Sensitivity to the neutrino mass hierarchy
with the JUNO experiment

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1 Introduction

The Standard Model of particle physics 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, 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., and a deficit of their flux as compared with the prediction from the Standard Solar Model was also established by Bahcall et al. 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 \( m_{\nu_3} > m_{\nu_1} \) 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, \( \nu_e \) and \( \bar{\nu}_e \), muon, \( \nu_\mu \) and \( \bar{\nu}_\mu \), and tauon, \( \nu_\tau \) and \( \bar{\nu}_\tau \). The flavour is conserved: so \( \nu_e \) is the neutrino which is produced with \( e^+ \), or produces an \( e^- \) in CC weak interaction processes; \( \nu_\mu \) is the neutrino which is produced with \( \mu^+ \), or produces \( \mu^- \), 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 \( \nu_e, \nu_\mu, \nu_\tau \) (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 \( \nu_\alpha \), with energy \( E \) is produced in some weak interaction process, at a sufficiently large distance \( L \) from the \( \nu_\alpha \) source, the probability to find a neutrino of a different flavour, say \( \nu_\beta \), \( P(\nu_\alpha \rightarrow \nu_\beta; E, L) \), 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 \( \nu_\mu \) will not change into a neutrino of a different flavour, i.e., the “\( \nu_\mu \) survival probability” \( P(\nu_\alpha \rightarrow \nu_\alpha; E, L) \) will be smaller than one. If only muon neutrinos \( \nu_\alpha \) 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 \( \nu_\alpha \) 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 \( \nu_1, |\nu_1\rangle \), 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 $3 \times 3$ 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 $U \rightarrow U^\ast$ (complex conjugate).

If the $3 \times 3$ 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 $c_{ij} := \cos \theta_{ij}$ and $s_{ij} := \sin \theta_{ij}$ 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:

$$P(\bar{\nu}_e \rightarrow \bar{\nu}_e; E, L) := |A(\bar{\nu}_e \rightarrow \bar{\nu}_e)|^2$$

$$= 1 - 4 \left( \sum_i |U_{ei}|^2 \right)^{-2} \sum_{i<j} \left( |U_{ei}|^2 |U_{ej}|^2 \sin^2 \frac{\Delta m^2_{ij} L}{4E} \right), \quad \text{(1.1)}$$

where $\Delta m^2_{ij} = m_i^2 - m_j^2$ (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, $\sin^2 2\theta_{12}$ and $\sin^2 2\theta_{13}$, and two mass-square differences $\Delta m^2_{31}$ and $\Delta m^2_{21}$. Past experiments \[11\] have measured all these parameters, but not the sign of $\Delta m^2_{31}$, 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 $\Delta m^2_{31}$ 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
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” Photomultipliers 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 $\gamma$-rays, which gives a prompt signal. The neutron scatters in the detector until being thermalized. It is then captured by a proton, on average $200\mu s$ later, and releases a 2.2-MeV $\gamma$-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 $^{235}\text{U}$, $^{238}\text{U}$, $^{239}\text{Pu}$ and $^{241}\text{Pu}$ in the

![Figure 1: Neutrino mass hierarchy hypothesis.](image)
Thermalization

Prompt

Delayed

\[ \bar{\nu}_e \]

\[ p \]

\[ n \]

\[ e^+ \]

\[ e^- \]

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

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

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

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

\[ f_k := \frac{N_{fiss}^k}{\sum_i N_{fiss}^i}. \]

 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: \(f_{235U} = 0.58, f_{239Pu} = 0.30, f_{238U} = 0.07\) and \(f_{241Pu} = 0.05\) [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

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

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

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

\[
P_{\bar{\nu}_e \to \bar{\nu}_e} \equiv P_{ee} = 1 - \cos^4 \theta_{13} \sin^2 2\theta_{12} \sin^2 \Delta_{21} \\
- \cos^2 \theta_{12} \sin^2 2\theta_{13} \sin^2 \Delta_{31} \\
- \sin^2 \theta_{12} \sin^2 2\theta_{13} \sin^2 \Delta_{32},
\]

(2.2)
The variables \( m_i \) and \( E_i \) are the mass and energy of the corresponding mass eigenstate, while \( \vartheta_{ij} \) represent the neutrino mixing angles. The oscillation phases \( \Delta_{ij} \) are defined as
\[
\Delta_{ij} := \frac{\Delta m_{ij}^2 L}{4E_{\bar{\nu}_e}} = 1.27 \cdot \frac{[\text{eV}^2][m]}{[\text{MeV}]} ,
\]
with a baseline length from the reactor \( L \).

While \( \Delta m_{21}^2 \) is fixed, \( \Delta m_{31}^2 \) 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 \( E_{\text{thr}} \sim m_n - m_p + m_e \sim 1.804 \text{ MeV} \), and the cross section is [17]
\[
\sigma_{\text{IBD}} = E_{e^+} p_{e^+} \cdot 9.52 \cdot 10^{-44} \text{ cm}^2,
\]
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 \( E_{\text{vis}} \) 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 \( E_{\text{vis}} \) 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 \( \delta E_{\text{vis}} \)
\[
G(E_{\text{vis}} - E_{\text{vis}}^{\text{obs}}, \delta E_{\text{vis}}) = \frac{1}{\sqrt{2\pi\delta E_{\text{vis}}}} \exp \left\{ -\frac{(E_{\text{vis}} - E_{\text{vis}}^{\text{obs}})^2}{2(\delta E_{\text{vis}})^2} \right\}.
\]
Generally the detector energy resolution can be modeled as
\[
\frac{\delta E_{\text{vis}}}{E_{\text{vis}}} = \sqrt{\frac{a^2}{E_{\text{vis}}} + b^2}
\]
(\( E_{\text{vis}} \) 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

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

(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.
\( \chi^2 \) analysis. To set the stage, we introduce the \( \chi^2 \) function as
\[
\chi^2 = \chi^2_{\text{stat}} + \chi^2_{\text{para}} \quad \text{(3.1)}
\]
The first term represents the statistical fluctuation, when we introduce binning w.r.t. \( E_{\text{vis}} \) it looks like
\[
\chi^2_{\text{stat}} = \sum_i \left( \frac{N_i - N^*_i}{\sqrt{N^*_i}} \right)^2,
\]
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 \( \sin^2 2\theta_{12} \) and \( \sin^2 2\theta_{13} \), and the two mass-square differences, \( \Delta m^2_{21} \) and \( \Delta m^2_{31} \) (the latter with the sign depending on the mass hierarchy assumption), whose contributions look like
\[
\chi^2_{\text{para}} = \left( \frac{\sin^2 2\theta_{12} - (\sin^2 2\theta_{12})_{\text{input}}}{\delta \sin^2 2\theta_{12}} \right)^2 + \left( \frac{\sin^2 2\theta_{13} - (\sin^2 2\theta_{13})_{\text{input}}}{\delta \sin^2 2\theta_{13}} \right)^2 + \left( \frac{\Delta m^2_{21} - (\Delta m^2_{21})_{\text{input}}}{\delta \Delta m^2_{21}} \right)^2 + \left( \frac{\Delta m^2_{31} - (\Delta m^2_{31})_{\text{input}}}{\delta \Delta m^2_{31}} \right)^2.
\]

We then define \( \Delta \chi^2 \) as
\[
\Delta \chi^2 := \chi^2_{\text{wrong}} - \chi^2_{\text{true}},
\]
where \( \chi^2_{\text{wrong}} \) is the minimum of the \( \chi^2 \) calculated assuming the wrong hierarchy and \( \chi^2_{\text{true}} \) is the minimum of the \( \chi^2 \) 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 \texttt{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/day [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.

\[1\] With two reactors of 36 GW thermal power at 53 km, a 20-kton Liquid Scintillator detector.
Table 1: The input values $Y^{\text{input}}$ and their uncertainties $\delta Y$ taken from [18, 19, 20].

| $Y$ | $\sin^2 2\theta_{12}$ | $\sin^2 2\theta_{13}$ | $\Delta m^2_{21}$ $eV^2$ | $|\Delta m^2_{31}|$ $eV^2$ |
|-----|---------------------|---------------------|---------------------|---------------------|
| $Y^{\text{input}}$ | 0.857 | 0.089 | $7.50 \times 10^{-5}$ | $2.32 \times 10^{-3}$ |
| $\delta Y$ | 0.024 | 0.005 | $0.20 \times 10^{-5}$ | $0.11 \times 10^{-3}$ |

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 $E_{i}^{\text{true}}$ 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 $E_{i}^{\text{true}}$ 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 algorithm. 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 $\delta L$ 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

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\[\text{See the appendix.}\]

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

Figure 6: Fitting NH data with IH theory at $L = 50$ 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 $\sin^2 2\theta_{13}$, 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 $\chi^2_{\text{min}}$ values, $\Delta \chi^2$, is a powerful statistic test to discriminate between normal and inverted hierarchy. In this specific case we find $\Delta \chi^2 / \text{ndf} \sim 0.6$.

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 $\Delta \chi^2 / \text{ndf} \sim 0$. 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 $\sim 55$ km,
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 trivial\(^3\) 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

\(^3\)Details of the chosen programming method in the appendix.
### Table 1: Sensitivity in arbitrary units ($\Delta \chi^2$ normalized to 1 at 3% resolution) with respect to the statistical factor $a$ in the resolution expression at $L = 53\,\text{km}$.

<table>
<thead>
<tr>
<th>$a$ (%)</th>
<th>Sensitivity (a.u.)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>0.5</td>
</tr>
<tr>
<td>2.5</td>
<td>1.0</td>
</tr>
<tr>
<td>3</td>
<td>1.5</td>
</tr>
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<tr>
<td>4.5</td>
<td>3.0</td>
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<td>5</td>
<td>3.5</td>
</tr>
<tr>
<td>5.5</td>
<td>4.0</td>
</tr>
</tbody>
</table>

Figure 13: Sensitivity in arbitrary units ($\Delta \chi^2$ normalized to 1 at 3% resolution) with respect to the statistical factor $a$ in the resolution expression at $L = 53\,\text{km}$.

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 $\chi^2$ 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 $\chi^2$ 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 $\chi^2$ 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 $\chi^2$

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

but a more general function, which we will from now on call $\chi^2$, 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 $\chi^2_{\text{std}}$ and efficiently provide results. Thus our general approach needs to explicitly write the chosen $\chi^2$ 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 $\chi^2$ function, with the command

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

The implementation of myFCN must be standard [21]:

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

where $f$ is the value of the $\chi^2$ 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 $\chi^2_{\text{std}}$ using a for loop:

```c++
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 $N_{\text{div}}$ number of bins between the range $[\text{Emin, Emax}]$ 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 $\chi^2$ value during the minimization procedure. This implementation allows to change the $\chi^2_{\text{std}}$ 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 to write the integrand of (2.6)

\[ f(E, \hat{E}) = \frac{N_\nu T}{4\pi L^2} \frac{dN}{dE}(E) P_{ee}(E) \sigma_{IBD}(E) G(E - 0.8 - \hat{E}, \delta \hat{E}) \]

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

```c++
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 \( \text{par}[7] \) 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 \text{TF1::Integral} is the tolerance set in the ROOT’s integration algorithm. We also must call back the parameters value from the \text{TMinuit} object before every calculation of \( \chi^2 \) inserting

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

inside the \text{myFCN} declaration before the \text{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 \( \chi^2 \) 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.
References


