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**Sensitivity to the neutrino mass hierarchy
with the JUNO experiment**

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1 Introduction

The Standard Model of particle physics [1] is a successful theory which not only unifies the electromagnetic and weak interactions but also explains almost all the phenomena of this nature observed at or below the electroweak scale. When this theory was first formulated by Weinberg in 1967 [2], its particle content was so economical that the neutrinos were assumed to be massless and hence there was no lepton flavor mixing. But just one year later solar neutrinos were observed by Davis *et al.* [3], and a deficit of their flux as compared with the prediction from the Standard Solar Model was also established by Bahcall *et al.* [4]. Such an anomaly turned out to be solid evidence for new physics beyond the Standard Model, because it was found to be attributed to neutrino oscillations, a spontaneous and periodic change from one neutrino flavor to another, which does not take place unless neutrinos have finite masses. By investigating these oscillations only the absolute value of the differences between the mass-square eigenstates can be measured, and thus we don't know which one is heavier — actually nowadays we know that $m_{\nu_2} > m_{\nu_1}$ thanks to matter effects in neutrino solar oscillations. One of the main aims of neutrino physics today is to determinate the correct mass hierarchy of the neutrino mass eigenstates.

1.1 The neutrino mass hierarchy

It is a well-established experimental fact that neutrinos and antineutrinos which take part in charged current (CC) and neutral current (NC) weak interaction are of three varieties (types) or flavours: electron, ν_e and $\bar{\nu}_e$, muon, ν_μ and $\bar{\nu}_\mu$, and tauon, ν_τ and $\bar{\nu}_\tau$. The flavour is conserved: so ν_e is the neutrino which is produced with e^+ , or produces an e^- in CC weak interaction processes; ν_μ is the neutrino which is produced with μ^+ , or produces μ^- , etc.

The experiments with solar, atmospheric, reactor and accelerator neutrinos have provided compelling evidences for the existence of neutrino oscillations, transitions in flight between the different flavour neutrinos ν_e, ν_μ, ν_τ (antineutrinos $\bar{\nu}_e, \bar{\nu}_\mu, \bar{\nu}_\tau$), caused by nonzero neutrino masses and neutrino mixing.

The existence of flavour neutrino oscillations implies that if a neutrino of a given flavour, say ν_α , with energy E is produced in some weak interaction process, at a sufficiently large distance L from the ν_α source, the probability to find a neutrino of a different flavour, say ν_β , $P(\nu_\alpha \rightarrow \nu_\beta; E, L)$, can be different from zero. $P(\nu_\alpha \rightarrow \nu_\beta; E, L)$ is called the $\nu_\alpha \rightarrow \nu_\beta$ oscillation or transition probability. If $P(\nu_\alpha \rightarrow \nu_\beta; E, L) \neq 0$, the probability that ν_μ will not change into a neutrino of a different flavour, i.e., the “ ν_μ survival probability” $P(\nu_\alpha \rightarrow \nu_\alpha; E, L)$ will be smaller than one. If only muon neutrinos ν_μ are detected in a given experiment and they take part in oscillations, one would observe a “disappearance” of muon neutrinos on the way from the ν_α source to the detector.

Oscillations of neutrinos are a consequence of the presence of flavour neutrino mixing, or lepton mixing, in vacuum. This means that the flavour eigenstate of the neutrino $\nu_l, |\nu_l\rangle$, will be a coherent superposition of the mass eigenstates $|\nu_j\rangle$ with

the physical mass m_j :

$$|\nu_l\rangle = \sum_{j=1}^3 U_{lj} |\nu_j\rangle, \quad l = e, \mu, \tau,$$

where U is the 3×3 Maki-Nakagawa-Sakata-Pontecorvo (MNSP) matrix [5, 6]. The unitarity of U depends on the mechanism of neutrino mass generation [7], however the bottom line is that any possible deviation of U from unitarity must be small, at most at the percent level, as constrained by the available experimental data [8, 9]. That is why U is simply assumed to be unitary in dealing with current neutrino oscillation data. The same formulas can be applied to describe the antineutrino mixing by making the replacement $U \rightarrow U^*$ (complex conjugate).

If the 3×3 MNSP matrix U is exactly unitary, it can be parametrized in terms of three flavor mixing angles and three CP-violating phases in the following ‘standard’ way:

$$U = \begin{pmatrix} 1 & 0 & 0 \\ 0 & c_{23} & s_{23} \\ 0 & -s_{23} & c_{23} \end{pmatrix} \begin{pmatrix} c_{13} & 0 & s_{13}e^{-i\delta} \\ 0 & 1 & 0 \\ s_{13}e^{-i\delta} & 0 & c_{13} \end{pmatrix} \begin{pmatrix} c_{12} & s_{12} & 0 \\ -s_{12} & c_{12} & 0 \\ 0 & 0 & 1 \end{pmatrix} P_\nu,$$

where $c_{ij} := \cos \vartheta_{ij}$ and $s_{ij} := \sin \vartheta_{ij}$ are defined, and $P_\nu := \text{Diag}\{e^{i\varphi}, e^{i\sigma}, 1\}$ denotes the diagonal Majorana phase matrix which is not sensitive to neutrino oscillations.

Therefore we define the survival probability $P(\bar{\nu}_e \rightarrow \bar{\nu}_e; E, L)$ as the square of the oscillation amplitude:

$$\begin{aligned} P(\bar{\nu}_e \rightarrow \bar{\nu}_e; E, L) &:= |A(\bar{\nu}_e \rightarrow \bar{\nu}_e)|^2 \\ &= 1 - 4 \left(\sum_i |U_{ei}|^2 \right)^{-2} \sum_{i < j} \left(|U_{ei}|^2 |U_{ej}|^2 \sin^2 \frac{\Delta m_{ij}^2 L}{4E} \right), \end{aligned} \quad (1.1)$$

where $\Delta m_{ij}^2 = m_i^2 - m_j^2$ (eV^2), L (km) is the distance from the antineutrino production site and E (MeV) the antineutrino energy. Thus we can characterize the antineutrino survival probability with four parameters (in addition to the baseline length L and the antineutrino energy E): two mixing angles, $\sin^2 2\vartheta_{12}$ and $\sin^2 2\vartheta_{13}$, and two mass-square differences Δm_{31}^2 and Δm_{21}^2 . Past experiments [11] have measured all these parameters, but not the sign of Δm_{31}^2 , of which we only know the absolute value. This leads to the two hypothesis summarized in figure 1, to which we will from now on refer to as ‘normal’ or ‘inverted’ hierarchy.

The JUNO experiment aims to measure the flux rate and energy spectrum of $\bar{\nu}_e \rightarrow \bar{\nu}_e$ oscillations to an unprecedentedly good degree of accuracy, especially to pin down the sign of Δm_{31}^2 or equivalently the neutrino mass ordering.

1.2 The JUNO experiment

The Jiangmen Underground Neutrino Observatory (JUNO) is a multi-purpose neutrino experiment [10]. It was proposed in 2008 to determine the neutrino mass hierarchy by detecting reactor antineutrinos from the Daya Bay nuclear power plant (NPP), thus formerly known as ‘Daya Bay II experiment’. The mass hierarchy determination requires equal baselines from the detector to all reactor cores to avoid

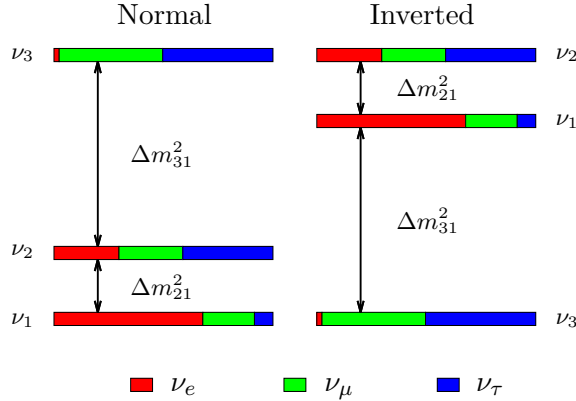


Figure 1: Neutrino mass hierarchy hypothesis.

cancellation of the oscillation dephasing effect. Due to the complex and unclear layout of the future nuclear power plants in the neighborhood, the experiment was moved to Jiangmen city in Guangdong province in August 2012, and named as JUNO in 2013. The site location is at 53 km from both the Yangjiang and Taishan NPPs. The JUNO project was approved by Chinese Academy of Sciences in February 2013. Data taking is expected in 2020.

JUNO consists of a central detector, a water Cherenkov pool and a muon tracker. The central detector is a Liquid Scintillator (LS) detector of 20 kton fiducial mass with a designed energy resolution of $3\%/\sqrt{E/\text{MeV}}$. Both central detector (CD) and water pool (WP) are equipped with 20" Photomultiplier Tubes (PMTs); they detect scintillation light in CD and Cherenkov light in WP.

JUNO measures the reactor neutrino signal via the inverse beta decay reaction

$$\bar{\nu}_e + p \rightarrow e^+ + n$$

in the LS. The reactor antineutrino $\bar{\nu}_e$ interacts with a proton, creating a positron (e^+) and a neutron. The positron quickly deposits its energy and annihilates into two 511-keV γ -rays, which gives a prompt signal. The neutron scatters in the detector until being thermalized. It is then captured by a proton, on average $200\mu\text{s}$ later, and releases a 2.2-MeV γ -ray. The coincidence of the prompt-delayed signal pair in such a short time significantly reduces backgrounds. The positron carries almost all energy of the neutrino in this reaction. All this may be visually schematized as in figure 2. The WP and an additional Top Tracker are used to measure muon tracks to reject cosmogenic backgrounds. Further details on JUNO physics can be found in [10].

2 Reactor antineutrino flux

In this section, it will be briefly discussed the evaluation of how many electron antineutrinos $\bar{\nu}_e$ would be detected at a far detector with medium baseline length from a reactor.

In a nuclear reactor, antineutrinos are mainly produced via beta decay of the fission products of the four radio-active isotopes ^{235}U , ^{238}U , ^{239}Pu and ^{241}Pu in the

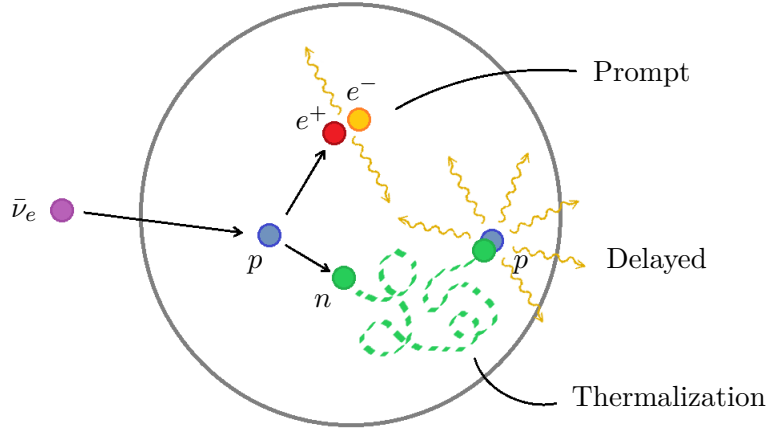


Figure 2: Antineutrino detection in JUNO via inverse beta decay: the prompt signal is first detected, the delayed one comes, on average, after $200\mu\text{s}$.

fuel. The number of antineutrinos produced per fission depends on their energy $E_{\bar{\nu}_e}$ (MeV) [14]

$$\begin{aligned} \varphi(E_{\bar{\nu}_e}) = & f_{235\text{U}} \exp(0.870 - 0.160E_{\bar{\nu}_e} - 0.0910E_{\bar{\nu}_e}^2) \\ & + f_{239\text{Pu}} \exp(0.896 - 0.239E_{\bar{\nu}_e} - 0.0981E_{\bar{\nu}_e}^2) \\ & + f_{238\text{U}} \exp(0.976 - 0.162E_{\bar{\nu}_e} - 0.0790E_{\bar{\nu}_e}^2) \\ & + f_{241\text{Pu}} \exp(0.793 - 0.080E_{\bar{\nu}_e} - 0.1085E_{\bar{\nu}_e}^2), \end{aligned}$$

where f_k denotes the relative fission contribution of the isotope k in a reactor fuel, derived from the fission rate N_k^{fiss} of isotope k as

$$f_k := \frac{N_k^{\text{fiss}}}{\sum_i N_i^{\text{fiss}}}.$$

Although f_k varies over time as the fuel is burned, it can be approximated for this type of experiments with the average value of the relative fission contributions: $f_{235\text{U}} = 0.58$, $f_{239\text{Pu}} = 0.30$, $f_{238\text{U}} = 0.07$ and $f_{241\text{Pu}} = 0.05$ [15]. The event rate of antineutrinos with energy $E_{\bar{\nu}_e}$ (MeV) at a reactor of P (GW_{th}) thermal power is then expressed as

$$\frac{dN}{dE_{\bar{\nu}_e}} = \frac{P}{\sum_k f_k \varepsilon_k} \varphi(E_{\bar{\nu}_e}) \cdot 6.24 \cdot 10^{21}, \quad (2.1)$$

where ε_k is the released energy per fission of the isotope k : $\varepsilon_{235\text{U}} = 201.7$ MeV, $\varepsilon_{239\text{Pu}} = 210.0$ MeV, $\varepsilon_{238\text{U}} = 205.0$ MeV and $\varepsilon_{241\text{Pu}} = 212.4$ MeV [16]. The numerical factor comes from unit conversion, $1\text{GW}/\text{MeV} = 6.24 \cdot 10^{21}$.

This rate is then modulated by oscillation. The $\bar{\nu}_e$ survival probability (1.1) is expressed as

$$\begin{aligned} P_{\bar{\nu}_e \rightarrow \bar{\nu}_e} \equiv P_{ee} = & 1 - \cos^4 \vartheta_{13} \sin^2 2\vartheta_{12} \sin^2 \Delta_{21} \\ & - \cos^2 \vartheta_{12} \sin^2 2\vartheta_{13} \sin^2 \Delta_{31} \\ & - \sin^2 \vartheta_{12} \sin^2 2\vartheta_{13} \sin^2 \Delta_{32}, \end{aligned} \quad (2.2)$$

The variables m_i and E_i are the mass and energy of the corresponding mass eigenstate, while ϑ_{ij} represent the neutrino mixing angles. The oscillation phases Δ_{ij} are defined as

$$\Delta_{ij} := \frac{\Delta m_{ij}^2 L}{4E_{\bar{\nu}_e}} = 1.27 \cdot \frac{[\text{eV}^2][m]}{[\text{MeV}]}, \quad (\Delta m_{ij}^2 := m_i^2 - m_j^2)$$

with a baseline length from the reactor L .

While Δm_{21}^2 is fixed, Δm_{31}^2 and then $\Delta m_{32}^2 = \Delta m_{31}^2 - \Delta m_{21}^2$ depend on the mass hierarchy:

$$\Delta m_{31}^2 = \begin{cases} m_3^2 - m_1^2 > 0 & \text{(NH)} \\ m_3^2 - m_1^2 < 0 & \text{(IH)} \end{cases}.$$

As we have seen before JUNO uses protons as targets to detect electron antineutrinos via the inverse neutron beta decay (IBD) process

$$\bar{\nu}_e + p \rightarrow e^+ + n.$$

The threshold neutrino energy of this process is $E_{\text{thr}} \sim m_n - m_p + m_e \sim 1.804$ MeV, and the cross section is [17]

$$\sigma_{\text{IBD}} = E_{e^+} p_{e^+} \cdot 9.52 \cdot 10^{-44} \text{ cm}^2, \quad (2.3)$$

where E_{e^+} and p_{e^+} are the energy and momentum of the positron in MeV, neglecting the kinetic energy of the proton and the neutron for a MeV scale antineutrino. The positron energy is roughly $E_{e^+} \sim E_{\bar{\nu}_e} - (m_n - m_p)$.

The produced positron then interacts with scintillator, converting its kinetic energy to photons. Eventually, the positron annihilates with an electron in the detector and emits two 0.5 MeV photons. The energies of those photons are the accumulated as the visible energy E_{vis} which is the sum of the positron's total and one electron's rest energies,

$$E_{\text{vis}} \sim E_{e^+} + m_e \sim (E_{\bar{\nu}_e} - 0.78) \text{ MeV}.$$

As we can see the antineutrino spectrum is simply shifted of 0.78 MeV circa from the visible spectrum.

Finite detector energy resolution may distort the true visible energy E_{vis} to the finally observed one $E_{\text{vis}}^{\text{obs}}$. This effect can be modeled by a gaussian detector response function $G(E_{\text{vis}} - E_{\text{vis}}^{\text{obs}}, \delta E_{\text{vis}})$ with the energy resolution δE_{vis}

$$G(E_{\text{vis}} - E_{\text{vis}}^{\text{obs}}, \delta E_{\text{vis}}) = \frac{1}{\sqrt{2\pi}\delta E_{\text{vis}}} \exp \left\{ -\frac{(E_{\text{vis}} - E_{\text{vis}}^{\text{obs}})^2}{2(\delta E_{\text{vis}})^2} \right\}. \quad (2.4)$$

Generally the detector energy resolution can be modeled as

$$\frac{\delta E_{\text{vis}}}{E_{\text{vis}}} = \sqrt{\frac{a^2}{E_{\text{vis}}} + b^2} \quad (2.5)$$

(E_{vis} in MeV) and is composed of two parts. The first term in the square-root represents the statistical uncertainty, and the second one gives the systematic uncertainty, which we will not consider in this study. The observed antineutrino distribution by

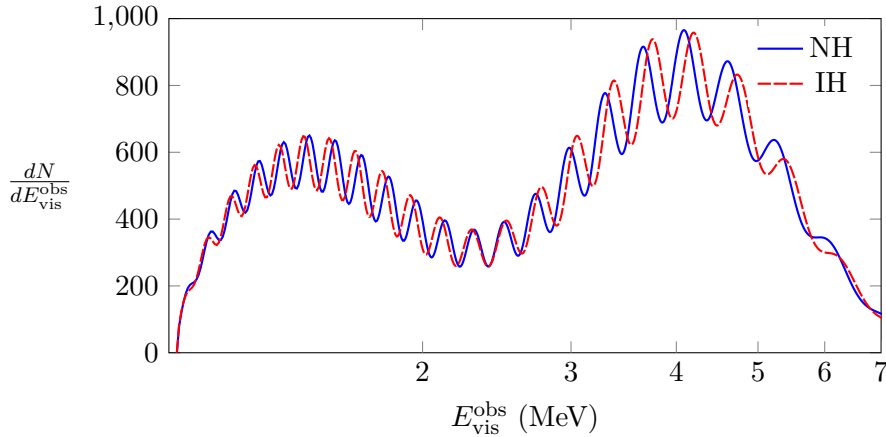


Figure 3: Energy spectrum for 10^5 events, infinite resolution, assuming normal (blue) or inverted (dashed red) hierarchy. It's clear how the $\sin^2 2\vartheta_{13}$ is responsible for the high frequency oscillation while $\sin^2 2\vartheta_{12}$ is responsible for the lowest one.

a detector with N_p free protons after an exposure time T can then be expressed, from (2.1), (2.2), (2.3) and (2.4), as

$$\frac{dN}{dE_{\text{vis}}^{\text{obs}}} = \frac{N_p T}{4\pi L^2} \int_{E_{\text{thr}}}^{\infty} dE_{\bar{\nu}_e} \frac{dN}{dE_{\bar{\nu}_e}} P_{ee}(L, E_{\bar{\nu}_e}) \sigma_{\text{IBD}}(E_{\bar{\nu}_e}) G(E_{\bar{\nu}_e} - 0.8 - E_{\text{vis}}^{\text{obs}}, \delta E_{\text{vis}}^{\text{obs}}). \quad (2.6)$$

The energy spectrum measured by JUNO assuming infinite and finite energy resolution is given in figures 3 and 4, respectively.

3 Sensitivity to the mass hierarchy

After obtaining the energy distribution of reactor antineutrinos, a further step concerns the estimation of the sensitivity in determining the mass hierarchy using a

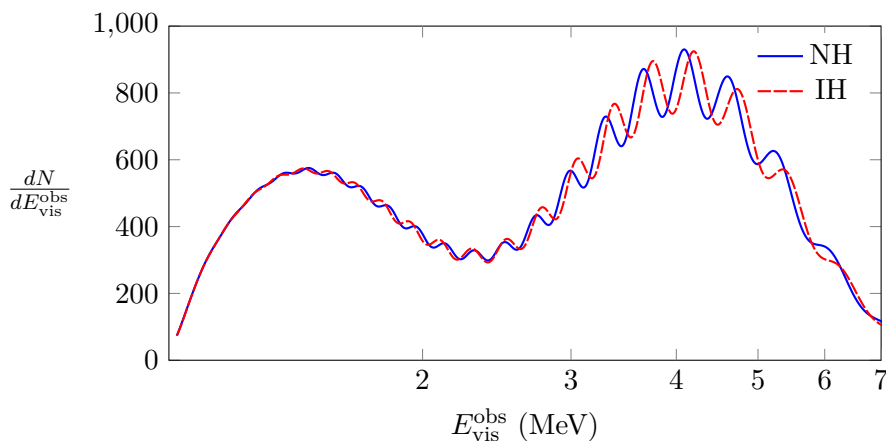


Figure 4: Energy spectrum assuming 10^5 events, $a = 3\%$, and normal (blue) or inverted (dashed red) hierarchy.

χ^2 analysis. To set the stage, we introduce the χ^2 function as

$$\chi^2 = \chi_{\text{stat}}^2 + \chi_{\text{para}}^2 . \quad (3.1)$$

The first term represents the statistical fluctuation, when we introduce binning w.r.t. $E_{\text{vis}}^{\text{obs}}$ it looks like

$$\chi_{\text{stat}}^2 = \sum_i \left(\frac{N_i - N_i^*}{\sqrt{N_i^*}} \right)^2 ,$$

with the summation running over all the bins. Here N_i^* is the number of events for the i_{th} bin while N_i is the theoretical prediction assuming right or wrong mass hierarchy.

The second term summarizes the prior knowledge on mixing parameters. In JUNO these are the mixing angles $\sin^2 2\vartheta_{12}$ and $\sin^2 2\vartheta_{13}$, and the two mass-square differences, Δm_{21}^2 and Δm_{31}^2 (the latter with the sign depending on the mass hierarchy assumption), whose contributions look like

$$\begin{aligned} \chi_{\text{para}}^2 = & \left(\frac{\sin^2 2\vartheta_{12} - (\sin^2 2\vartheta_{12})_{\text{input}}}{\delta \sin^2 2\vartheta_{12}} \right)^2 \\ & + \left(\frac{\sin^2 2\vartheta_{13} - (\sin^2 2\vartheta_{13})_{\text{input}}}{\delta \sin^2 2\vartheta_{13}} \right)^2 \\ & + \left(\frac{\Delta m_{21}^2 - (\Delta m_{21}^2)_{\text{input}}}{\delta \Delta m_{21}^2} \right)^2 \\ & + \left(\frac{\Delta m_{31}^2 - (\Delta m_{31}^2)_{\text{input}}}{\delta \Delta m_{31}^2} \right)^2 . \end{aligned}$$

We then define $\Delta\chi^2$ as

$$\Delta\chi^2 := \chi_{\text{wrong}}^2 - \chi_{\text{true}}^2 ,$$

where χ_{wrong}^2 is the minimum of the χ^2 calculated assuming the wrong hierarchy and χ_{true}^2 is the minimum of the χ^2 calculated assuming the correct hierarchy. It will be often scaled with the number of degrees of freedom, which is clearly equal to the number of fitted data minus the constraints: $n_{\text{bin}} - 6$.

4 Data simulation and fitting

All the datasets are simulated within the ROOT Data Analysis Framework [21], mainly using the built-in Monte Carlo method `TF1::GetRandom` to take into account statistical fluctuations. They are all composed of 10^5 events, which is a realistic assumption considering that the expected event rate calculated with the JUNO experimental setup is about 60 events/day¹ [10] for a total running time of 5 years. The values of the two mixing angles and the two mass-square differences used in the generation are listed in Table 1.

¹With two reactors of 36 GW thermal power at 53 km, a 20-kton Liquid Scintillator detector.

Table 1: The input values Y^{input} and their uncertainties δY taken from [18, 19, 20].

Y	$\sin^2 2\vartheta_{12}$	$\sin^2 2\vartheta_{13}$	$\Delta m_{21}^2 \text{ eV}^2$	$ \Delta m_{31}^2 \text{ eV}^2$
Y^{input}	0.857	0.089	7.50×10^{-5}	2.32×10^{-3}
δY	0.024	0.005	0.20×10^{-5}	0.11×10^{-3}

To take into account the finite detector resolution (2.5) it has been chosen an alternative generation method. The adoption of the method presented above for the infinite resolution case implies having an explicit analytic form (without the integral) of the spectrum function, which is basically impossible to be found, cfr. eq. (2.6). We have proceeded as follows: random events (say E_i^{true} their energies) were generated following the ideal distribution ($a = 0$); then, before filling the histogram, assigned to a new energy value following the normal gaussian distribution with mean E_i^{true} and variance $a\sqrt{E_i^{\text{true}}}$. That is a highly efficient way to generate datasets with finite resolution.

Considering now the fitting procedure, the minimization of (3.1) has required to use directly the ROOT minimization libraries, in particular the `TMinuit` algorithm². In the minimization procedure all the parameters were left free to vary in their physical limits, except for the baseline length, which was fixed, assuming a very small error δL on it.

5 Results

5.1 Infinite resolution: sensitivity and baseline length

We now discuss the results obtained assuming infinite resolution ($a = 0$). In this case the number of bins can be arbitrary high, so it has been chosen 200 as unit. We present the study of the baseline length influence on the sensitivity in distinguishing between the two theoretical hypothesis, namely between normal and inverted hierarchy.

Two different setups, at $L = 50$ km and $L = 30$ km, have been considered and a

²See the appendix.

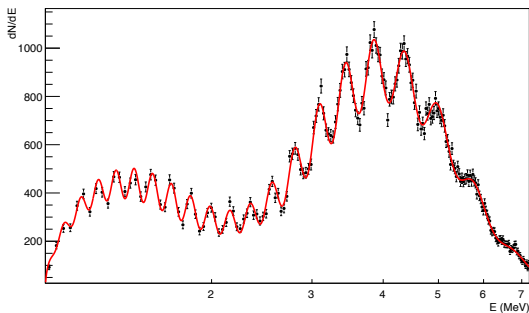


Figure 5: Fitting NH data with NH theory at $L = 50$ km.

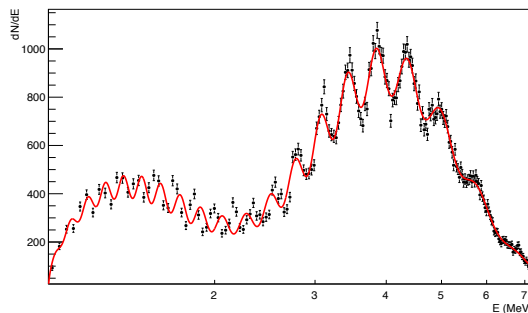


Figure 6: Fitting NH data with IH theory at $L = 50$ km.

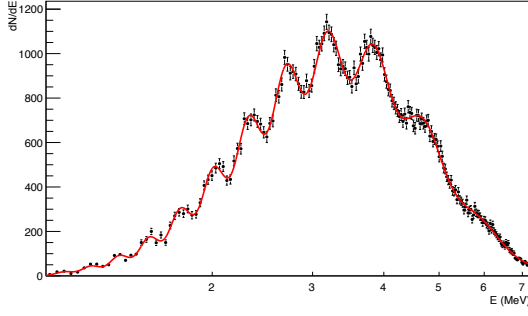


Figure 7: Fitting NH data with NH theory at $L = 30$ km.

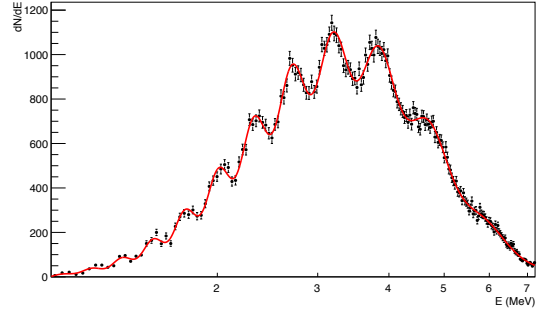


Figure 8: Fitting NH data with IH theory at $L = 30$ km.

fit on NH data, assuming normal or inverted mass hierarchy, has been performed. As it can be noticed with an overlook to the results at 50 km presented in figures 5 and 6, the parameter that changes most at the end of the minimization procedure is $\sin^2 2\vartheta_{13}$, namely the one who is responsible for the tiny oscillations; moreover the total lack of adherence to data in the first 2 MeV it's self-evident. Thus the difference in the two χ^2_{\min} values, $\Delta\chi^2$, is a powerful statistic test to discriminate between normal and inverted hierarchy. In this specific case we find $\Delta\chi^2/ndf \sim 0.6$.

On the other side, comparing results at 30 km in figures 7 and 8 it can be noticed how it's harder to establish which fit is the “wrong” one. That's because both theoretical hypothesis seem to fit data accurately, as a matter of fact we find $\Delta\chi^2/ndf \sim 0$. As a consequence the sensitivity in discriminating between the two theories is higher in the first case and we can say for sure that it strongly depends on the baseline length L .

Distances from 1 to 140 km between the detector and the reactor cores, L , were considered in the calculation of $\Delta\chi^2$. It has been chosen to fit Asimov datasets [13] (where the bin's content is equal to the expectation value) with 400 bins to avoid useless fluctuations in the plot. Looking at the results presented in figure 9 it's clear that exists a range, from 50 km to 70 km, with a maximum in ~ 55 km,

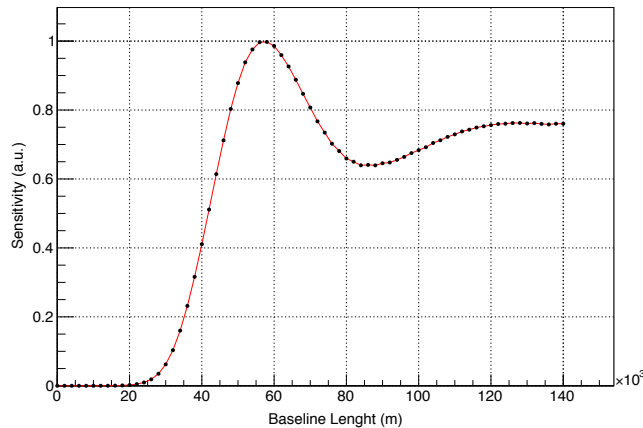


Figure 9: Sensitivity in arbitrary units ($\Delta\chi^2$ normalized to 1 at the peak) w.r.t. the baseline length, assuming $a = 0$.

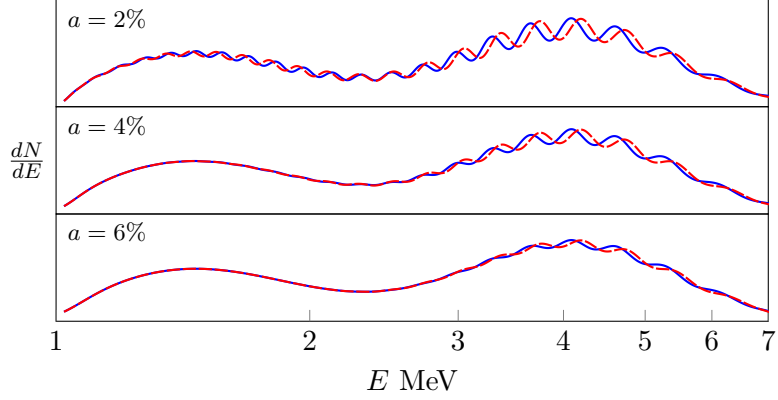


Figure 10: Plots of the spectrum 2.6 with $a = 2\%$, 4% , 6% resolution.

in which the sensitivity is maximized, and we can better discriminate between the two hypothesis. One may look for a second, higher, maximum for $L > 140$ km, but since antineutrino rate (see 2.6) is proportional to $1/L^2$, the time spent collecting the same amount of data would be prohibitive.

As it's described in the next section, the study of the sensitivity w.r.t. the baseline length with finite resolution has not been performed, due to a high computational time request. In [12] this analysis shows how the peak's position moves toward smaller values of L increasing the a value in (2.5).

5.2 Finite resolution: sensitivity and resolution

Introducing a finite resolution in the fitting process is not trivial³, as seen before the problem involves the the presence of an integral in the expression (2.6). Now we study the sensitivity comparing spectrums collected at the same baseline length (at 53 km, as planned) but with different values of the experimental resolution. As it can be seen with just an overlook to the three graphs in figure 10, the difference between the two curves (normal and inverted hierarchy) vanishes when decreasing the resolution parameter, from a (experimentally challenging) $2\%/\sqrt{E/\text{MeV}}$ resolution

³Details of the chosen programming method in the appendix.

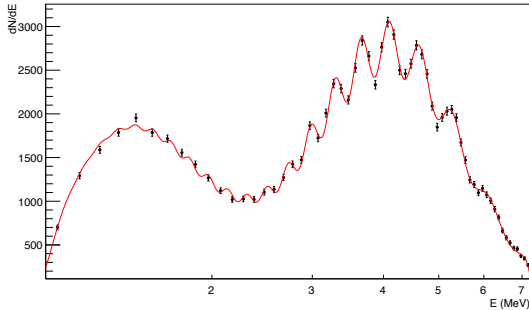


Figure 11: Fitting NH data with NH theory, $a = 3\%$ at $L = 53$ km.

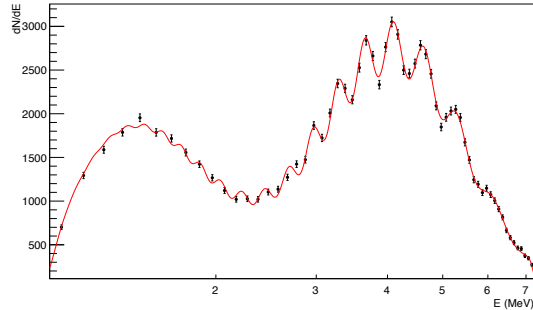


Figure 12: Fitting NH data with IH theory, $a = 3\%$ at $L = 53$ km.

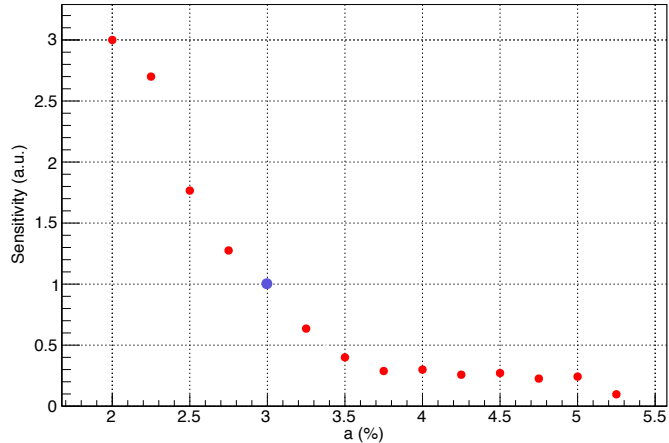


Figure 13: Sensitivity in arbitrary units ($\Delta\chi^2$ normalized to 1 at 3% resolution) w.r.t. the statistical factor a in the resolution expression at $L = 53$ km.

to a much worse $6\%/\sqrt{E/\text{MeV}}$. The resolution effect is strong especially in the first part, between 0 and 2 MeV, where the tiny oscillations are totally smeared out and the curves are nearly indistinguishable.

Let's consider the binning issue. Choosing an arbitrary high number of bins makes not much sense with a finite resolution, therefore a realistic binning of 100 keV, which signifies more or less 62 bins through the entire spectrum, has been adopted.

We discuss as a preliminary analysis the results of fitting NH data (resolution $3\%/\sqrt{E/\text{MeV}}$) either with correct or wrong theory (figures 11 and 12). As can be seen, comparing this with figures 5 and 6, now it's very hard to say which fit is the "wrong" one, the bins are less than the ideal case and the two datasets almost superimposable. This means that the two χ^2 are very similar and we have a smaller $\Delta\chi^2$.

Now, this kind of minimization, because of the complexity of the expression (2.6), requires a not negligible amount of computational time when performing it with the available processors. This drawback is a big limit to the kind of analysis we want to perform; a plot with the same number of points as the one in figure 9 is, computationally speaking, too expensive. In addition to this, with $a < 2\%$ the ROOT algorithm fails to compute the integral with the standard level of tolerance, and increasing it obviously more time-consuming. Then, in the perspective to study the sensitivity ($\Delta\chi^2$) w.r.t. the resolution parameter a we can only make a qualitative analysis.

Let's take a look at the graph in figure 13. Each of the fourteen points is the result of the mean of three different values of $\Delta\chi^2$, computed considering three datasets simulated with the same set of parameters but with a different seed in the generation of the random events. This cuts out the effects of the statistical fluctuations on χ^2 and any bad-precision issues in the minimization algorithm. The $\Delta\chi^2$ results are normalized to make the sensitivity at the JUNO designed resolution $3\%/\sqrt{E/\text{MeV}}$ unitary. It can be easily noticed how the sensitivity rapidly goes to zero when the resolution gets worse, in particular with $a > 3\%$ the ability to

discriminate between the two hypothesis is practically zero. We can easily conclude that any resolution equal or better than the designed one is essential to determinate the mass hierarchy with an acceptable degree of accuracy.

6 Discussion and conclusions

The sensitivity of the JUNO experiment to the determination of the neutrino mass hierarchy has been studied by performing a χ^2 analysis. The results are applicable to all medium baseline reactor electron-antineutrino oscillation experiments with scintillation detection system and a similar active mass.

Taking into account statistical fluctuations in the data and finite bin-size effects, we study the impacts of the baseline length and the energy resolution on the sensitivity, represented by the $\Delta\chi^2$, and find that it strongly depends on them. With infinite resolution ($a = 0$) the optimal baseline length is found to be in the interval [50, 70] km and at the actual indicated JUNO site ($L = 53$ km) an energy resolution better or equal to the $3\%/\sqrt{E/\text{MeV}}$ level is needed to determine the neutrino mass hierarchy pattern. With these experimental settings, JUNO would determine the mass hierarchy after five or more years of running.

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Appendix: Fitting with the TMinuit class

When we need to perform a fit not only minimizing the standard χ^2

$$\chi_{\text{std}}^2 = \sum_i \left(\frac{N_i - N_i^*}{\sqrt{N_i^*}} \right)^2$$

but a more general function, which we will from now on call χ^2 , the build-in ROOT method `TF1::Fit` is no longer exploitable. This method, given data and a `TF1` object, uses the `TMinuit` libraries in order to minimize χ_{std}^2 and efficiently provide results. Thus our general approach needs to explicitly write the chosen χ^2 function and directly use `TMinuit` to minimize it.

Let's first consider the infinite resolution case of the spectrum (2.6), when its expression it's explicit. After declaring the `TMinuit` object who performs the minimization (`gMinuit`), we first need to set the function to minimize (`myFCN`), the χ^2 function, with the command

```
gMinuit->SetFCN(myFCN);
```

The implementation of `myFCN` must be standard [21]:

```
void myFCN( int &npar, double* gin, double &f, double* par, int flag) {
    f = ... ; // implement the function
}
```

where `f` is the value of the χ^2 function, `par` is the pointer to the array of function parameters and `npar` its dimension; `TMinuit` will vary the values it points in the minimization procedure. We will not consider the other arguments. In the specific case we are considering, we can calculate the value of the χ_{std}^2 using a `for` loop:

```
double chi2 = 0;
double* xi = new double[1];
for ( int i = 1 ; i <= dataset->GetSize()-2 ; ++i ) {

    xi[0] = Emin + (Emax-Emin)/(2*N_div) + (i-1)*(Emax-Emin)/N_div;
    // bin's mid point
    chi2 += pow( (spectrum->EvalPar( xi , par ) -
        dataset->GetBinContent(i)) / datasetNH->GetBinError(i) ,2);
}
f = chi2;
```

where `spectrum` is a `TF1` object corresponding to our theoretical distribution with its parameters and `dataset` is a `TH1F` object with `N_div` number of bins between the range `[Emin, Emax]` containing the data to fit. The important thing here is that `spectrum` is evaluated in each bin's mid point and in the parameter values pointed to by `par`, the latter passed from the `TMinuit` object to `myFCN` to test the χ^2 value during the minimization procedure. This implementation allows to change the χ_{std}^2 expression or add other terms.

Now that the function to minimize is set we can initialize the parameters and add their limits with the `mnparm` function, then launch the minimization procedure with the `mnexcm` function. There are three algorithms in `TMinuit`: `SEEK` consists in

a preliminary Monte Carlo search of the minimum, **SIMPLEX** finds quickly its rough value and **MIGRAD** improves the estimation. **SEEK** it's often used to find the absolute minimum before improving it, as a matter of fact it can happen that **SIMPLEX** or **MIGRAD** converges to a minimum which is only local.

Applying this procedure to the finite resolution case needs a little code enhancement to bypass the fact we do not have an explicit form of the spectrum as above, therefore using a **TF1** object to represent it is now obviously impossible. The sleight of hand we use is to write the integrand of (2.6)

$$f(E, \hat{E}) = \frac{N_p T}{4\pi L^2} \frac{dN}{dE}(E) P_{ee}(E) \sigma_{\text{IBD}}(E) G(E - 0.8 - \hat{E}, \delta \hat{E})$$

as a **TF1** with E as the main variable x and \hat{E} as a parameter, then in the calculation of χ^2 integrate in dE for every point using \hat{E} as the main variable. Here is the code:

```
for ( int i = 1 ; i <= dataset->GetSize()-2 ; ++i ) {

    par[7] = Emin + (Emax-Emin)/(2*N_div_R) + (i-1)*(Emax-Emin)/N_div_R;
    chi2 += pow( ( integrand->Integral( Ethr , Max , par , 1E-13) -
        dataset->GetBinContent(i)) / dataset->GetBinError(i) ,2);
}
```

The parameter `par[7]` represents \hat{E} , `Ethr` it's the energy threshold, `Max` a sufficient high energy value where the antineutrino distribution is practically zero, and the fourth argument of `TF1::Integral` is the tolerance set in the **ROOT**'s integration algorithm. We also must call back the parameters value from the **TMinuit** object before every calculation of χ^2 inserting

```
for ( int i = 0 ; i < npar ; ++i ) {
    gMinuit->GetParameter( i , par[i] , parErr[i] );
}
```

inside the `myFCN` declaration before the `for` loop.

This code, unlike the first one, which requires a negligible computational time in this kind of analysis, it's definitely time consuming: several minutes are required for a complete minimization. Responsible for this is the calculation of the integral, which is not trivial even with numerical methods, and decreasing the tolerance value obviously worsen things.

The limits given by the complex kind of integration also influence the range of resolution values we want to test. For $a < 2\%$ the integration algorithm with a tolerance of 10^{-13} hardly converges all over the points, and the value of χ^2 is then distorted. Clearly a smaller tolerance value prevents this issue, but the amount of requested time for each iteration becomes unacceptable when we need to iterate it in a program that acts in a large number of situations.

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