Hongyue Duyang **Shandong University** On behalf of the JUNO Collaboration

GeV Events

TIPP2023 @ Cape Town, South Africa



JUNO's Potential for



Introduction to JUNO

- Jiangmen Underground Neutrino 700 Observatory (JUNO):
 - Determine the neutrino mass ordering (NMO) with reactor neutrinos.
 - Measure neutrino oscillation parameters to sub-percent level
 - · Supernova, solar, geo., atm. v, etc.
- Currently under construction. Physics run Central Detector to start in 2024.





\$\$: 43.5m

The largest liquid scintillator detector ever built.









Why Capability in GeV is Needed?



- Cosmic muons produce backgrounds to reactor neutrino IBD events.

Precise track and shower vertex reconstruction is needed to reduce these backgrounds.

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- Cosmic muons produce backgrounds to reactor neutrino IBD
 - Precise track and shower vertex reconstruction is needed to
- Atmospheric neutrinos provide independent sensitivity to NM (directionality and flavor identification are mandatory).
- Other physics topics like indirect dark matter search.





A Liquid Scintillator Detector for GeV Events?

- LS detectors are traditionally good for low-energy topics: reactor/solar neutrinos etc.
 - Low threshold, high energy-resolution.
- But
 - No direct tracking information.
 - Scintillation light is isotropic, Cherenkov light is only a few percent: no direct directional information.

Typical LS detectors are designed with low-threshold, good energy resolution, ideal for low-energy physics.



Daya Bay







Kamland 3



Borexino





Scintillation light from a point source is isotropic

- PMTs at different angles wrt the track see distinct shapes of nPE(t)
- Exactly how nPE(t) looks depends on:
- Track direction;
- Track starting and stopping points;
- Track dE/dx...
- Event topology information in the PMT waveform.



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NSIDE: 16. Theta: 1.75, Phi: 0.033630







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5



PMT Waveforms (After deconvolution and noise-removing)



And more features....

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Pictures of PMT Features





And more features....

PMT Waveforms (After deconvolution and noise-removing)

Pictures of PMT Features Machine Learning Models (Planer: EfficientNetV2; Spherical: Deepsphere; 3D: PiontNet++)



PMT Waveforms (After deconvolution and noise-removing)

Pictures of PMT Features

flavor/vertex etc. from the feature patterns.

(Planer: EfficientNetV2; Spherical: Deepsphere; 3D: PiontNet++)

Models are trained with large number of PMT feature pictures and learn to find direction/energy/





Charged Lepton Direction Reconstruction



 Performances on simulated single charged leptons evaluated by the angle between true and reconstructed directions.

 ~1.5° angular resolution for electrons/muons at 1 GeV.



Cosmic Muon Track Reconstruction





• Cosmic muon tracks are reconstructed by the incident and exit points on the detector sphere.

 Performances evaluated by the angle between true and reconstructed tracks (α) and the distance between the track midpoints (d).







Atmospheric v: Directionality Reconstruction

- Both lepton and hadron informations are used in the directionality reconstruction.
- Low-threshold in LS detectors allows for more information from hadrons.
- the charged lepton direction.
- measurements.

• The reconstructed neutrino direction deviates less from true neutrino direction compared with

• An advantage for an LS detector with this method for atmospheric neutrino oscillation

- Two possible strategies on energy reconstruction:
- Strategy 1: Reconstruct the visible energy Strategy 2: Reconstruct the neutrino energy (after quenching in the LS).

ML result with PMT features and summed nPE(t) information

Detailed study is on-going on their impact on oscillation analysis.

*i***: Energy Reconstruction**

For fully-contained events only.

ML result with PMT features

Atmospheric v: Vertex Reconstruction

 Information of interaction vertex is useful for external background rejection and as input to other reconstruction algorithms.

 Performances evaluated by the distance between true and reconstructed vertices:

• Resolution: ~22 cm for ν_{μ} -CC and ~33 cm for ν_{e} -CC

• It is possible to statistically separate ν and $\bar{\nu}$ with neutron-capture informations.

Feature Extraction

For more details see Talk by Yongpeng Zhang: "The methodology of atmospheric neutrino identification in JUNO"

Feature Extraction

Strategy 2: Directly input PMT features from multiple triggers into ML.

Neutrino vs Anti-neutrino Performance

 Input features from both the prompt trigger and delayed triggers into ML. • ν and $\bar{\nu}$ can be statistically separated with the help from neutron-capture informations.

Neutrino Flavor Identification Performances

Summary and Outlook

- (MeV) topics.
- more.
- Stay tuned for more excited GeV physics from JUNO!

Liquid scintillator detectors are traditionally limited to low-energy

 With machine learning techniques we have greatly expanded their capabilities in GeV energy region: cosmic muons, atmospheric v and

• Performance comparable or even better than traditional large detectors for GeV physics (for example a water Cherenkov detector.)

Back up slides

Planer Model: EfficientNetV2

Spherical Model: DeepSphere

X4

3D Model: PointNet++

Alternative Generator Check

