Vertex Reconstruction in JUNO



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Zi		LI –	(SYSU	
	,		10.00	

Vertex Reconstruction

Jiangmen Underground Neutrino Observatory



- Location : Kaiping, Jiangmen city, Guangdong Province, China.
- Baseline optimized for Neutrino Mass Ordering determination : 53 km from Taishan and Yangjiang NPP
- Reactor $\bar{\nu_e}$: \sim 60 / day
- Data taking start : 2021

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The JUNO Detector



Top Tracker

- ▷ 3 layers of plastic scintillator
- \triangleright Precise μ tracking

Water Cherenkov Detector

- > 25 kton ultra-pure water
- 2.4k 20" PMTs
- \triangleright High μ detection efficiency
- Protects CD from external radioactivity

Central Detector

- > 35.4 m diameter acrylic sphere
- > 20 kton liquid scintillator
- ▷ 18k 20" PMTs + 25k 3" PMTs, 78% PMT coverage

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Physics of JUNO

Detection Principle





Physics

- Mass Hierarchy (MH)
- Precise measurement of parameter $(\theta_{12}, \Delta m_{21}^2, \Delta m_{31}^2)$
- Reduce uncertainty on δ_{CP}
- Understand requirement for $0\nu\beta\beta$ experiment





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Motivation of Vertex Reconstruction Study

To determine MH, require energy resolution better than $3\%/\sqrt{E}$



Base on the current experiment setup, could reach 3 - 4σ sensitivity in 6 years

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Vertex Reconstruction

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Motivation of Vertex Reconstruction Study



To achieve 3% energy resolution, Vertex resolution better than 12cm is needed



Illustration of Vertex



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- ▷ Charge Center
- Time Likelihood

▷ Deep Learning

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Charge Center Algorithm Principle

Use charge information as weight for each PMT, the vertex position can be estimated by $\vec{r_{rec}}$:

 $\vec{r}_{rec} = \alpha \sum_i q_i \vec{r}_i / \sum_i q_i$

where :

 q_i : i PMT collected charge;

 $\vec{r_i}$: i PMT position;

 α : 1.2 correction factor

Mathematic calculation :

$$z_{rec} = \frac{1}{4\pi} \int zd\Omega$$

= $\frac{1}{4\pi} \int_{0}^{2\pi} (z_0 + r \cdot \cos\theta) \sin\theta d\theta$
= $\frac{1}{2} \int_{0}^{2\pi} (z_0 + (\sqrt{R^2 - z_0^2 \sin^2\theta} - z_0 \cos\theta) \cdot \cos\theta) \sin\theta d\theta$
= $\frac{2}{3} z_0$



- Light source deployed at z_0
- Gamma evenly emitted in 4π solid angle of z_0
- Charge center method reconstructs vertex at 2/3 z₀

Charge Center Performance

For $R_{edep} > 16$ m, due to total refraction, the charge distribution change, charge center method doesn't work well



- Charge Center
- Time Likelihood

Deep Learning

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Time Likelihood Algorithm Principle

For point-like events like e^+ , e^- , γ deposit energy in liquid scintillator and emit photons, define residual hit time :

$$t_{i,res} = t_i - tof - t_0$$

where :

- t_i : First hit time of i PMT;
- tof : Time of flight for scintillating photon;
- t_0 : Real time of an event.



Time Likelihood Algorithm Principle

Charge Center Method to get initial vertex

For each iteration step with vertex position $\vec{R_0}$, calculate $t_{i,res}$ for each PMT

$$t_{i,res}(ec{R_0},t_0)=t_i-\sum_lpharac{D_lpha(ec{R_0},ec{R_i})}{c_lpha}-t_0$$

where :

 α : Different material;

 c_{α} : Light speed in material;

 D_{α} : Photon travel length in material;

 R_i : Position of i PMT.

Define Joint Likelihood Function

$$\mathcal{L}(ec{R_0},t_0)=-\sum_i \ln(f(t_{i,res}))$$

Minimize likelihood function to get the best vertex position

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Possibility Density Function

The quality of vertex reconstruction depends on how well our PDF could describe the physical processes

- Put 4.4MeV γ source at the center, get the t_{res} distribution from simulation for PMT with nPE=1
- In case nPE > 1, use the following equation to calculate the PDF



GridSearch Minimization

Step :

- * Get initial vertex position from Charge Center Method
- @ For each direction (X,Y,Z) and iteration, ± 1 step to find the minimum grid in the current step_length
- & step_length divided by 2 and go to 0, stop iterate until step_length < 0.1 mm



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Vertex Reconstruction Bias

Apparent correlations between z_{rec} - z_{edep} mean bias and z_{edep}



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Effective Light Speed

Time of Flight is not correct : $t_{res} = t_i - tof - t_0$

- The actual path length that a photon traveled is longer than the length of straight line between hit and vertex
- \Box Using straight line and c_{LS} $(\frac{c}{n_{LS}})$ makes *tof* smaller than actual, $t_{res} > 0$

□ Both near and far ends want to recover its correct path, pushing vertex recostruct to the other end



Effective Light Speed

Time of Flight is not correct : $t_{res} = t_i - tof - t_0$

- □ To correct *tof*, should use true path length (unknown)
- \Box Alternative is to introduce the effective speed of light c_{eff} $\left(\frac{c}{n_{eff}}\right)$

 \Box tof = $\frac{distance}{c_{eff}}$, an appropriate c_{eff} can correct tof back



Effective Light Speed

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 \Box tof = $\frac{distance}{c_{eff}}$, an appropriate c_{eff} can correct tof back

□ *c*_{eff} is a parameter to simplify the entire optical model, including both geometrical and physical effect



Time Likelihood Performance

Reconstruction in each direction



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Time Likelihood Performance

Resolution vs. Energy



Time Likelihood Performance

Vertex Reconstruction Bias



More factors to consider

In real experiment, the following effects will worsen resolution

- PMT Time Transit Spread (TTS)
- > Waveform Reconstruction
- Dark Noise



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Time Likelihood Performance with TTS



Charge Center

Time Likelihood

Deep Learning

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Deep Learning



Deep Learning Performance



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Summary

Conclusion

- Three methods have been developed for vertex reconstruction in JUNO : Charge Center, Time Likelihood, Deep Learning
- ▷ Vertex Resolution is 60mm @ 1MeV, and 90mm @ 1MeV with TTS
- Vertex mean bias is less than 40mm

Outlook

- Further investigate the physic behind vertex bias
- Reconstruct with Dark Noise
- Deep Learning with TTS and Dark Noise

The End



Thanks for your attention !

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