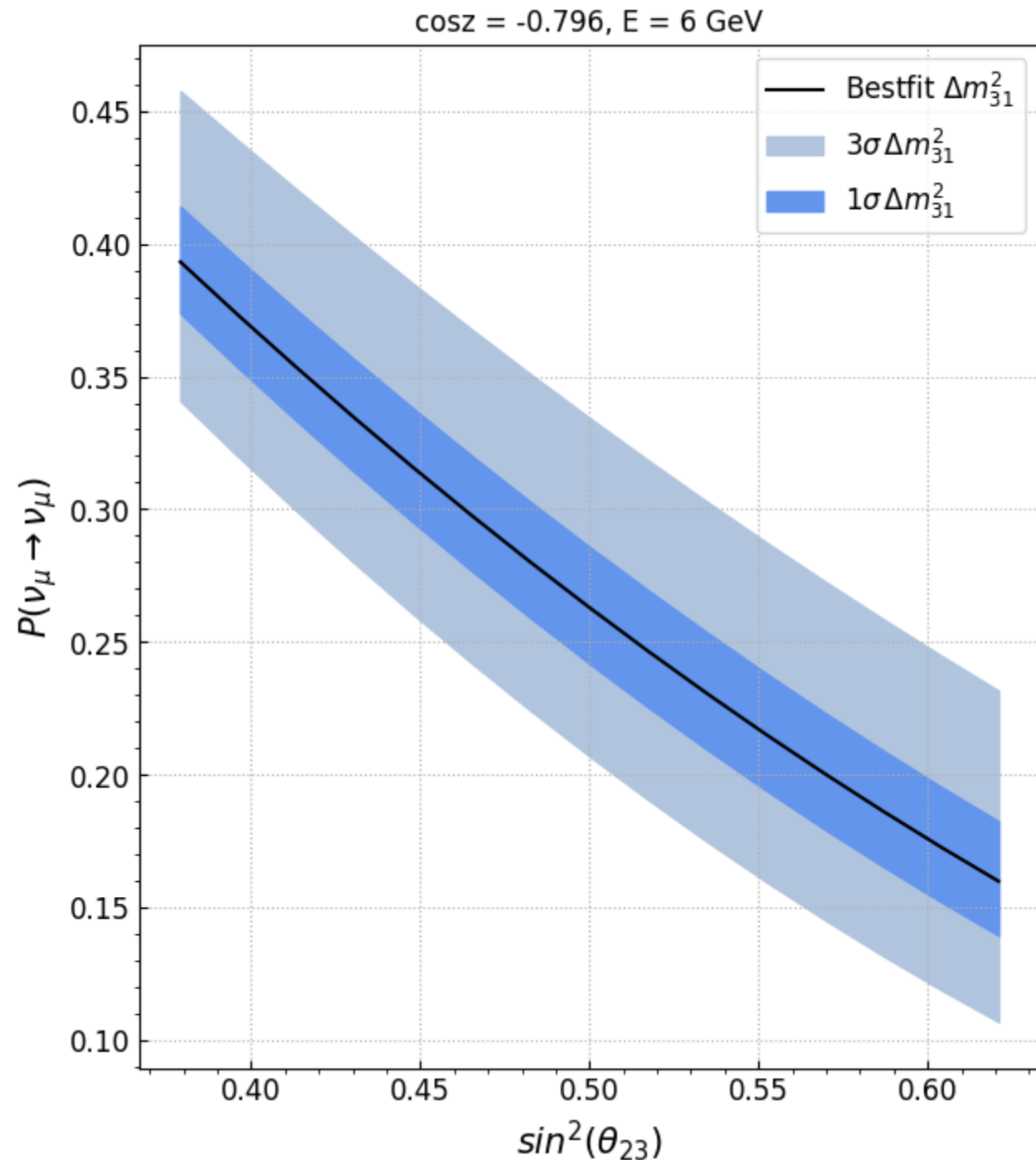
- Leading probability term is *asymmetric* about  $\theta_{23} = 45^\circ$



$$A = \frac{2E}{\Delta m_{31}^2} (7.56 \times 10^{-14} \text{eV}) \left( \frac{\rho(x)}{g/cm^3} \right) Y_e(x)$$

# Breaking the Octant Degeneracy

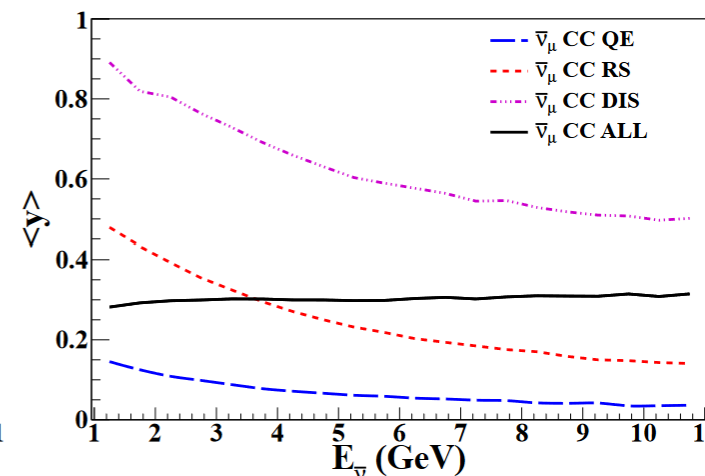
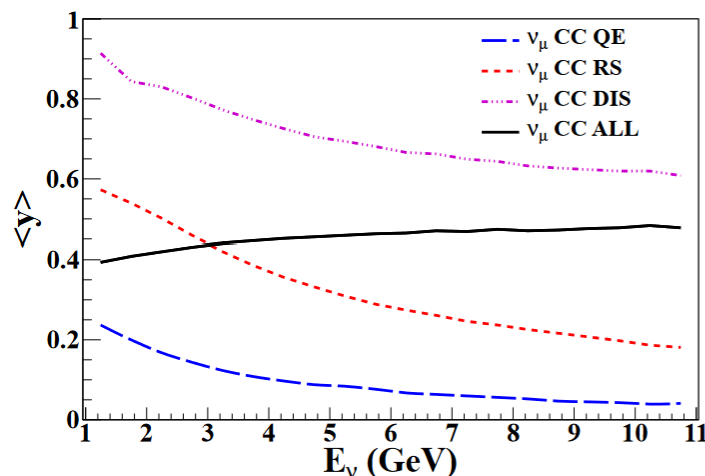
- Referring to the oscillogram difference plots on slide 2, I wanted to check the baseline corresponding to the maximum difference from  $\theta_{23} = 45^\circ$
- $\Delta m_{31}^2$  seems to have a significant impact on the survival probability, and introduces degenerate solutions
- The next plots will show the degeneracy gets worse with increasing energy



# Inelasticity

- Bjorken  $y := \frac{p_2 \cdot (p_1 - p_3)}{p_2 \cdot p_1}$
- In the high energy limit:

$$y = \frac{E_\nu - E_\mu}{E_\nu}$$



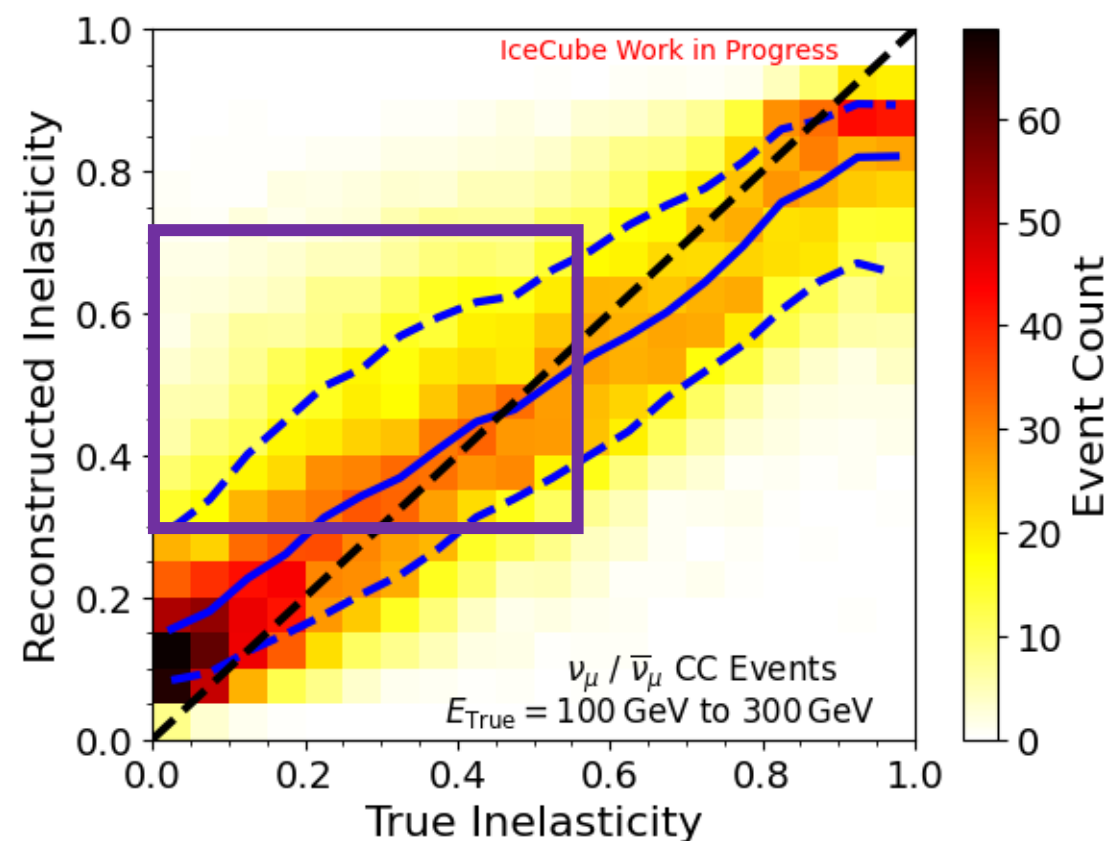
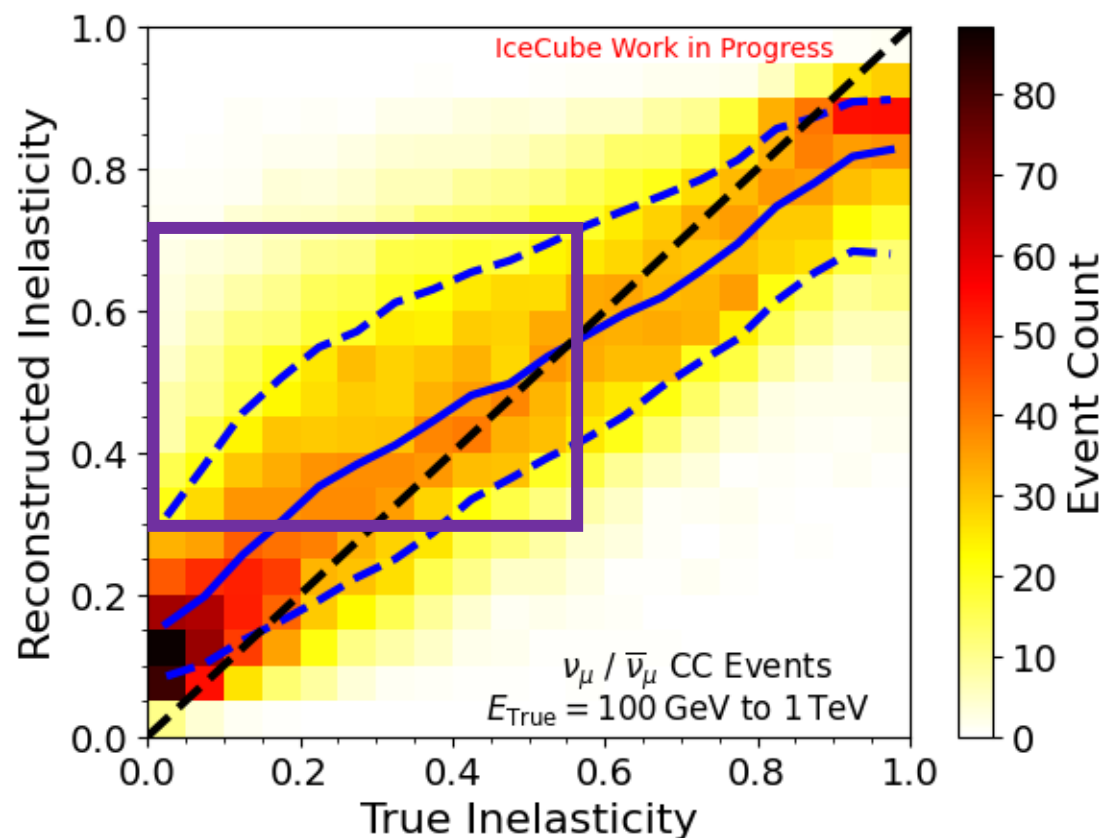
[arXiv:1406.3689](https://arxiv.org/abs/1406.3689)

- Neutrinos have a higher average inelasticity than antineutrinos
- At GeV energies, the outgoing muon is a minimum ionizing particle  $\Rightarrow$  average track length is directly proportional to muon energy
  - Determining inelasticity could then also be determined through topology, so a natural candidate for our CNN

# Other Things I Tried That Didn't Help

- Jason Koskenan suggested I try having the NN output a beta probability distribution instead of a measured  $y$ 
  - NN outputs the concentrations,  $\alpha$  and  $\beta$ , of the beta distribution
  - The loss you minimize is  $-\text{Log}(\text{Beta}(y_{\text{true}}; \alpha, \beta))$
  - Your measurement is then the mode of the distribution
- I tried incorporating spatial information through additional channels per image, and then you convolve with a 4-dimensional kernel
  - Kind of strange since I'm only adding information

# Impact of $>300$ GeV Events



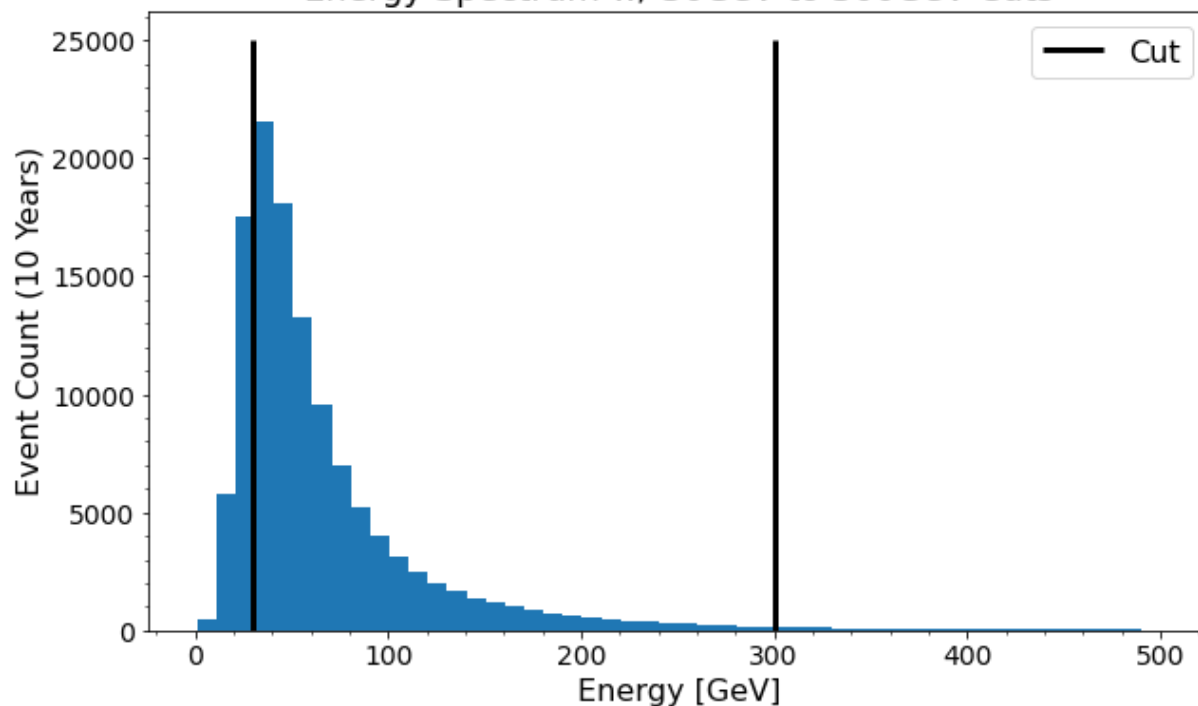
- Better fit at lower inelasticity values



# Cut Results

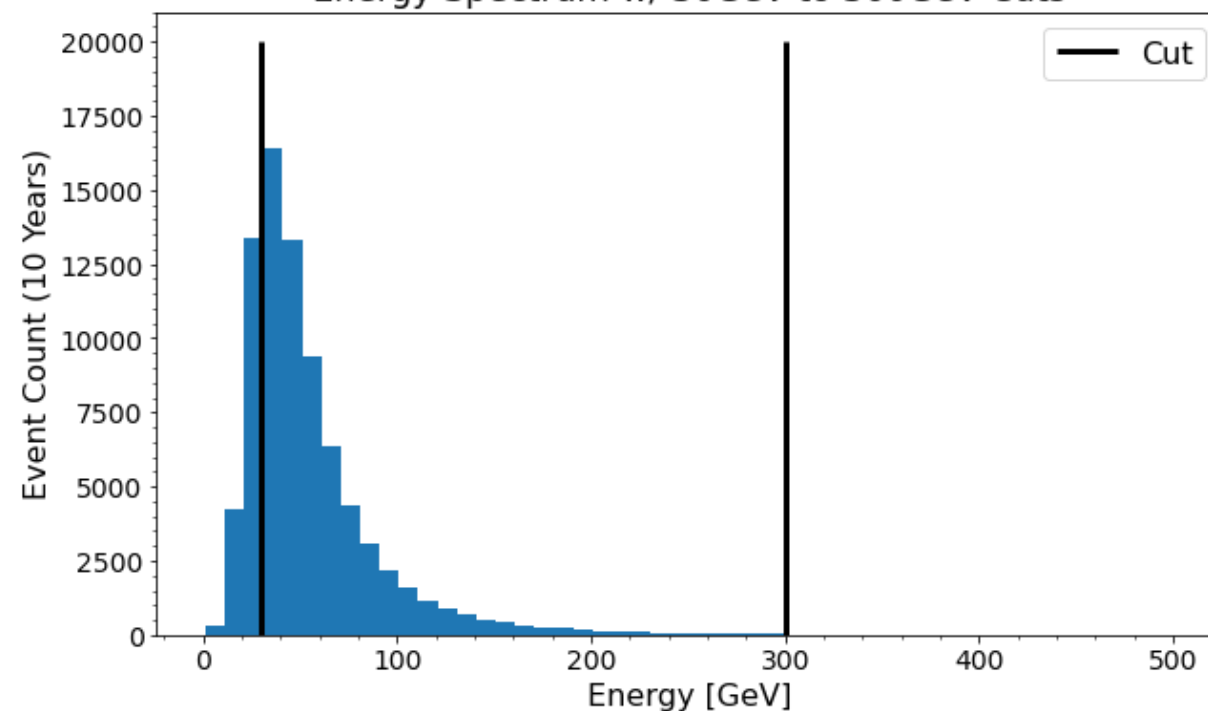
## Before containment cuts

Energy Spectrum w/ 30GeV to 300GeV Cuts



## After containment cuts

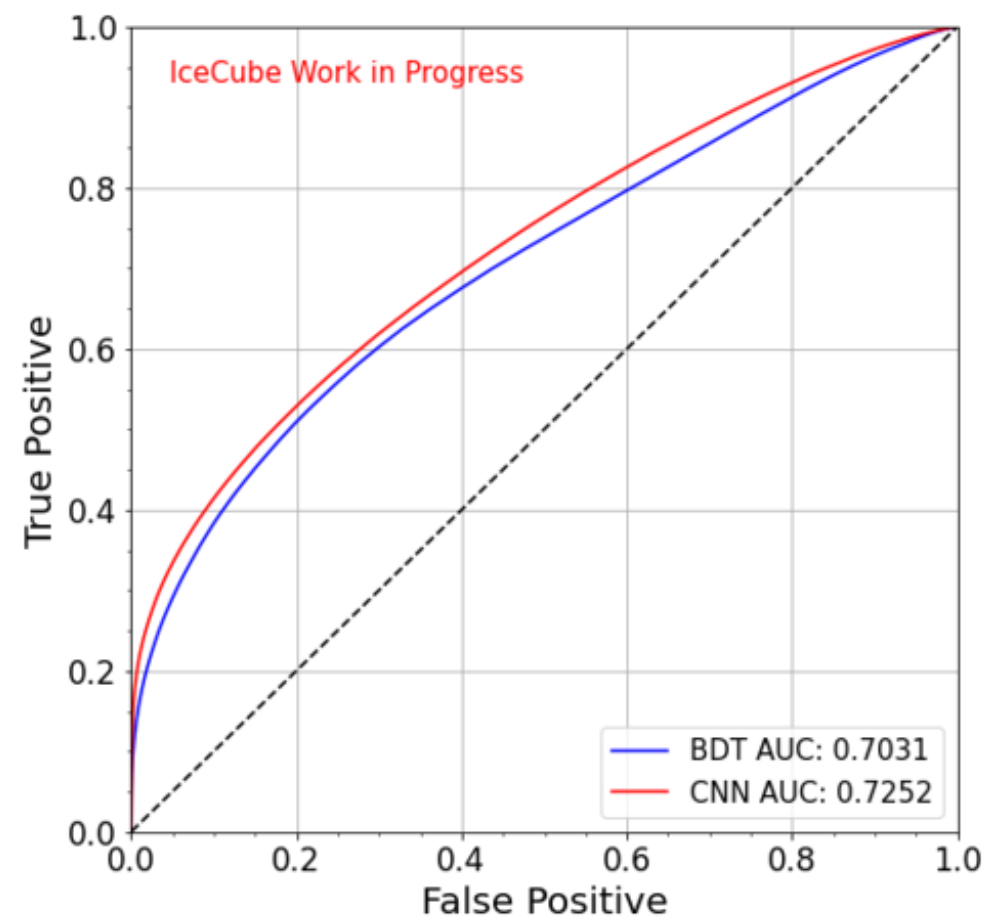
Energy Spectrum w/ 30GeV to 300GeV Cuts



- With the new cuts we also see a significant reduction in the tail of the distribution towards high energies

# PID Results

- Receiver Operating Characteristic (ROC) Curves for the 2D CNN and BDT
- Plots include numu, nue, and nutau events
  - Nutau events are labeled as cascade
- Events are weighted with the atmospheric weights (assume 10 years of livetime)



# PID Results

- Histogram of test set events binned by assigned track score
- Same events as the previous plot
- Events are weighted with the atmospheric weights (assume 10 years of livetime)

