Observing Supernova Neutrino Light Curves with Super-Kamiokande. V. Distance Estimation with Neutrinos Alone

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ABSTRACT

Neutrinos are pivotal signals in multi-messenger observations of supernovae. Recent advancements in the analysis method of supernova neutrinos, especially in quantitative analysis, have significantly broadened scientific possibilities. This study demonstrates the feasibility of estimating distances to supernovae using quantitative analysis techniques for supernova neutrinos. This estimation utilizes the direct relationship between the radius of a neutron star and the distance to the supernova. The radius of a neutron star is determined with an approximate uncertainty of 10% through observations such as X-rays and gravitational waves. By integrating this information, the distance to the supernova can be estimated with an uncertainty of within 15% at a 95% confidence level. It has been established that neutrinos can pinpoint the direction of supernovae, and when combined with distance estimates, threedimensional localization becomes achievable. This capability is vital for follow-up observations using multi-messenger approaches. Moreover, more precise distance determinations to supernovae through follow-up observations, such as optical observations, allow for accurate measurements of neutron star radii. This data, via the neutron star mass-radius relationship, could provide various insights into nuclear physics.

Keywords: Supernova neutrinos (1666); Neutrino astronomy (1100); Neutrino telescopes (1100); Corecollapse supernovae (304); Neutron stars (1108);

1. INTRODUCTION

Corresponding author: Yudai Suwa suwa@yukawa.kyoto-u.ac.jp A supernova (SN) produces a large number ($\sim 10^{58}$) of neutrinos. These neutrinos, originating from thermal processes, are emitted almost isotropically. As a result, neutrinos generated in nearby supernovae (SNe) ensure their detectability (see Janka 2017; Horiuchi & Kneller 2018; Müller 2019; Vartanyan & Burrows 2023; Yamada et al. 2024, and references therein). It is noteworthy that neutrinos are emitted before electromagnetic radiation, as their production occurs while the shock wave is still confined within the star (Kistler et al. 2013; Abe et al. 2016). Therefore, observing SN neutrinos is an initial step for further multiband observational studies that utilize the broad electromagnetic wavelength, extending from radio to gamma rays. Within the realm of multi-messenger studies in SN research, neutrinos are deemed paramount from diverse scientific perspectives.

Recently, advances have been made in the methods used for the quantitative analysis of neutrinos from SNe. This field particularly focuses on the period starting a few seconds after the formation of the protoneutron star (PNS) from the SN. At this point, the PNS is no longer contracting,¹ the falling back accretion onto it from the ejecta has become minor, and neutrinos start to come out from the deep inside of the PNS in a simpler way, through diffusion. This makes the situation much easier to describe physically, and quantitative analysis becomes possible. We mainly aim to understand this stage with long-term simulations (Suwa et al. 2019; Mori et al. 2021; Nakazato et al. 2022), find analytic ways to describe how neutrinos are emitted (Suwa et al. 2021), and construct the pipeline code for data analysis (Suwa et al. 2022; Harada et al. 2023).

In this paper, we investigate the potential of determining the distance to an SN through quantitative analysis of SN neutrinos. A pivotal element of this study is the direct relationship between the distance of the SN and the radius of the PNS. Essentially, determining the radius of the PNS facilitates the estimation of the distance to the SN, and vice versa. This study evaluates the precision with which the distance to an SN can be measured using NS radii derived from alternative observations such as X-rays.

This paper is organized as follows. Section 2 describes the model used to generate mock data, while Section 3 details the analysis of these mock samples for parameter estimation. Section 4 discusses the implications of the findings, and Section 5 provides a summary of the main results.

2. MOCK SAMPLING

In this work, we use the same method to perform parameter estimation as Suwa et al. (2022) and Harada et al. (2023), in which we use the solution for the neutrino light curve derived in Suwa et al. (2021). The time evolution of the event rate and positron average energy are given by analytic functions of time. The parameter dependence on the mass and radius of the PNS are also explicitly presented.

The event rate is given by

$$\mathcal{R} = 720 \,\mathrm{s}^{-1} \left(\frac{M_{\mathrm{det}}}{32.5 \,\mathrm{kton}}\right) \left(\frac{D}{10 \,\mathrm{kpc}}\right)^{-2} \\ \times \left(\frac{M_{\mathrm{PNS}}}{1.4 \,M_{\odot}}\right)^{15/2} \left(\frac{R_{\mathrm{PNS}}}{10 \,\mathrm{km}}\right)^{-8} \left(\frac{g\beta}{3}\right)^{5} \\ \times \left(\frac{t+t_0}{100 \,\mathrm{s}}\right)^{-15/2}, \qquad (1)$$

where M_{det} is the detector mass with 32.5 kton corresponding to the entire inner detector volume of Super-Kamiokande (SK),² D is the distance between the SN and Earth, M_{PNS} is the PNS mass, R_{PNS} is the PNS radius,³ g is the surface structure correction factor, and β is the opacity boosting factor from coherent scattering (see Suwa et al. 2021, for details). The timescale t_0 is given by

$$t_0 = 210 \,\mathrm{s} \left(\frac{M_{\mathrm{PNS}}}{1.4 \, M_{\odot}}\right)^{6/5} \left(\frac{R_{\mathrm{PNS}}}{10 \,\mathrm{km}}\right)^{-6/5} \\ \times \left(\frac{g\beta}{3}\right)^{4/5} \left(\frac{E_{\mathrm{tot}}}{10^{52} \,\mathrm{erg}}\right)^{-1/5}, \tag{2}$$

where E_{tot} is the total energy emitted by all flavors of neutrinos. By integrating Eq. (1) with Eq. (2), the expected total number of events is

$$N = \int_0^\infty \mathcal{R}(t) dt$$

= 89 $\left(\frac{M_{\text{det}}}{32.5 \,\text{kton}}\right) \left(\frac{D}{10 \,\text{kpc}}\right)^{-2} \left(\frac{M_{\text{PNS}}}{1.4 \,M_\odot}\right)^{-3/10}$
 $\times \left(\frac{R_{\text{PNS}}}{10 \,\text{km}}\right)^{-1/5} \left(\frac{g\beta}{3}\right)^{-1/5} \left(\frac{E_{\text{tot}}}{10^{52} \,\text{erg}}\right)^{13/10}.$ (3)

For the canonical parameters used in this paper ($M_{\rm det} = 32.5$ kton, D = 8 kpc, $M_{\rm PNS} = 1.52 M_{\odot}$, $R_{\rm PNS} = 12.4$ km, $g\beta = 1.6$, and $E_{\rm tot} = 10^{53}$ erg), the expectation number becomes N = 2940.

¹ This implies that the radius of the PNS has already converged to that of the cold neutron star (NS). In the subsequent discussion, the radius of the PNS and that of NS are not distinguished.

 $^{^2}$ In this study, we employ the full 32.5 kton volume of the SK inner detector. This is because, at least for a Galactic SN, the timescale of data analysis is short, and we can avoid significant contamination from backgrounds (Mori et al. 2022). For a more detailed discussion, see Suwa et al. (2019).

 $^{^3}$ This corresponds to the radius after PNS contraction by neutrino cooling.

The average energy of created positrons is given by

$$E_{e^+} = 25.3 \,\mathrm{MeV} \left(\frac{M_{\mathrm{PNS}}}{1.4 \, M_{\odot}}\right)^{3/2} \\ \times \left(\frac{R_{\mathrm{PNS}}}{10 \,\mathrm{km}}\right)^{-2} \left(\frac{g\beta}{3}\right) \left(\frac{t+t_0}{100 \,\mathrm{s}}\right)^{-3/2}.$$
(4)

For the energy distribution, we employ the Fermi-Dirac distribution function for neutrinos, which allows us to calculate the distribution of the positron. Note that the above estimates are based on the simple expression for the cross section of the inverse beta decay. More precise expressions are given in Vogel & Beacom (1999); Strumia & Vissani (2003), which we used for our numerical estimates in Nakazato et al. (2022).

Importantly, by combining Eqs. (1) and (4) we find

$$D = 10 \operatorname{kpc} \left(\frac{\mathcal{R}}{720 \operatorname{s}^{-1}} \right)^{-1/2} \left(\frac{E_{e^+}}{25.3 \operatorname{MeV}} \right)^{5/2} \times \left(\frac{M_{\operatorname{det}}}{32.5 \operatorname{kton}} \right)^{1/2} \left(\frac{R_{\operatorname{PNS}}}{10 \operatorname{km}} \right).$$
(5)

Note that it is independent of t, g, and β . This equation has two meanings. If we know the radius R_{PNS} , we can estimate the distance D and vice versa. In Suwa et al. (2022); Harada et al. (2023), we assumed that Dis known by other means so that we could constrain the radius. In the following, we will show how accurately we can estimate the distance before the other observations are available.

Based on the equations presented here, we perform 100 Monte Carlo simulations of SN neutrinos, each using different random seeds (see Figure 1 in Suwa et al. 2022, for a specific example). In the next section, we explain the data analysis method.

3. PARAMETER ESTIMATION

Parameter estimation is conducted by fitting the Monte-Carlo data described in Section 2 to analytic solutions, aiming to assess the accuracy and reliability of our method (see Figure 2 in Suwa et al. 2022). Detailed descriptions of the numerical setting, the fitting procedure, and the statistical methods used for the analysis are provided below.

Here, we employ the Gaussian-approximation analysis (see Harada et al. 2023, for details), i.e., the chi-square fitting using the event rate and the average energy. This is because we are interested in Galactic SNe, so the expected event number is large enough for the Poisson distribution to be approximated by the Gaussian distribution. Also, in this work, we increase the number of parameters from three to four so that computational cost becomes greater than the previous work and spectral likelihood analysis becomes time-consuming. For binning and the probability density function definition, see Suwa et al. (2022). For completeness, we summarize the following. The time bins are calculated by the equations:

$$t_i = t_{i-1} + \Delta t_i, \tag{6}$$

$$\Delta t_i = A \Delta t_{i-1},\tag{7}$$

where Δt_i represents the time width of the *i*-th time bin, and A is a constant. This constant is determined using the first time bin, t_1 , and the last time bin of the analysis, t_{end} . For this paper, we set Δt_1 to 0.5 and t_{end} to 100 seconds. We chose the number of bins, N, to be 20, resulting in an approximate value of A equal to 1.206. The χ^2 is calculated as follows:

$$\chi^2 = \sum_{i=1}^{N} \frac{(O_i - X_i)^2}{\sigma_i^2},$$
(8)

where O_i , X_i , and σ_i represent the observed value, the expected value, and the variance, respectively, each associated with the time index *i*. For the event rate $(X = \mathcal{R})$, the variance is calculated using $\sigma_i^2 = \mathcal{R}_i^2/N_i$, where N_i denotes the number of events in the *i*-th bin. For the average energy $(X = E_{e^+})$, we adopt $\sigma_i^2 = (0.05E_{e^+})^2$, as outlined by Nakazato et al. (2022), which demonstrates that the statistical error of the average energy is at the level of several percent.

In Suwa et al. (2022), we employed the joint probability density function (PDF) for the measured parameters as

$$\mathcal{P}(M_{\rm PNS}, R_{\rm PNS}, E_{\rm tot}) \propto e^{-\chi^2 (M_{\rm PNS}, R_{\rm PNS}, E_{\rm tot})/2}.$$
 (9)

Instead of Eq. (9) that assumes the uniform prior for all parameters, we employ the following PDF, including Gaussian prior for the radius as

$$\mathcal{P}(M_{\text{PNS}}, R_{\text{PNS}}, E_{\text{tot}}, D)$$

$$\propto e^{-\chi^2 (M_{\text{PNS}}, R_{\text{PNS}}, E_{\text{tot}}, D)/2} \times e^{-(R_{\text{PNS}} - \bar{R})^2/(2\sigma_R^2)}, \quad (10)$$

where \bar{R} and σ_R are the expected mean value and its uncertainties of PNS radius. In this work, we employ $\bar{R} = 12.4$ km and $\sigma_R = 0.7$ km based on Miller et al. (2021), suggesting that the NS radius is not strongly dependent on its mass. For the other parameters, we employ uniform prior within $1.0 < M_{\rm PNS}/M_{\odot} < 2.0$, $0.5 < E_{\rm tot}/(10^{53} \,{\rm erg}) < 2.0$, and $3 < D/\rm{kpc} < 13$, respectively.

Figure 1 presents the distribution of \mathcal{P} . Solid and dashed contours in this figure represent $\mathcal{P}/\mathcal{P}_{\text{max}} = 1/e$



Figure 1. A sample of probability density function (PDF) determined by Eq. (10). Contours with solid and dashed lines correspond to $\mathcal{P}/\mathcal{P}_{\text{max}} = 1/e$ (0.368, corresponding to 1σ) and $1/e^2$ (0.135, 2σ), respectively, where \mathcal{P}_{max} is the maximum value of the PDF.

(equivalent to 0.368 or 1σ) and $1/e^2$ (0.135, corresponding to 2σ), respectively, with \mathcal{P}_{max} denoting the peak value of \mathcal{P} . It is important to note that the uncertainties depicted are based on a single realization. Given that the observed data can exhibit variations due to Poisson statistics, conducting Monte Carlo simulations for multiple realizations is crucial to accurately assess the expected parameter sensitivity (or expected error) in preparation for actual observational data.

The expected parameter sensitivity is assessed through 100 realizations of the aforementioned model.⁴ Each realization, processed via Monte Carlo simulations, yields a variety of best-fit values in accordance with Poisson statistics. However, the cumulative average demonstrates that the preset input values are most likely to be accurate, with a significant decrease in probability density for values diverging from the initial inputs. To estimate the expected parameter sensitivity, we use the median of the compiled average PDF. The uncertainty levels of 68% and 95% are determined by the range of

 Table 1. Expected Values and Statistical Errors with only

 Neutrinos

	input	Median	68%	95%
$M_{\rm PNS}~(M_{\odot})$	1.52	1.58	$^{+0.13}_{-0.12}$	$^{+0.26}_{-0.24}$
$R_{\rm PNS}~({\rm km})$	12.4	12.5	$^{+0.7}_{-0.7}$	$^{+1.4}_{-1.4}$
$E_{\rm tot} \ (10^{53} \ {\rm erg})$	1.00	1.05	$^{+0.15}_{-0.13}$	$^{+0.31}_{-0.25}$
$D \; (\mathrm{kpc})$	8.00	8.10	$^{+0.60}_{-0.56}$	$^{+1.24}_{-1.08}$

⁴ The calculation with 1,000 realizations yields almost identical results, indicating that the calculation has reasonably converged (see Suwa et al. 2022).



Figure 2. Flowchart depicting the process of supernova (SN) discovery using neutrinos (ν) and subsequent analyses. The chart outlines steps from initial detection through data analysis involving specific equations, integration of multimessenger follow-up observations, distance measurement with electromagnetic (EM) signals, and culminating in the determination of the mass-radius (M-R) relationship of neutron stars (NS).

parameters that correspond to these specific probability levels, centered around the median. The findings are compiled in Table 1. Due to the uncertainty imposed by the priors on the radius, it is evident that the precision in determining the mass and total energy is subject to greater uncertainty compared to our previous study (see Table 1 in Suwa et al. 2022). However, it is noteworthy that the distance to the SN has been determined with a precision of within 15% at a 95% confidence level, representing a significant new piece of information. To assess the impact of the number of detected neutrinos on uncertainty, we repeated the procedure for a case where D = 5 kpc. The uncertainty in this case is 14%, indicating that the uncertainty is primarily determined by the prior imposed on R. We also investigate the impact of prior distribution, assuming $\sigma_R = 1$ km. The resulting error in the distance estimation is approximately 20% at a 95% confidence level.

4. IMPLICATION

Here, we discuss the application of distance estimation. The flowchart of the analysis is shown in Figure 2.

After observing neutrinos, it takes time for the shock wave to propagate through the star. Hence, there is a delay before the onset of electromagnetic radiation: ap-



Figure 3. A schematic image for identifying the supernova candidate based on the neutrino signals detected at SK (blue cylinder at the center). Analysis of the signals gives directional information (orange triangle region) and distance measurement (green circle). Combining them, the three-dimensional position of the supernova progenitor may be identified.

proximately 10^5 seconds for red supergiants, 10^4 seconds for blue supergiants, and 10^2 seconds for Wolf-Rayet stars (Kistler et al. 2013). On the other hand, in principle, neutrino detectors can issue an alert within a few minutes (Abe et al. 2016; Al Kharusi et al. 2021; Kashiwagi et al. 2024). If the SN progenitor can be identified through neutrino observations, conducting multimessenger observations of the shock breakout becomes possible. Previously, direction determination by neutrinos has been discussed in various studies (Beacom & Vogel 1999; Ando & Sato 2002; Abe et al. 2016; Linzer & Scholberg 2019; Mukhopadhyay et al. 2020; Kashiwagi et al. 2024), in which the direction would be determined within several degrees. On the other hand, this study has revealed that it is possible to estimate distances using neutrinos alone (Eq. 5) (see Segerlund et al. 2021; Bendahman et al. 2024, for a different approach). Combining the estimated direction and distance makes it possible to determine the three-dimensional position of the SN progenitor (see Figure 3). If there is only one massive star at the estimated position, it becomes possible to uniquely determine the progenitor star before electromagnetic observation of the explosion.⁵ This information is essential, particularly for follow-up observations by telescopes with limited sky coverage. At

this stage, the accuracy of distance estimation depends on the accuracy of determining the NS radius, which is $\mathcal{O}(10)\%$.

Next, we consider the use of neutrino data after the realization of electromagnetic observations. Suppose that the distance can be determined with an accuracy of $\mathcal{O}(1)\%$ through observations by, for instance, optical telescopes.⁶ In that case, imposing a prior distribution on the distance and estimating the NS radius is possible. Here, Eq. (5) should be changed as

$$R_{\rm PNS} = 10 \,\mathrm{km} \left(\frac{\mathcal{R}}{720 \,\mathrm{s}^{-1}}\right)^{1/2} \left(\frac{E_{e^+}}{25.3 \,\mathrm{MeV}}\right)^{-5/2} \\ \times \left(\frac{M_{\rm det}}{32.5 \,\mathrm{kton}}\right)^{-1/2} \left(\frac{D}{10 \,\mathrm{kpc}}\right). \tag{11}$$

Furthermore, as demonstrated in previous studies (Suwa et al. 2022; Harada et al. 2023), it is also possible to independently estimate the mass of the PNS. Combining these makes it possible to constrain the mass-radius relationship of NSs from neutrino observations. For instance, if the distance to an SN can be determined with 1% accuracy through optical observations, the precision in determining the NS radius would also approximate $\sim 1\%$ for a nearby SN. This represents a tenfold improvement in precision compared to the most stringent current observational limits, significantly enhancing the constraints on nuclear physics.

5. SUMMARY

In this study, we show that the quantitative analysis of neutrinos, which has recently become possible, can independently estimate the distance to a supernova explosion with an accuracy of 10%. This methodology relies on prior information about the neutron star radius derived from supplementary observations. When combined with neutrino-based determinations of the supernova's direction, this approach enables threedimensional localization, which is crucial for follow-up observations. Moreover, if the supernova's distance is further refined through electromagnetic observations, this enhanced distance accuracy can reciprocally refine parameter estimations, thereby enabling a highly precise determination of neutron star radii.

⁵ A list of nearby massive stars that may produce supernovae has been compiled in Healy et al. (2024), and matching the threedimensional positions provided by neutrinos with this list should allow us to identify the progenitor star of the supernova.

⁶ When considering parallax measurements with Gaia, bright objects can achieve distance estimates with approximately 1% precision, even at distances exceeding 1 kpc (Gaia Collaboration et al. 2023).

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