



# Spatiotemporal Machine Learning for Atmospheric Neutrino Interactions in JUNO

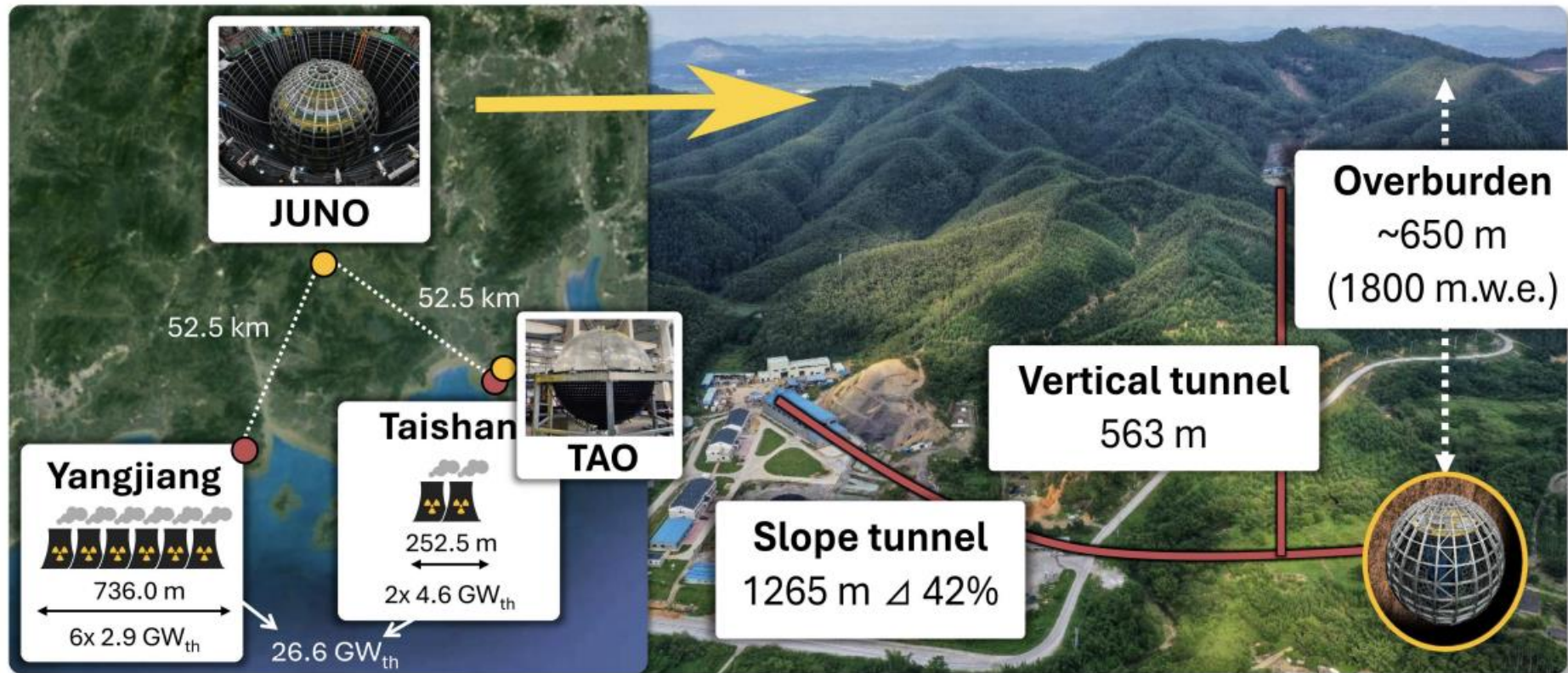
Liam Francis Jones

Supervisors: Prof. Costas Andreopoulos & Prof. Xianguo Lu



# JUNO Experiment

- Jiangmen **U**nderground **N**eutrino **O**bservatory (JUNO) is a next-generation multi-purpose, 20 kton Liquid Scintillator (LS) detector



# JUNO Detector

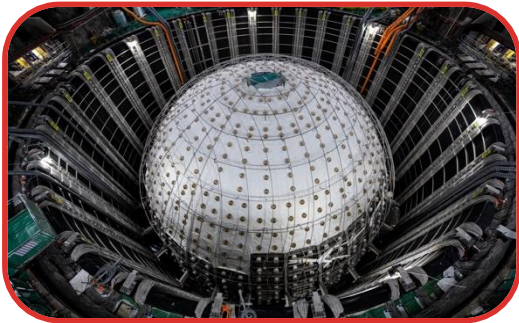
## Top Tracker

3 plastic scintillation layers  
~**650m** rock overburden



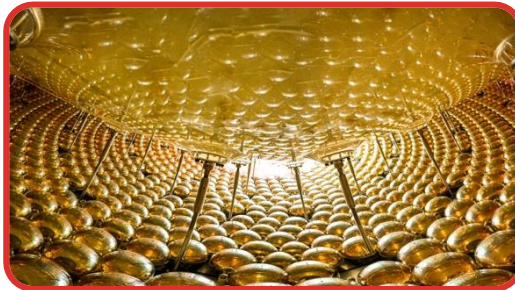
## Outer Cherenkov Detector

35 kton of ultrapure water  
>**2500** 20-inch PMTs



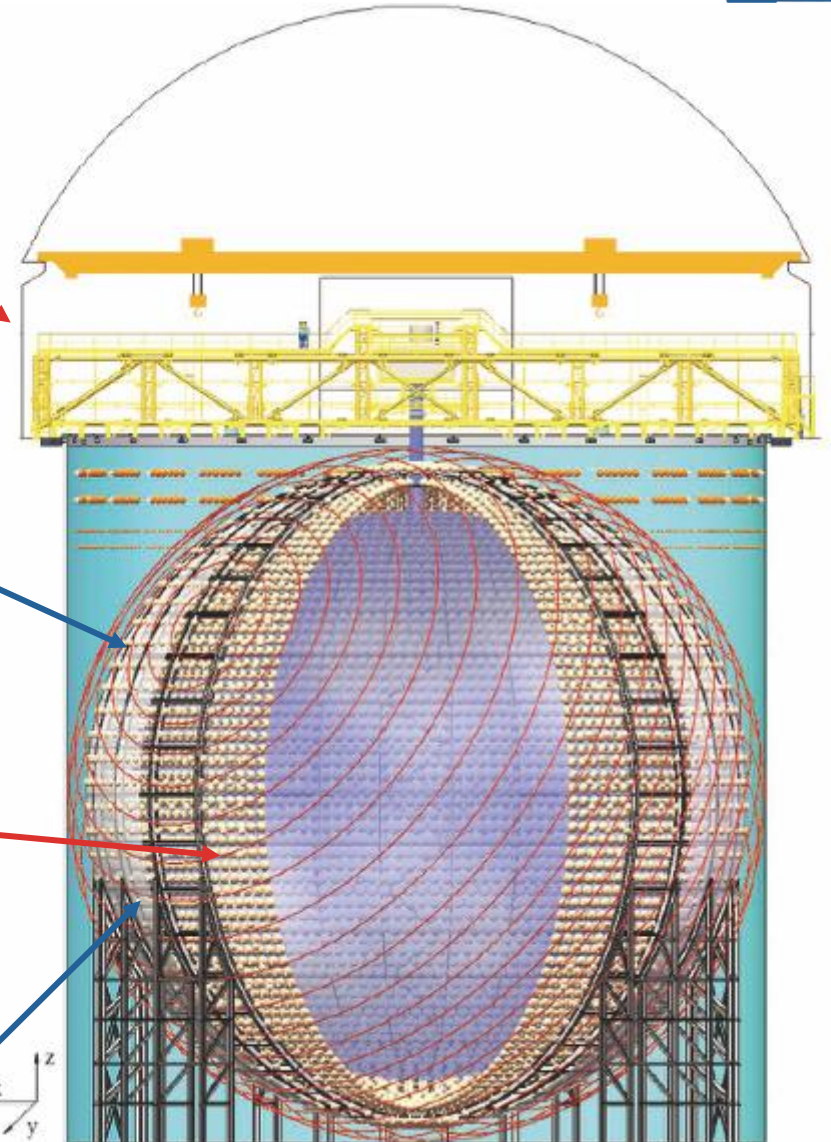
## PMTs

**17,596** 20-inch + **25,600** 3-inch



## Acrylic Vessel

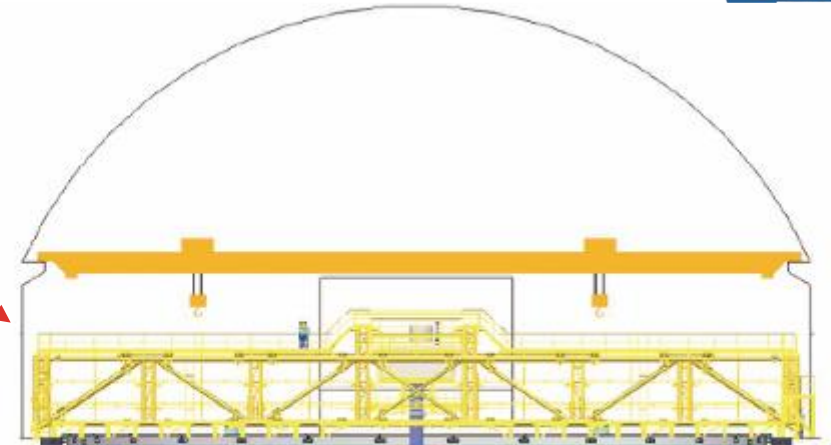
17.7m radius  
20 kton of liquid scintillator  
LAB + PPO + bis-MSB + BHT



# JUNO Detector

## Top Tracker

lastic scintillation  
rock overb



## Excellent Energy Resolution

78% Photocoverage  
~3% per MeV  
~1300 p.e. per MeV

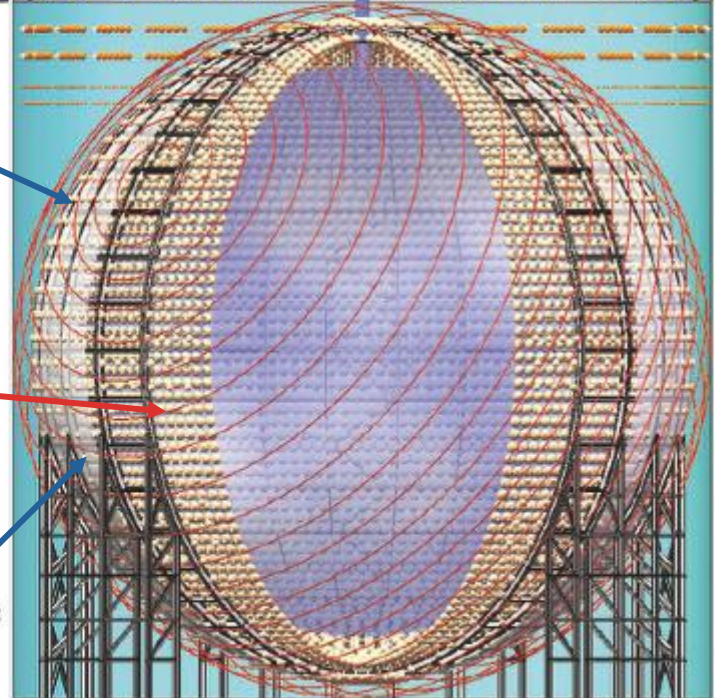


17,596 20-inc



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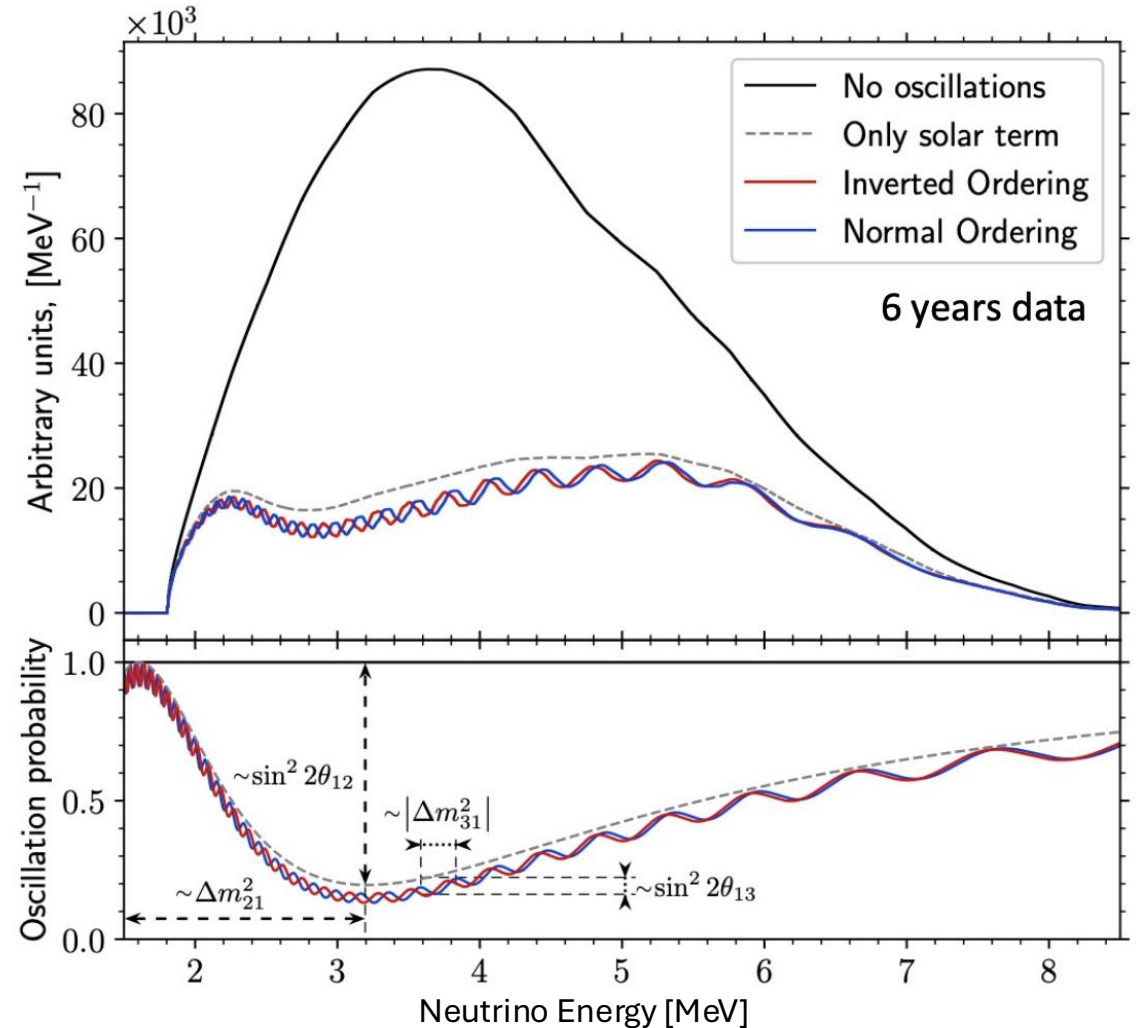


# Physics Programme

- NMO with  $3\sigma$  significance in **6 years**
- Precision of oscillation parameters
  - **<0.5%** for  $\sin^2 \theta_{12}$ ,  $\Delta m_{21}^2$  and  $\Delta m_{31}^2$  in **6 years**



Neutrino Source	Event Rate	Energy Region (MeV)
Reactor	60 / Day	0.1 - 10
Atmosphere	100s / Year	10 - 10 <sup>4</sup>
Supernova Burst	5000 at 10 kpc	0.1 - 10
Sun	2-4 / Year	0.1 - 10
Earth Crust/Mantle	400 / Year	0.1 - 10

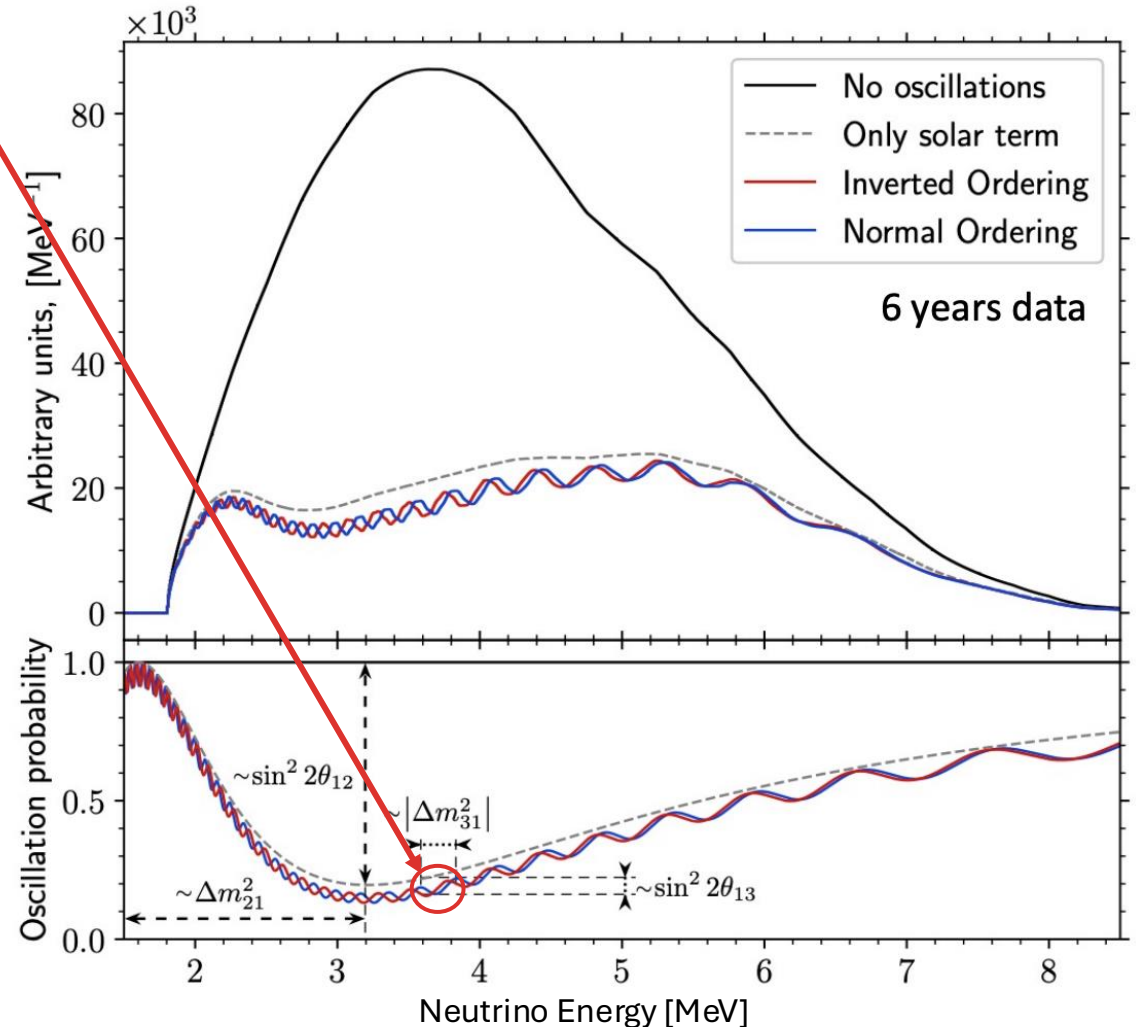


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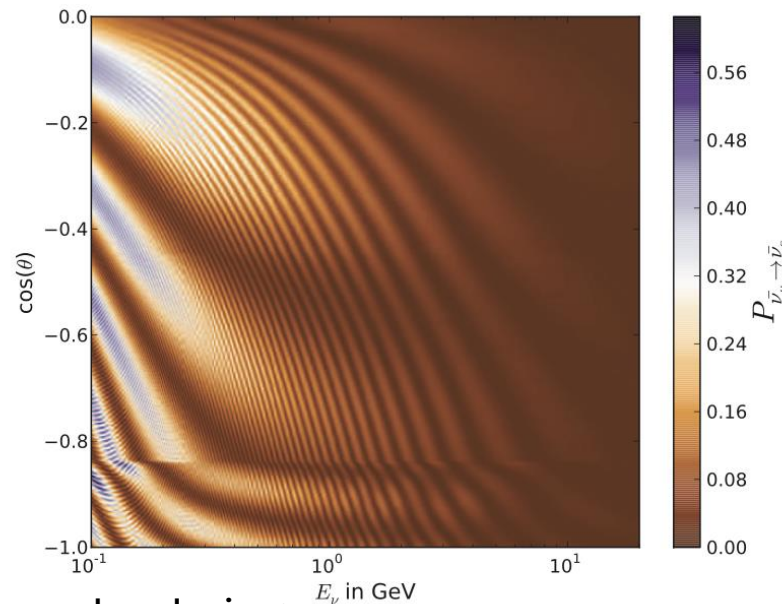
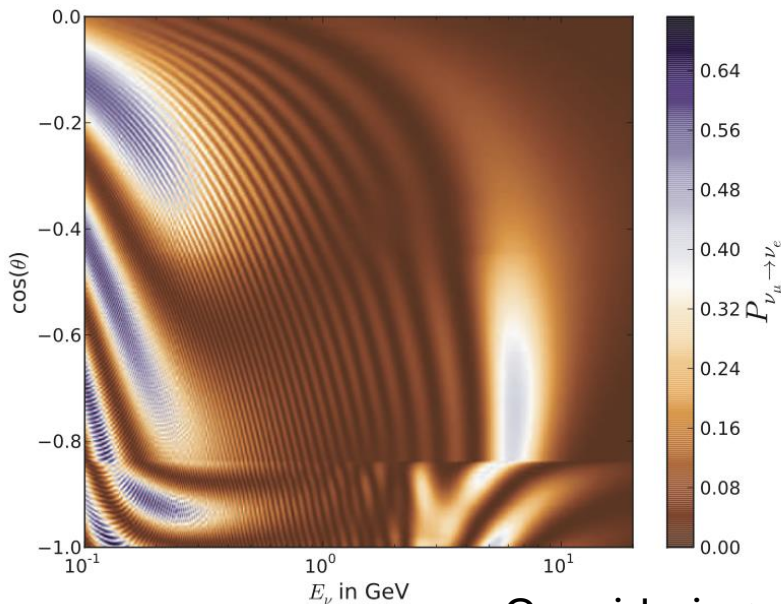
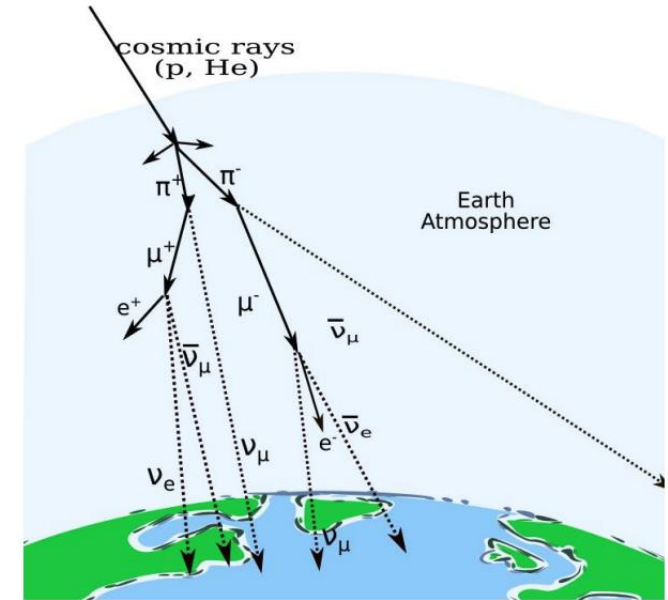


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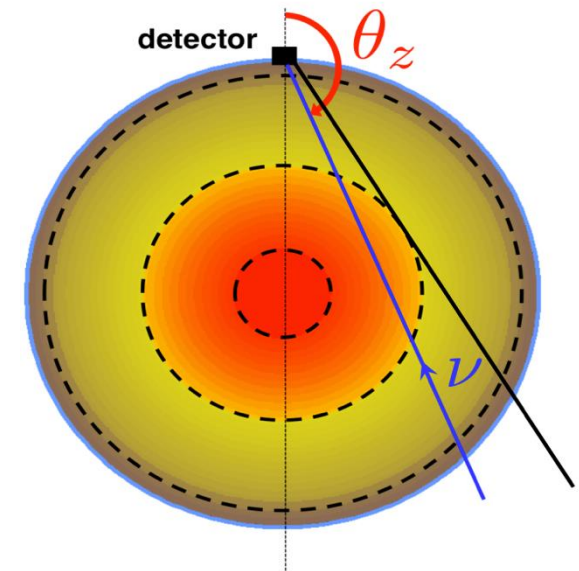


# Atmospheric Neutrinos

- Atmospheric neutrinos give insight to NMO through matter effects
- Offers a complementary channel to boost sensitivity
- Oscillation probability,  $P = f\left(\frac{L}{E}\right)$ ,  $L \sim \cos\theta \rightarrow$   
Neutrino **flavour**, **direction** and **energy** are crucial

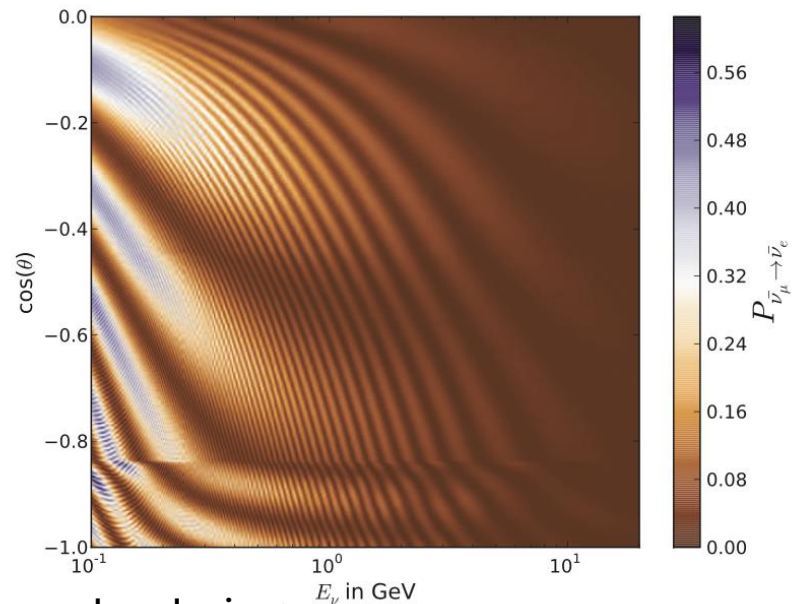
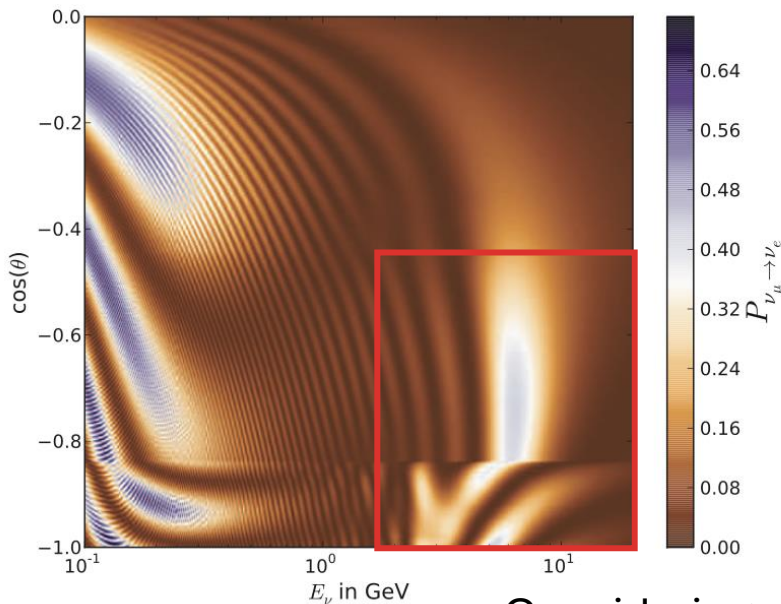
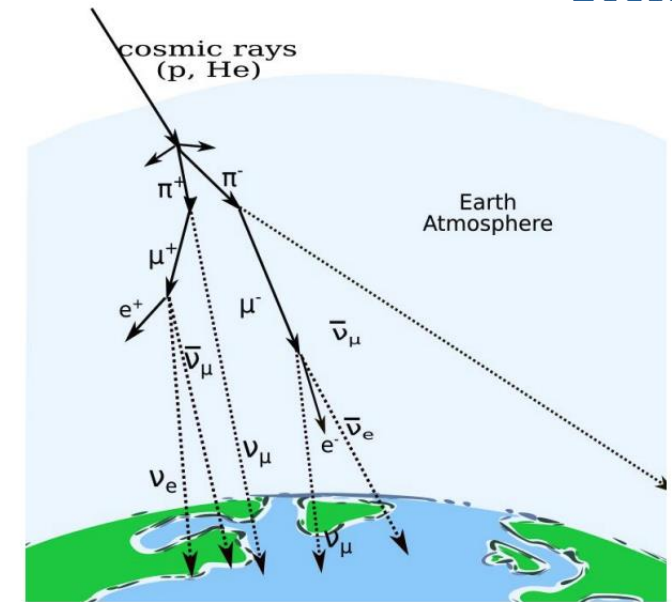


Considering normal ordering

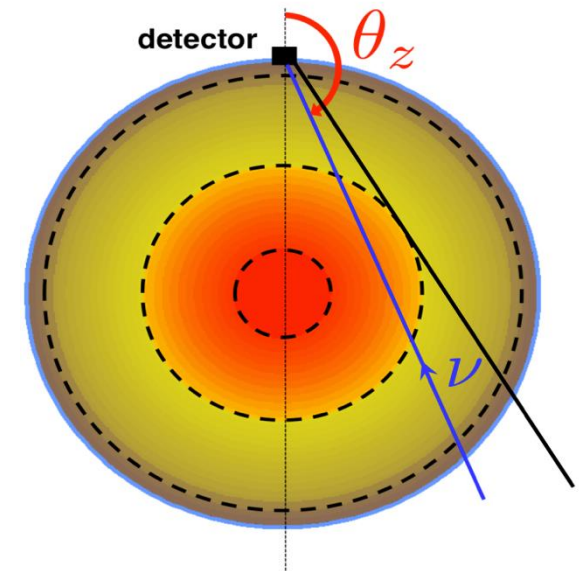


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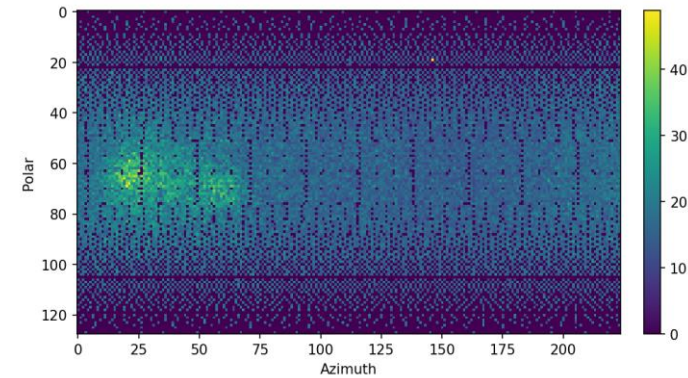
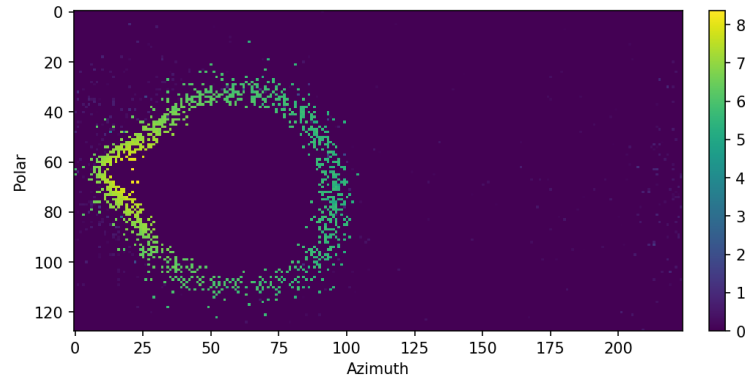
Considering normal ordering



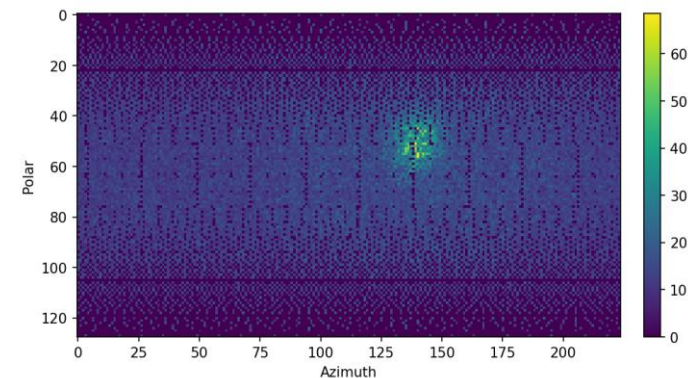
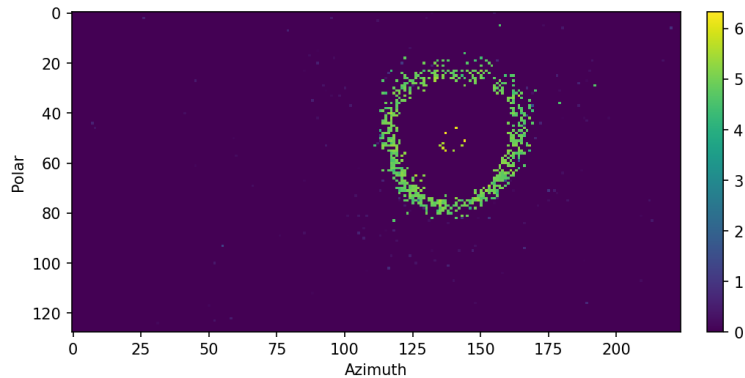
# Why Use a Spatiotemporal Method?

- Handcrafted features throw away information
- Using raw **time/charge** pairs allows the model to decide for itself
- Can capture the highly complex patterns in superpositions of light

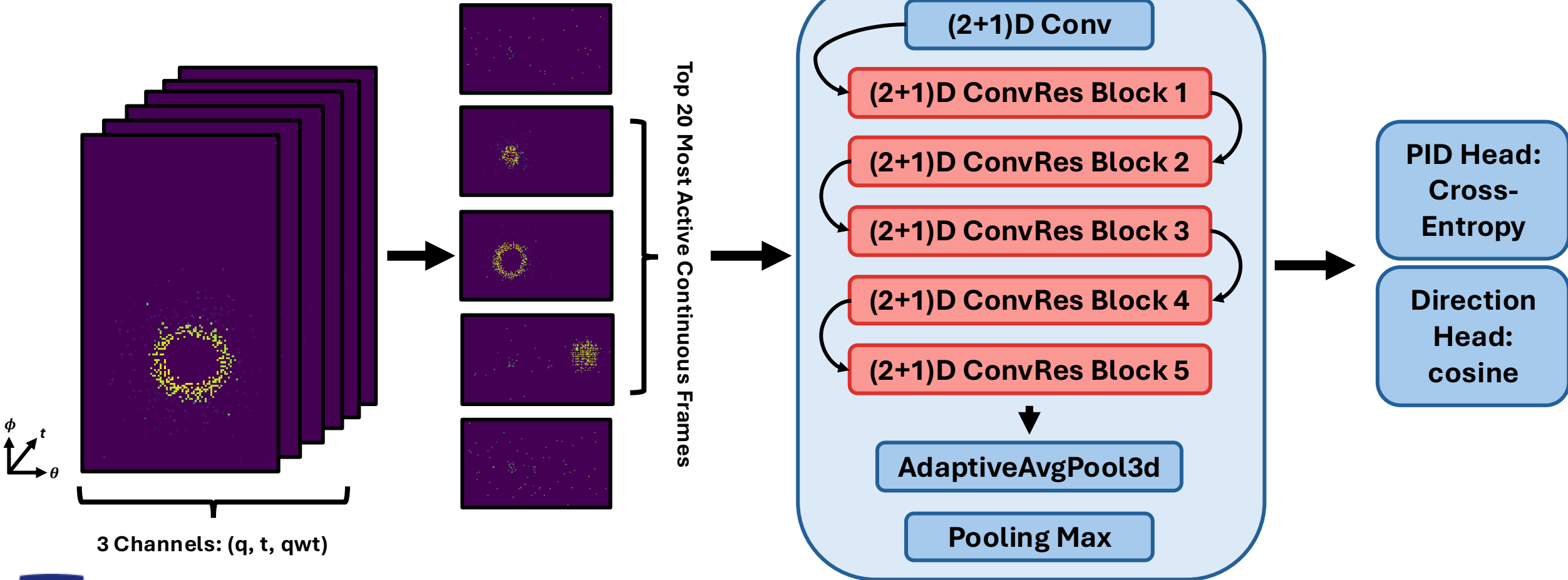
$\nu_{\mu}$  – like



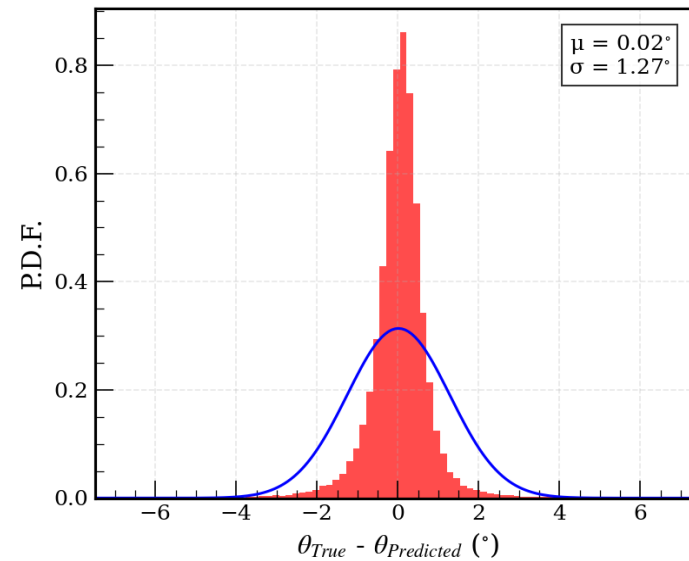
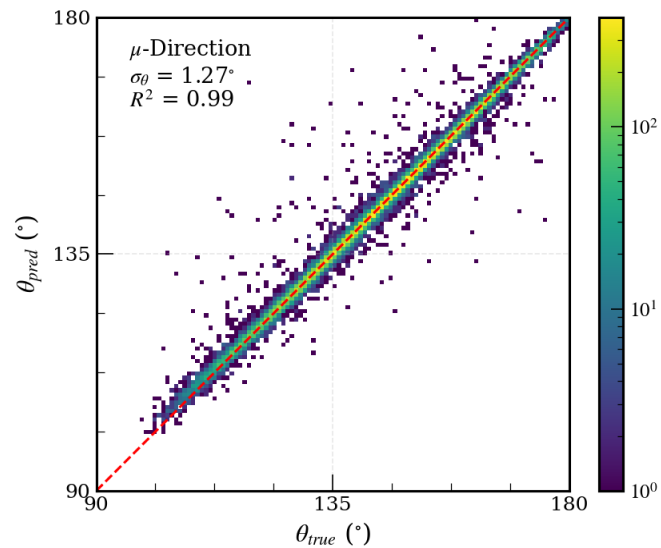
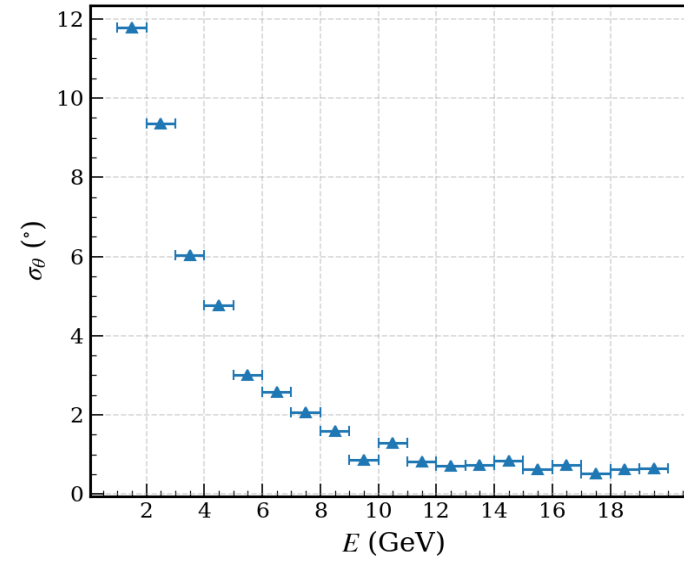
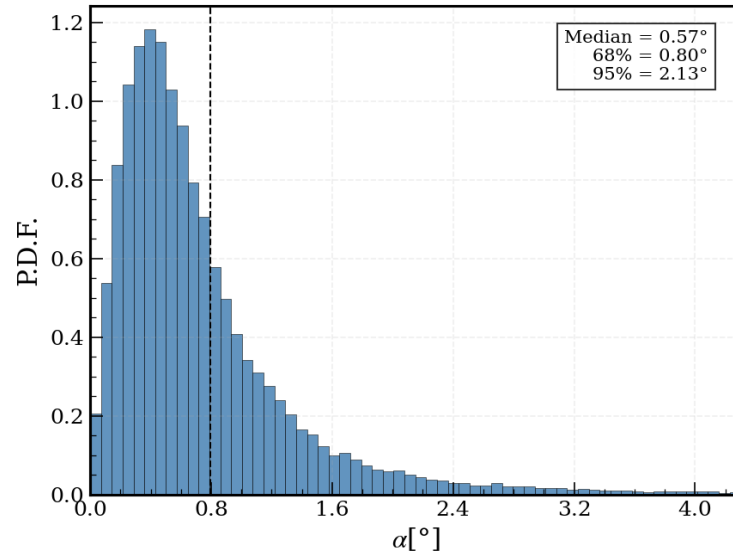
$\nu_e$  – like



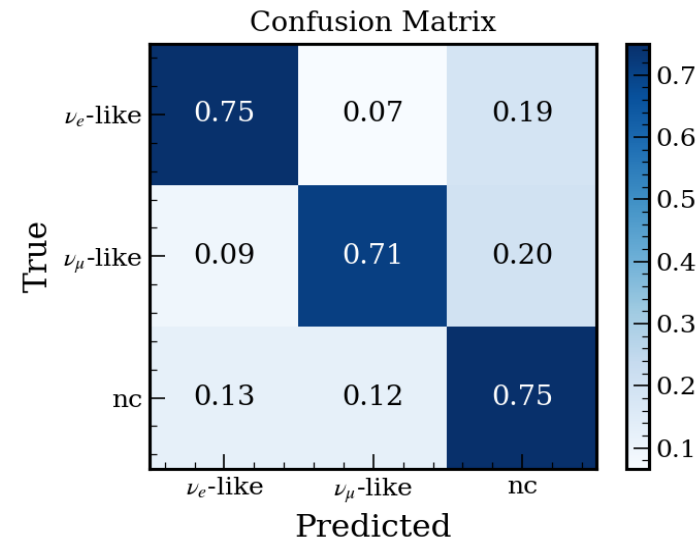
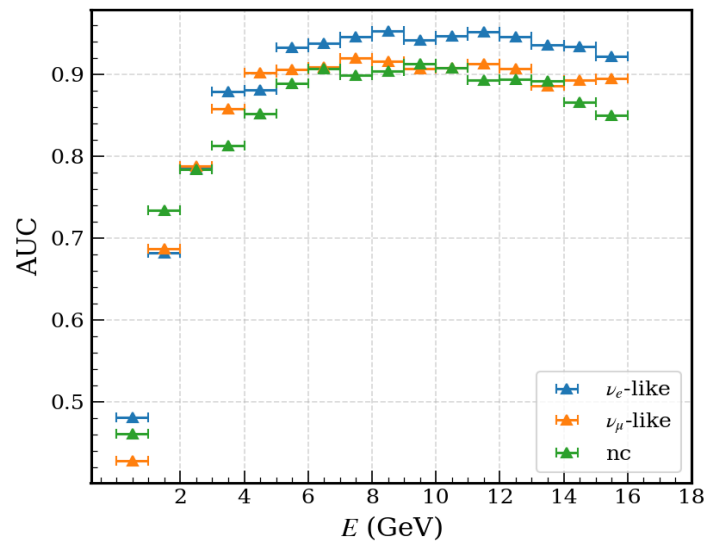
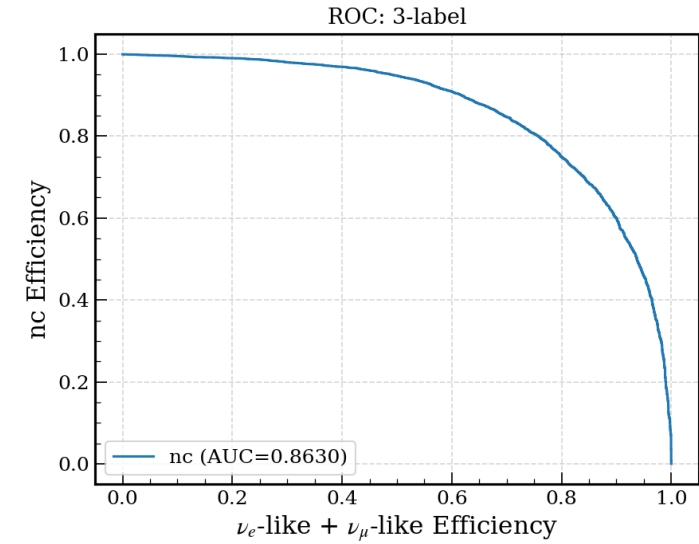
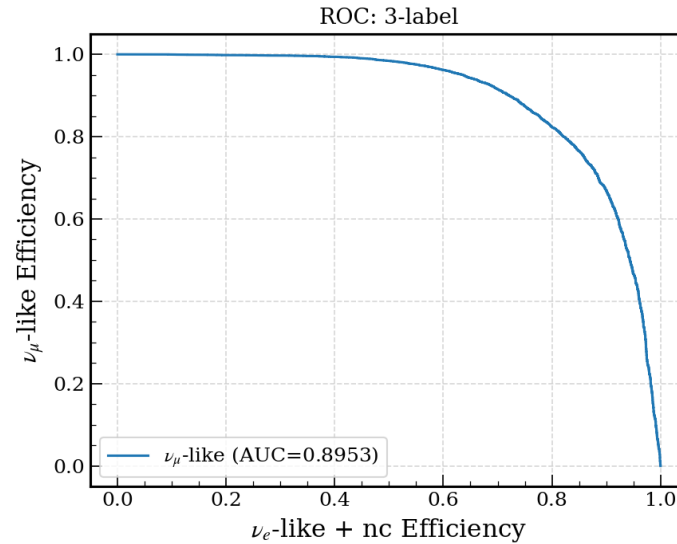
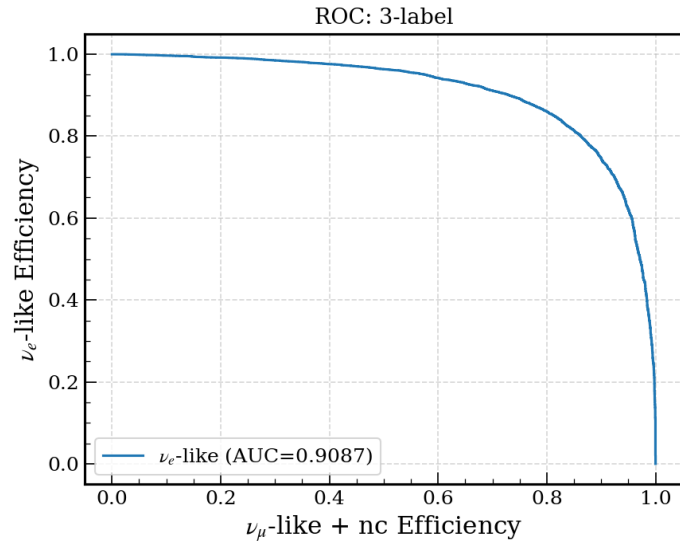
# (2+1)D Convolution for Reconstruction



# Muon Direction



# Atmospheric Flavour Classification



# Conclusion

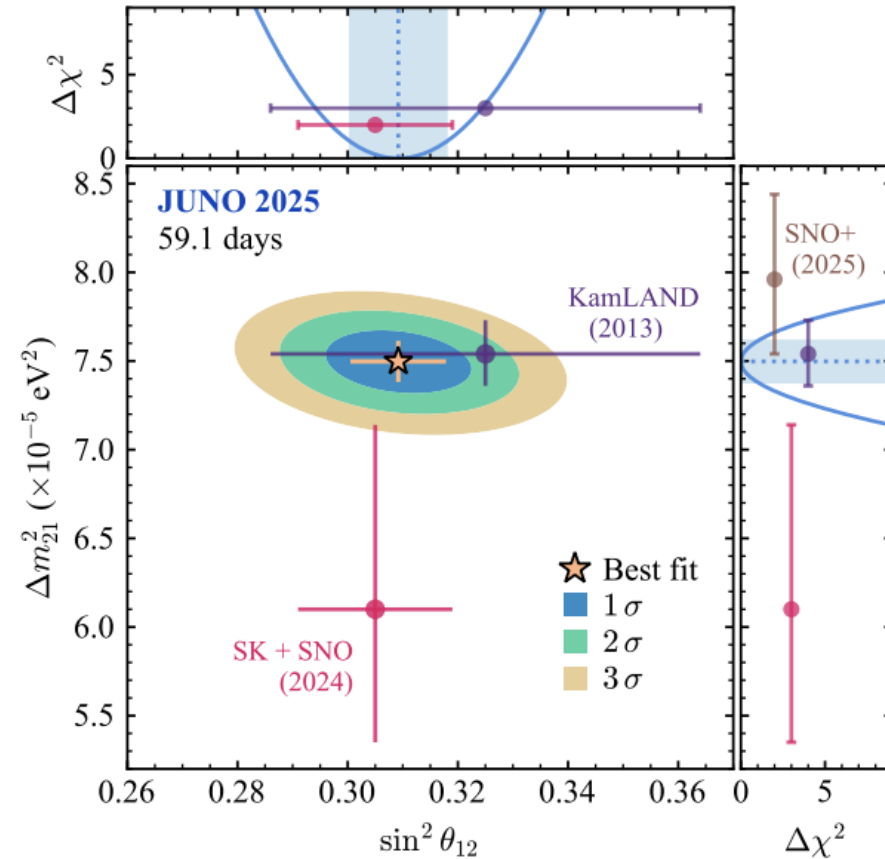
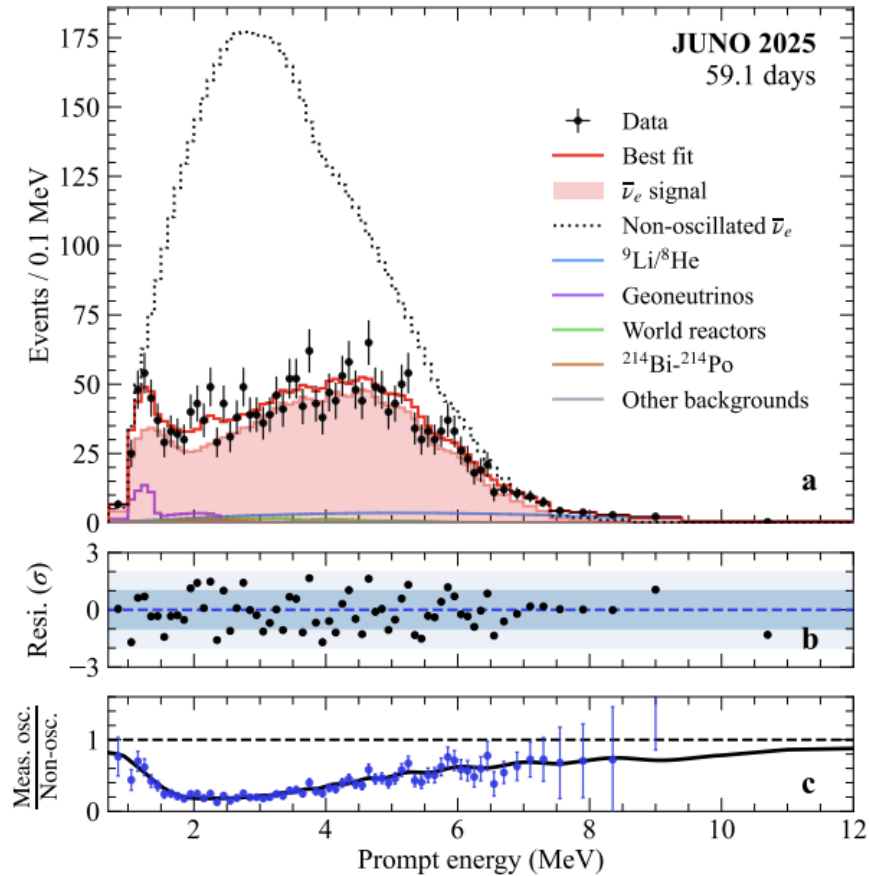
- Summary
  - The first attempt at spatiotemporal learning shows promise
  - For both case low energy events make results worse
  - NC events underperform
- Next Steps
  - Move to atmospheric neutrino direction reconstruction
  - Test auxiliary losses to improve performance
  - Look to real events
  - Extend to 4D spatiotemporal learning



# Backup



# JUNO First Results



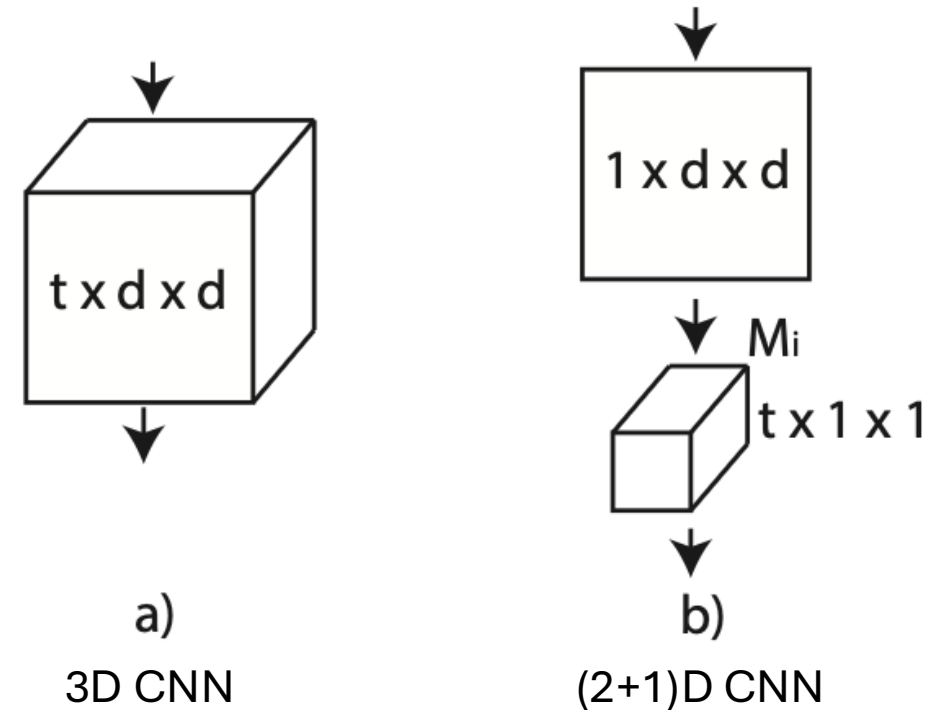
$$\sin^2 \theta_{12} = 0.3092 \pm 0.0087$$

$$\Delta m_{21}^2 = (7.50 \pm 0.12) \times 10^{-5} \text{ eV}^2$$

# SpatioTemporalConv

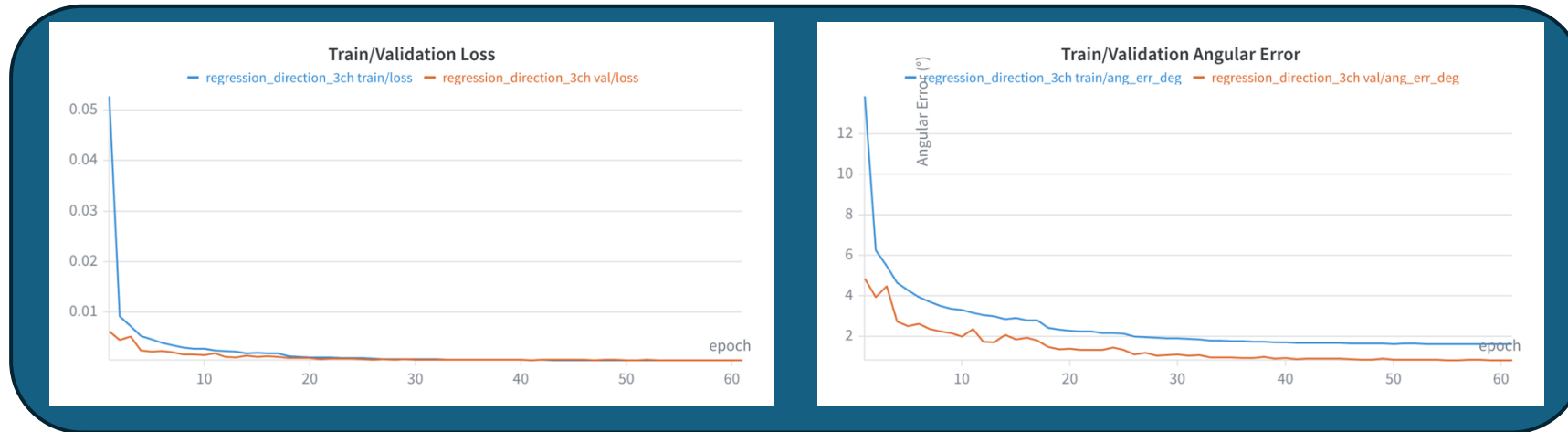
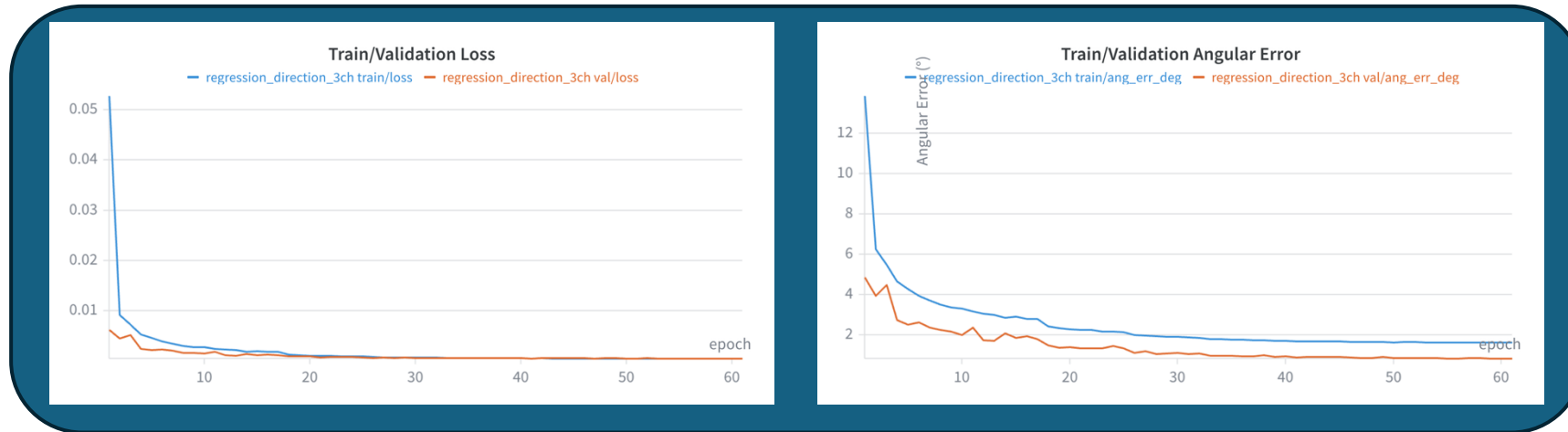
- Instead of a full 3DConv (kernel [kT, kH, kW]) we split the time and space components:
  - **SpatioConv:** Looks at 2D space bins (kernel [1, kH, kW])
  - **TemporalConv:** Looks at 1D space bins (kernel [kT, 1, 1])
- Intermediate channel count ( $M_i$ ) is chosen to match param count of 3DConv:

$$M_i = \frac{(kT \cdot kH \cdot kW \cdot C_{in} \cdot C_{out})}{(kH \cdot kW \cdot C_{in} + kT \cdot C_{out})}$$



# Training

## Muon Direction (40k/10k/40k)



## Atmospheric Classification (24k/6k/15k)