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Characterizing a supernova's standing accretion shock instability with neutrinos and gravitational waves

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We perform a novel multimessenger analysis for the identification and parameter estimation of the standing accretion shock instability (SASI) in a core-collapse supernova with neutrino and gravitational-wave (GW) signals. In the neutrino channel, this method performs a likelihood ratio test for the presence of SASI in the frequency domain. For gravitational-wave signals we process an event with a modified constrained likelihood method. Using simulated supernova signals, the properties of the Hyper-Kamiokande neutrino detector, and O3 LIGO interferometric data, we produce the two-dimensional probability function (PDF) of the SASI activity indicator and calculate the probability of detection $P_{\rm D}$ as well as the false identification probability $P_{\rm FI}$. We discuss the probability to establish the presence of the SASI as a function of the source distance in each observational channel, as well as jointly. Compared to a single-messenger approach, the joint analysis results in a $P_{\rm D}$ (at $P_{\rm FI} = 0.1$) of SASI activities that is larger by up to $\approx 40\%$ for a distance to the supernova of 5 kpc. We also discuss how accurately the frequency and duration of the SASI activity can be estimated in each channel separately. Our methodology is suitable for implementation in a realistic data analysis and a multimessenger setting.

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I. INTRODUCTION

With the recent detection of gravitational signals from binary systems, we have entered the era of multimessenger astronomy with gravitational waves (GWs) [1,2]. There is great hope that many more classes of sources will eventually be detected in GWs in the future. One such class, with great scientific potential, is core-collapse supernovae (CCSNe) [3,4].

A CCSN is a prime multimessenger source, involving GWs, neutrinos, electromagnetic signals at several wavelengths, and possibly cosmic rays. In particular, neutrinos and gravitational waves play an important role, as they carry information about the early stages in the collapse as well as the causes for shock revival, while instead electromagnetic observations mainly probe the later, postshock breakout physics [3].

In the event of a supernova in our galaxy, a detection in the neutrino channel is guaranteed, as long as neutrino detectors at or above the kiloton scale are operational [5]. The detection of GWs is possible and might be achieved at ground-based laser interferometers depending on the detailed GW morphology [6]. The physics potential of a joint detection of neutrinos and GWs from a galactic (or otherwise nearby) supernova has been explored by some pioneering works [7,8], but has not been fully studied yet.

The simulated gravitational waves from CCSNe appear as stochastic processes in the time domain but also present deterministic features in the time-frequency domain. These features include the frequency evolution of the fundamental vibrational mode (f/q - mode) of the proto-neutron star (PNS) [9–11], as well as a deterministic imprint related to the hydrodynamic instability called the standing accretion shock instability (SASI) [12–23]. The latter has a distinctive signature in neutrinos, in the form of quasiperiodic fluctuations of the neutrino luminosity, and therefore it is a natural candidate for multimessenger studies. As its name indicates, SASI is a large-scale, sloshing motion of the stalled shock front, which typically lasts for a fraction of a second postcollapse. Depending on its amplitude and duration, it could have a critical role in promoting convection and therefore aiding the shock revival mechanism, especially in more massive progenitors that are prone to collapse into a black hole [12].

The SASI descriptive parameters include the frequency (in both the GW and neutrino channels), duration, amplitude, and GW polarization state. The average SASI frequency contains information about the average radius of the stalled shock front and coupling mechanism between the shock wave and the PNS [12,24]. Longer SASI durations could also appear in failed supernovae [24].

It is of interest to analyze the statistical conditions needed to detect the presence of the SASI and to estimate its parameters for realistic detectors where noise is present. The noises, as well as the signal processing artifacts, of neutrino and GW detectors are different. However, in both the neutrino and GW channels, noise can induce some energy in the SASI time-frequency regions, thus complicating the analyses. Spectral properties of the SASI features in the neutrino luminosity were described in Refs. [25–27] for a specific set of progenitors. The question of the detectability of SASI was discussed, not with respect to a specific algorithm, but in terms of the spectral amplitude relative to a Cherenkov detector's shot noise. In Refs. [26,27], the shot noise was estimated by Fourier transforming a neutrino time series. The estimated shot noise became independent of frequency. Note, however, that when the statistical fluctuation of the neutrino signals itself dominates over the noise of the detector's background, the frequency independence assumption may only serve as a rough approximation. The extension to a full SASI detection methodology in the neutrino channel was performed in Ref. [5], where some of the current authors proposed a procedure (which we named the "SASI-meter") to detect the presence of SASI with a desired statistical confidence, as well as obtain an estimate of the frequency for SASI candidates that pass a desired confidence threshold. In that work, we also pointed out that there is an intrinsic uncertainty in the frequency of SASI in both the GW and neutrino channels related to the finite duration of the SASI episode.

In Ref. [28], an application of the Hilbert-Huang transform to a three-dimensional CCSN GW was proposed for the SASI frequency and duration estimation with simulated Gaussian noise.

In Ref. [29] a Bayesian method that uses a training process on an existing database of GWs was proposed to identify the presence of SASI. Magnetorotational emission models were assumed not to contain the SASI. Parameter estimation and the false identification probability were not involved in that study. In this paper, we further extend the theme of SASI detectability and parameter estimation in real interferometric noise as well as analyze the probability with which the presence of the SASI can be established, the intrinsic uncertainties of the SASI frequency, and the SASI false identification probability. We use frequentist inference, which does not apply prior information about the SASI from any specific numerical simulation. We use theoretical knowledge to identify conservative boundaries of the time-frequency region of a GW signal where SASI contributions would be present.

We consider a scenario where a CCSN detection has been established in both the (time-coincident) neutrino and GW channels. In this framework, we focus on the estimation of physical parameters of the SASI hydrodynamic instability from the recorded neutrino luminosity and GW signatures. We extend our recent neutrino analysis [5] with an estimate of the duration of the SASI and by including the GW channel for both the detection and the estimation of the deterministic parameters. The wavelet decomposition of the GW data, recorded at a laser interferometer like LIGO [30], is performed using the coherent WaveBurst (cWB) algorithm [31], while the postprocessing of the (simulated) data is a novel element of this work. We introduce a new quantitative metric for the detection of the SASI in the GW channel, which computes the ratio of a collective signal-to-noise ratio (SNR) of the wavelet components in the SASI time-frequency region versus the total SNR of the event.

In Sec. II we review the physical origins of SASI and its deterministic parameters, and discuss an illustrative example. In Sec. III we discuss the methodology of analyzing SASI-induced neutrino and GW modulations separately, and the results of the single-messenger analyses. In Sec. IV we present a novel methodology of jointly analyzing the neutrino and GW signatures, and the results of this analysis. We also discuss the expected precision in estimating the oscillation frequencies, amplitude, starting time, and duration of SASI-induced modulations on neutrino and GW signatures separately. For the frequency estimation, we present the results of the two channels separately [32]. The roles of the detectors and the distance to the CCSN are investigated. Finally, we discuss the application of this novel joint analysis to future CCSNe observations and conclude in Sec. V.

II. TIMING, FREQUENCY, AND DURATION OF SASI

In this section, we review the origin of the signal processing features of the SASI like the timing with respect to the initial collapse, the duration, the frequency content in the neutrino and GW channels, as well as some considerations on the amplitude and related GW polarization state (even if the current analysis does not make use of the GW polarization state).

The electron neutrino luminosity increases to its peak level over a ~1–10-ms time scale after the core bounce due to neutronization. Over the same time frame, for a nonrotating progenitor, we do not expect a relatively strong GW emission, because the collapsed core and its immediate surroundings are nearly spherically symmetric (see, for example, the GW signals in Ref. [6]). A turbulent phase is expected to start after the shock stalls, at $t \gtrsim 50$ ms postbounce, with parts of the shock collapsing under aspherical mass accretion. Such turbulence corresponds to a relatively stronger GW emission, which will be initially due to the fundamental g/f oscillatory mode of the PNS. When the accretion on the shock breaks spherical symmetry, parts of the shock are also susceptible to tangential forces that can amplify the SASI; see, for example, Ref. [12] and references within.

As an illustrative example of our methodology, we use the results of the three-dimensional general-relativistic (GR) simulation by Kuroda, Kotake, Hayama, and Takiwaki (KKHT) [33], in which SASI was found to leave an imprint in both neutrinos and GWs. Specifically, we use the numerically calculated neutrino event rate for Hyper-Kamiokande (Hyper-K) and the GW time series for the model S15.0 (SFHx) which is for a nonrotating progenitor with mass $15.0M_{\odot}$ and the equation of state SFHx. The KKHT model exhibits vigorous sloshing (as opposed to spiral) SASI motion. We obtain both the simulated neutrino and GW signals for a representative observer direction, which is generic, and not special with respect to the sloshing SASI motion. The investigation of models with fast-rotating progenitors that show spiral SASI motion will be left for future works.

A. Physical origin of SASI

Let us begin by reviewing basic analytical arguments on SASI. There is agreement in stating that the SASI period depends on the mechanism that couples the shock wave and the surface of the PNS, as well as the total mass behind the shock wave. However, more discussion is ongoing on the details as well as the best definitions for parameters like the PNS radius. Two equations related to these issues are discussed in the following. In Ref. [12], the coupling was stated to be acoustic because the advective effect is expected to operate on slower time scales. According to Refs. [21,24], SASI is due to an advective-acoustic cycle whose period is given by the sum of the advective and acoustic time scales for perturbations traveling between the (angle-averaged) shock radius $r_{\rm sh}$ and the radius of maximum deceleration r_{∇} on the surface of the protoneutron star:

$$T_{\text{SASI}} = \int_{r_{\nabla}}^{r_{\text{sh}}} \frac{dr}{|V_r|} + \int_{r_{\nabla}}^{r_{\text{sh}}} \frac{dr}{c_s - |V_r|}.$$
 (1)

Here V_r is the average radial velocity of the outgoing material within the average shock radius and c_s is the average speed of sound in the same region. Equation (1) shows that fluctuations in the shock radius will induce variations in the period of the SASI. The size of these variations is about 20% in some numerical simulations of CCSNe with direct collapse into black holes [24]. The dependence of the central frequency on the mass behind the shock has also been estimated, by either physical arguments [6,25,34,35] or mode analysis [11,36–39]. According to the fitting formulas derived in Ref. [11], the typical frequency of the GW emission associated with SASI activity [40] is given by

$$f_{\rm GW}^{\rm SASI} = 2 \times 10^2 \ {\rm Hz} \sqrt{\frac{m_{\rm sh}}{r_{\rm sh}^3}} - 8.5 \ {\rm Hz} \left(\frac{m_{\rm sh}}{r_{\rm sh}^3}\right).$$
 (2)

Here $m_{\rm sh} = M_{\rm sh}/M_{\odot}$ and $r_{\rm sh} = R_{\rm sh}/100$ km, where $M_{\rm sh}$ and $R_{\rm sh}$ are the total mass behind the shock and average shock radius, respectively. $M_{\rm sh}$ can be approximated by the PNS mass, $M_{\rm sh} \simeq M_{\rm PNS} \simeq 1.4 M_{\odot}$, because most of the mass is confined within the central object. Before shock revival, typically $R_{\rm sh} \approx 150$ km. With this choice of parameters, we find $f_{\rm GW}^{\rm SASI} \approx 125$ Hz [40]. The connection between f_{GW}^{SASI} and the actual frequency of the hydrodynamical instability is still an open question. Since each full cycle of the SASI amounts to two periods of an associated quadrupole deformation, the frequency of the GWs generated directly by the SASI, or by the resonant excitation of the PNS by the SASI, is expected to be roughly twice the frequency of the instability itself, which is also the expected frequency of the fluctuations in the neutrino luminosity, f_{ν}^{SASI} . Therefore, one might expect $f_{\text{GW}}^{\text{SASI}} \sim 2 f_{\nu}^{\text{SASI}}$. Some authors also found that, for observing directions along the SASI axis, the SASI signature in the GW channel could have two main components: one at $f \sim f_{\nu}^{\text{SASI}}$ and the other at $f \sim 2f_{\nu}^{\text{SASI}}$ [25]. In other numerical works, a simpler connection, $f_{\text{GW}}^{\text{SASI}} \sim f_{\nu}^{\text{SASI}}$, has emerged [33]. Motivated by the currently evolving discussion, here f_{GW}^{SASI} and f_{ν}^{SASI} will therefore be treated independently, without any assumption linking them.

We note (see also Ref. [5]) that the SASI frequency suffers from a minimum uncertainty given by the inverse of the duration of the SASI episode (e.g., a 100-ms SASI would have an intrinsic uncertainty of 10 Hz) because of the line broadening induced by temporal windowing. In this regard, given that the duration is the same for the neutrino and GW signatures, this fixed duration's relative impact would be smaller in the GW channel if $f_{GW}^{SASI} \sim 2f_{\nu}^{SASI}$. Given that the GW features related to the SASI are expected to develop a specific preferential rotational axis, the polarization is expected to be mainly elliptical [41]. Here, however, the polarization signatures will not be considered in detail; their inclusion is postponed to future work.

B. Illustrative example

To illustrate the methodology, we use the results of the KKHT model S15.0, for both the neutrino and GW channels. The simulated GW waveform contains both the g-mode and the SASI. In two-dimensional time-frequency maps, the g-mode has a slope of $\sim 3000 \text{ s}^{-2}$, and it starts roughly before 100 ms postbounce. The SASI episode has a duration of approximately 50 ms, and a frequency of ~ 127 Hz.

For the neutrino event rates, the signal with the SASI removed (no-SASI from here on) was generated in the way described in Ref. [5]. The original time sequence is made smoother (thus eliminating the high-frequency oscillations

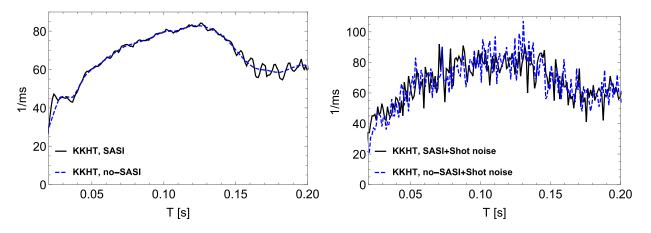


FIG. 1. Neutrino event rates with and without SASI-induced imprints predicted by KKHT [33] at Hyper-K for a supernova at distance D = 10 kpc. The curve without SASI is obtained by applying a low-pass filter on the original signal. In the left (right) panel the rates are shown without (with) the realistic Poisson fluctuations—i.e., the neutrino shot noise—which are driven by the signal itself.

due to SASI) by taking the event rates averaged over eight time bins, each of width $\Delta = 1$ ms, and performing a polynomial interpolation of these averaged rates. See Fig. 1 for the predicted original and smoothed-out neutrino event rates. The figure also shows a realistic version of the same rates where, for illustration, we include a realization of the statistical fluctuations of the number of events in each bin (neutrino shot noise).

To illustrate the GW pipeline we prepare two sets of simulated GW waveforms and corresponding neutrino event rates (number of events in time bins of 1-ms width) at the Hyper-K detector. One set includes the original KKHT output containing the SASI for neutrinos and GWs. The other is produced from the original one by artificially removing the SASI imprints. We use the two waveforms to derive the receiver operating curves (ROCs) to identify the SASI in GW data (see Sec. III). The two sets are shown in Figs. 2 and 3; a brief description of how they are prepared is given below.

For GWs, we resample the original KKHT waveform (run S15.0) to 16 384 Hz to match the standard sampling frequency of the LIGO noise and used waveform data after 57.37 ms, as the earlier part has neither g-mode nor SASI components. We apply a twofold filtering to the S15.0 waveform as follows:

(1) The S15.0 waveform is split into two segments based on time intervals: one (t < 156.25 ms, S15.0-E for Early) where the SASI is not present, and the other (t > 156.25 ms, S15.0-L for Late) where it is present. This "cutoff" was chosen particularly for this waveform (to produce the no-SASI injection). The GW SASI-meter does not include any such cut-off. Then, a band-stop Butterworth digital filter [42] is applied to the S15.0-L segment to remove frequencies from 60 to 200 Hz, thus removing the SASI (fS15.0-L segment, where f is for filtered).

(2) We rejoin the S15.0-E and fS15.0-L segments, and apply a low-pass filter to the resulting waveform to address the residual discontinuity in the junction point by removing frequency components greater than 2000 Hz. The final result is the new, filtered waveform, fS15.0.

Figure 2 shows the two polarizations with the two models in the time domain, with their corresponding spectrograms in the time-frequency domain.

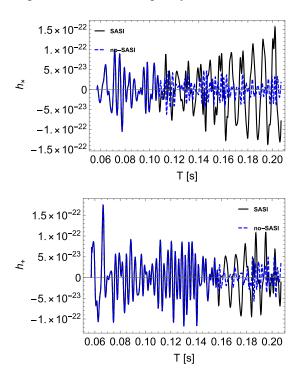


FIG. 2. The cross and plus gravitational-wave polarizations of the KKHT model [33] are plotted versus time (black solid). We also plot the same quantities where a time-frequency filter is applied to remove the GW SASI component according to the discussion in Sec. II B (blue dashed).

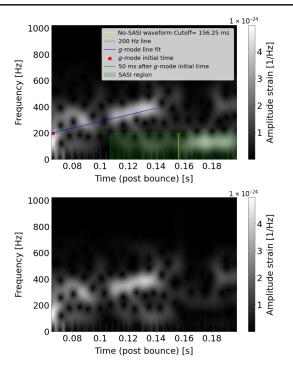


FIG. 3. S15 model spectrogram for the plus polarization, with SASI (top) and without SASI (bottom). In the top spectrogram, the blue line indicates where the *g*-mode is located (with the red dot indicating the time that the *g*-mode is around 200 Hz). In green is the allowed time-frequency SASI region (to start after 50 ms of the *g*-mode initial time) and below the 200 Hz line. In yellow we indicate the temporal interval where the actual SASI oscillations (see the location of the SASI in the left plot of Fig. 1 and bottom plot of Fig. 2) are removed, after the yellow vertical line at 156.25 ms.

Although the methodology of this paper is illustrated for the results of a specific numerical simulation, we designed it to be applicable to the range of three-dimensional simulations existing in the literature, which is summarized below for the interested reader.

In Ref. [13], results from a three-dimensional simulation of GW emission for a $15M_{\odot}$ star were presented. From the time-frequency plots, we can see that after the quiescent phase, which lasts until $t \approx 100$ ms postbounce, a lowfrequency signal, with frequencies below ≈ 200 Hz begins, after ≈ 50 ms of the g-mode component corresponding to 200 Hz, which persists through the remainder of the simulation (until ≈ 425 ms postbounce), which is due to aspherical mass motions in the gain layer from neutrinodriven convection and the SASI. Also, in the interval $\sim 150-200$ ms postbounce, an intermediate frequency emission ($\approx 400-600$ Hz) joins the low-frequency emission which is described as SASI-induced aspherical accretion.

In Ref. [33], three-dimensional GR with ν transport simulation was conducted for three nonrotating progenitors of 11.2, 15, and $40M_{\odot}$. Prompt convection can be seen in both low- and high-frequency regions in the early stage

until $t \approx 50$ ms postbounce of the simulation, with S11.2 showing the strongest prompt convection. In addition to the PNS *g*-mode, which is a relatively narrow-banded spectrum that can be seen for all models, in S15.0(SFHx) a SASIinduced low-frequency component is seen in the 100 < f < 150 Hz band for t > 150 ms postbounce, which is ≈ 75 ms after the *g*-mode component.

In Ref. [43] four models were discussed: s11.2, s20, s20s, and s27. In s11.2, no SASI growth is observed because of the large shock radius. The other models are SASI dominated. In s20, strong SASI activity (dominated by the spiral mode) is seen from 120-280 ms after core bounce which is the extended phase with a peak from 200-250 ms. After a period of transient shock expansion, SASI (much weaker) continues. In s20s, prior to shock revival, postshock flow is dominated by large-scale SASI sloshing motions between 120-280 ms post bounce. In s27, two episodes of pronounced SASI activity can be seen interrupted by a phase of transient shock expansion following infall of the Si/O interface. The first phase is from 120-260 ms postbounce and the second phase is from 410 ms postbounce to the end of the simulation (575 ms postbounce). Also, a low-frequency emission between 280-350 ms postbounce can be seen which is not associated with SASI.

SASI produces a sizable l = 1 mode in the range 50–100 Hz, and l = 2 components in 100–200 Hz range. From the literature, we see that noticeable SASI components begin after a delay of \approx 50 ms from the *g*-mode component corresponding to 200 Hz.

III. SINGLE-MESSENGER ANALYSIS: METHOD

We now illustrate the statistical method of analysis for each messenger separately. The flowchart in Fig. 4 provides a compact summary of the single-messenger analysis procedure, and its relationship with the multimessenger analysis, which will be discussed in Sec. IV.

We assume that the noises in the neutrino and GW channels are statistically independent. The main reason is that for the neutrino channel the noise is partly driven by the neutrino luminosity itself in the search for oscillatory signals in detected neutrino events, while in laser interferometers the noise is instrumental in origin.

The signals in the GW and neutrino channels have different dependences on the distance to the CCSN, D. The amplitude of the GW signal at an interferometer scales like D^{-1} , whereas the event rate in a neutrino detector is proportional to D^{-2} . This means that the dependence on D of the SASI meter, as well as the SASI parameter estimation, could be different in the neutrino and GW channels, resulting in a nontrivial dependence of the combination of the two channels on the distance, which needs to be properly investigated; see Sec. IV.

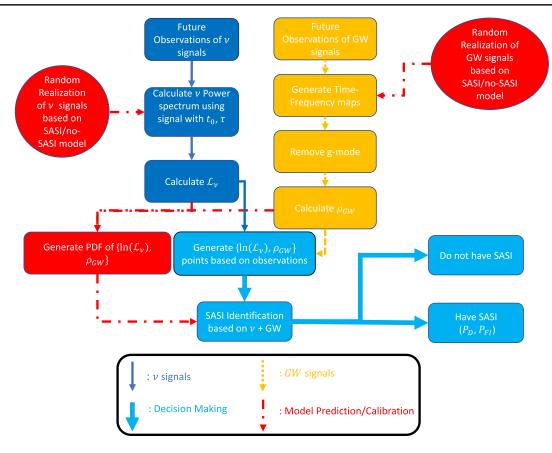


FIG. 4. Flowchart of the multimessenger SASI meter developed in this work. Due to the absence of current detections, we characterize the pipeline with random realizations of reconstructed ν and GW signals for the test example. For the ν , this means adding Poissonian fluctuations on signals that have a SASI as well as signals where the SASI was removed. In the case of future detections, we can achieve the same result by taking a smoothed-out version of the detected neutrino luminosities as the no-SASI ν signature, and randomize it with Poissonian fluctuations to identify the threshold for the desired $P_{\rm FI}$ used as a reference in this work for the single-channel or multiplechannel identification mode ($P_{\rm FI} = 0.1$). For the GW channel, in the test example used in this paper, we inject the GWs with and without SASI in real interferometric noise. In a realistic scenario, the no-SASI injections can be used to tune the threshold on the identification metric for the desired single or multimessenger $P_{\rm FI}$ ($P_{\rm FI} = 0.1$). The signals in the ν and GW channels with SASI are used in this paper to characterize the performance of the GW- ν SASI meter.

A. Neutrino-only analysis

In the present context, a neutrino detector like Hyper-K operates as a counting device, so the main observable is the number N of neutrinos detected ("events") in the detector's volume in a given time bin. N is a function of time, and it is affected by the physical fluctuations of the incoming neutrino flux (some due to SASI) as well as statistical (Poissonian) fluctuations. The latter scale as \sqrt{N} , so their relative amplitude increases with the distance: $\sqrt{N}/N \propto D$.

In the following we discuss our likelihood ratio methodology for identifying SASI-induced neutrino signals in detail. This methodology is inspired by the Neyman-Pearson detection method of a signal in Gaussian noise, and provides the probability to detect the SASI as a function of the corresponding false alarm probability. The method was described extensively in Ref. [5]; for completeness, here we briefly summarized its main points.

1. Computing receiver operating curves

First, we establish two parametric templates in the time domain which characterize the main features of neutrino signals with and without the SASI activity. For the case with SASI activity we choose a single frequency function:

$$N_{\rm S}(t) = (A - n)(1 + a\sin(2\pi f_{\rm S} t)) + n, \qquad (3)$$

where N_S is the number of neutrino events collected in a unit time bin centered at t, A is the time-averaged event rate (the "DC component") in the detector including instrumental noise (after possible experimental cuts), a is the relative SASI amplitude, n is the mean value of the background events ($n \simeq 0$ for Hyper-K), and f_S is the nominal frequency of the SASI. The second template, for the case without SASI, is a constant:

$$N_{\rm nS}(t) = A,\tag{4}$$

[with A having the same meaning as in Eq. (3)]. In the above templates, f_s and a are treated as free parameters, with respect to which the likelihood will be maximized. We note that there are two implicit variables in Eqs. (3) and (4): the starting time t_0 and the duration τ of the neutrino time series of interest (which is typically a subset of the entire neutrino burst, where SASI is likely to be found). We consider these as fixed for the time being, as was done in Ref. [5]; varying them is discussed later in Appendix A. The quantity A is treated as fixed as well, because it can be determined accurately by measuring the total number of neutrino events collected in a time series with t_0 and τ [5].

We consider the neutrino events that are recorded in a detector after an initial time t_0 , in time bins of width $\Delta = 1$ ms. The *j*th time bin then corresponds to the time $t_j = t_0 + j\Delta$. The observed number of events in the same bin will then be $N(t_j)$. The spirit of the method consists in establishing how well the time series $\{N(t_j)\}$ matches the templates in Eqs. (3) and (4). Considering the oscillatory character of SASI, the matching is done in frequency space.

Following Refs. [26,27], we perform a discrete Fourier transform of the series $\{N(t_j)\}$ over the time interval $[t_0, t_0 + \tau]$, containing $N_{\text{bins}} = \tau/\Delta$ time bins. The discrete frequency resolution is then

$$\delta = \frac{1}{\tau},\tag{5}$$

which represents the minimum width of frequency bins for which statistical independence between adjacent bins can be realized. The Nyquist frequency is

$$f_{\rm Nyq} = \frac{1}{2\Delta},\tag{6}$$

which corresponds to the frequency index

$$k_{\rm Nyq} = \frac{f_{\rm Nyq}}{\delta} = \frac{\tau}{2\Delta} = \frac{1}{2}N_{\rm bins}.$$
 (7)

We define the discrete Fourier-transformed neutrino signal as

$$h(k\delta) = \sum_{j=0}^{N_{\text{bins}}-1} N(t_j) e^{i2\pi j\Delta k\delta}$$
(8)

and the one-sided power spectrum as

$$P(k\delta) = \begin{cases} 2|h(k\delta)|^2/N_{\text{bins}}^2 & \text{for } 0 < k\delta < f_{\text{Nyq}}, \\ |h(k\delta)|^2/N_{\text{bins}}^2 & \text{for } k\delta = 0. \end{cases}$$
(9)

The factor of $1/N_{\text{bins}}^2$ is included in order to fix the normalization, so that at k = 0 we have $P(0) = (N_{\text{ev}}/N_{\text{bins}})^2$ [here $N_{\text{ev}} = \sum_{j=0}^{N_{\text{bins}}-1} N(t_j)$].

For a given quantity, a symbol with a tilde will be used when referring to an actual outcome of a measurement, which is affected by statistical fluctuations. The probability that an observed power at a specific frequency $k\delta$, $\tilde{P}(k\delta)$, is a realization of a certain hypothetical template with parameters set Ω is [5]

$$\operatorname{Prob}(\tilde{P}, \Omega) = \frac{N_{\text{bins}}^2}{4\sigma^2} \exp\left[-\frac{N_{\text{bins}}^2}{4\sigma^2}(\tilde{P}+P)\right] \times I_0\left(\frac{N_{\text{bins}}^2}{2\sigma^2}\sqrt{\tilde{P}P}\right), \quad (10)$$

where I_0 is the modified Bessel function of the first kind, P is the power calculated by the template Ω at frequency $k\delta$, and

$$\sigma^2 = \frac{N_{\rm ev}}{2}.\tag{11}$$

Let us now define the likelihood that a given observed power series vector, $\tilde{\mathcal{P}} = \{\tilde{P}(k\delta)\}$, is a realization of a certain hypothetical template Ω . It is defined as

$$L(\tilde{\mathcal{P}}, \Omega) = \prod_{k=K_{\min}}^{K_{\max}} \operatorname{Prob}(\tilde{P}(k\delta), \Omega), \qquad (12)$$

where K_{\min} and K_{\max} represent the minimum and maximum frequencies, with typical values being $K_{\min}\delta \approx 50$ and $K_{\max}\delta \approx 150$ Hz, which covers the SASI frequencies predicted by state-of-the-art simulations [13,25,33].

Given two hypotheses (i.e., two templates) with parameter sets Ω_S and Ω_{nS} , and an observed set $\tilde{\mathcal{P}}$, the likelihood ratio is

$$\mathcal{L}(\tilde{\mathcal{P}}) = \frac{\operatorname{Max}_{\Omega_{S}}[L(\tilde{\mathcal{P}}, \Omega_{S})]}{\operatorname{Max}_{\Omega_{mS}}[L(\tilde{\mathcal{P}}, \Omega_{nS})]}.$$
(13)

In the numerator (denominator), the template corresponding to SASI (no-SASI) is used and $\operatorname{Max}_{\Omega_{S(nS)}}[L(\tilde{\mathcal{P}}, \Omega_{S(nS)})]$ is the maximized likelihood with extremal parameters Ω_{S} (Ω_{nS}) with respect to the observed power spectrum $\tilde{\mathcal{P}}$. In this work, the templates in Eqs. (3) and (4) will be used as representative of the SASI and no-SASI cases. Their parameters are $\Omega_{S} = \{a, f_{S}\}$ and $\Omega_{nS} = \{\text{Null}\}$, respectively.

It is intuitive to see how the likelihood ratio in Eq. (13) is sensitive to SASI. Since our templates N_S [Eq. (3)] and N_{nS} [Eq. (4)] capture well the main features of the neutrino event rates of the models with and without SASI, respectively, as the SASI features in the data become more pronounced, the numerator (13) is likely to increase, while at the same time the denominator is likely to decrease (poorer fit for the N_{nS} template), so \mathcal{L} is likely to increase. Vice versa, \mathcal{L} will take lower values if the SASI signatures in the data become weaker. Therefore, Eq. (13) serves as our "SASI-meter" to identify the presence of SASI.

To assess the robustness of the SASI-meter method, we find the probability density distributions of \mathcal{L} (or, equivalently, $\ln \mathcal{L}$) under the two hypotheses as a test bed. This is done by simulating (using a Monte Carlo method) 10^3 sets of \mathcal{P} with shot noise based on the KKHT models with and without SASI.

We then obtain the probability density distributions of \mathcal{L} in the two scenarios: $\operatorname{Prob}_{\nu,S}(\ln(\mathcal{L}))$ and $\operatorname{Prob}_{\nu,nS}(\ln(\mathcal{L}))$. A useful way to describe these two distributions, and compare them with one another, is to examine the probabilities that—under the two hypotheses—the likelihood ratio exceeds a certain threshold value, Λ_{ν} :

$$P_{\rm D}^{\nu} = \int_{\Lambda_{\nu}}^{\infty} d\ln(\mathcal{L}) \operatorname{Prob}_{\nu,\rm S}(\ln(\mathcal{L})), \qquad (14)$$

$$P_{\rm FI}^{\nu} = \int_{\Lambda_{\nu}}^{\infty} d\ln(\mathcal{L}) \operatorname{Prob}_{\nu,\rm nS}(\ln(\mathcal{L})).$$
(15)

 Λ_{ν} represents a value of the likelihood ratio above which the SASI hypothesis is accepted as true ("detection"). Therefore, $P_{\rm D}^{\nu}$ takes the meaning of the SASI *detection probability*, because it represents the probability that the method accepts the SASI hypothesis as true when the SASI is in fact true. $P_{\rm FI}^{\nu}$ then represents the *false identification* probability, i.e., the probability that the SASI hypothesis is accepted when in fact the no-SASI hypothesis is the true one. The curve of the points $(P_D^{\nu}, P_{FI}^{\nu})$ for varying Λ_{ν} is the ROC, which allows us to evaluate the effectiveness of the method at a glance.

In addition to assessing the detectability of SASI, our SASI-meter can also be used for parameter estimation. Indeed, in every realization of the Monte Carlo simulations based on the KKHT model, the extremal parameters $\tilde{\Omega}_{SASI}$ are found when searching for the maximized $\tilde{L}(\mathcal{P}, \tilde{\Omega}_{SASI})$. Thus, the probability distribution of $\tilde{\Omega}_{SASI}$ is sampled by our Monte Carlo simulations. From it, one can find the uncertainty on the parameters f_s and a due to the statistical fluctuations of neutrino events in the detector. These uncertainties increase as the distance of the CCSNe to the Earth increases, as a result of the decreased number of neutrino events [5].

2. Results: Neutrino receiver operating curves

Our results for the neutrino-only analysis are shown in Figs. 5 and 6. In particular, Fig. 5 shows the statistical distributions of $\ln(\hat{\mathcal{L}})$ for the SASI and no-SASI cases, for different values of the distance *D*. The probability density function (PDF) of $\ln(\hat{\mathcal{L}})$ with its maximum in the SASI case being one is obtained by rescaling the distribution of $\ln(\mathcal{L})$. The corresponding ROCs are shown as well. We notice the expected trends (see Sec. III A 1), namely,

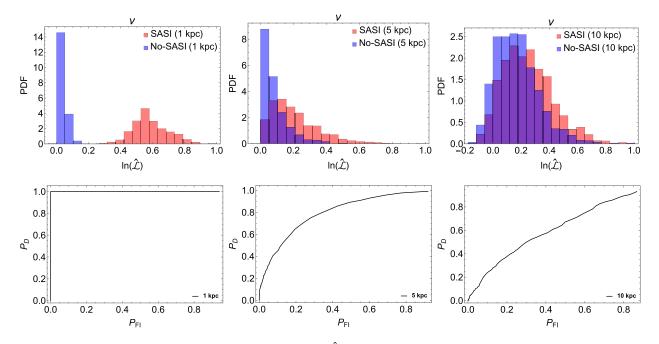


FIG. 5. Top row: examples of distributions of the test statistics $\ln(\hat{\mathcal{L}})$ obtained from Monte Carlo–generated neutrino data at Hyper-K with the starting time $t_0 = 150$ ms and duration $\tau = 50$ ms. The blue histograms correspond to the one from a simulation where the SASI component has been previously filtered out and the red histograms correspond to the one from a simulation where the SASI component has been kept. Here, $\ln(\hat{\mathcal{L}})$ is the rescaled logarithmic likelihood ratio with its maximum in the SASI case being 1. The three columns correspond to distances D = 1, 5, 10 kpc to the supernova. Bottom row: corresponding receiver operating curves describing the identification probability versus the false identification probability.

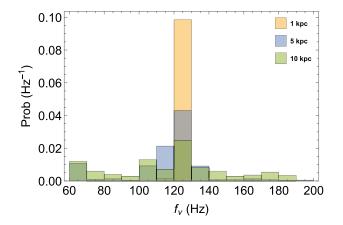


FIG. 6. Histograms of the estimated SASI frequency from neutrino signatures. As expected, the variance decreases for closer distances since the amplitude of the Poissonian fluctuations decreases with closer distances, while the amplitude of the SASI fluctuations with respect to the DC component is distance independent.

the two distributions have an increasingly large overlap as D increases, which results in a worsening (i.e., approaching the line $P_D^{\nu} = P_{\rm Fl}^{\nu}$) of the ROCs. For example, at 5 kpc, $P_D^{\nu} \approx 0.5$ (for $P_{\rm Fl}^{\nu} = 0.1$). At 10 kpc, $P_D \approx 0.2$, indicating that it would be difficult to identify SASI in Hyper-K. The methodology introduced here is model independent, namely, the identification thresholds for $P_{\rm Fl}^{\nu} = 0.1$ are determined automatically from the smoothed detected luminosity. However, the performance will be model dependent from the amplitude of the neutrino luminosity fluctuations and the mean neutrino luminosity.

We then discuss SASI parameter estimation using the neutrino SASI-meter. The estimation of parameters characterizing the SASI starting time and duration is discussed in Appendix A. We estimate the SASI frequency (oscillation amplitude) in the KKHT model by calculating the mean and uncertainties of the extremal frequency (extremal oscillation amplitude) in the time interval where the monochromatic feature is observed (see Appendix A for a more detailed discussion of the *monochromatic* feature). We plot the probability density distribution of the estimated SASI frequency at various CCSNe distances in Fig. 6. We further summarize the estimated SASI frequency and oscillation amplitude of the KKHT model (with uncertainties) at various CCSNe distance in Table I. To conclude, the extremal frequencies in the monochromatic region indicate the SASI frequency (oscillation amplitude) $f_{\nu} \approx 120 \text{ Hz}$ $(a_{\nu} \approx 0.05)$ when the neutrino signals are simulated using the KKHT model.

B. Gravitational wave–only analysis

In this section, we discuss the GW SASI identification ROC and estimate some of its parameters. We assume the

TABLE I. Estimated mean (median for GW SASI duration) and standard deviation of the SASI parameters in the neutrino and GW data analysis and the *g*-mode slope in the GW data analysis. The SASI frequency f_{ν} and amplitude a_{ν} in the neutrino analysis are estimated using neutrino events with starting time $t_0 =$ 150 ms and duration $\tau = 50$ ms, where P_D^{ν} is maximized at different CCSNe distances. The estimated SASI starting time t_0^{ν} (t_0^{GW}) and duration $\tau^{\nu}(\tau^{\text{GW}})$ in the neutrino (GW) analysis are also provided. See Appendix A for a detailed discussion of determining t_0^{ν} and τ^{ν} using neutrino signals. See Sec. III for a detailed discussion of determining τ^{GW} using GW signals. The reasons for the nonmonotonic behavior of the duration estimation with the distance are discussed in Sec. V.

g-mode slope	SASI	10 kpc	5 kpc	1 kpc
	f_{ν} (Hz)	113.38	111.03	119.85
	δf_{ν} (Hz)	32.9	22.6	1.22
	a_{ν}	0.063	0.047	0.044
	δa_{ν}	0.022	0.013	0.005
	$f_{\rm GW}$ (Hz)	120.08	120.42	122.36
	$\delta f_{\rm GW}$ (Hz)	18.65	13.80	5.48
	t_0^{ν} (ms)	N/A (due to	>150	>150
	0	large δf_{ν})		
	τ^{ν} (ms)	N/A (due to	>50	>50
		large δf_{ν})		
	$\tau^{\rm GW}$ (ms)	259	494	166
	$\delta \tau^{\rm GW}$ (ms)	347	552	261
$m_{\rm opt}^{\rm GW}(s^{-2})$		2564.84	2645.02	3190.68
$\delta m_{\rm opt}^{\rm GW}(s^{-2})$		1301.08	1132.72	929.62

existence of a detected GW event from the CCSN of interest in coincidence with a neutrino detection.

We restrict the analysis to the frequency range $16 \le f \le 2000$ Hz, and assume SASI starting times later than 50 ms after the beginning of the *g*-mode-related turbulence in the GW analysis, based on the literature review performed at the end of Sec. II B.

1. Receiver operating curves and parameter estimation

In detection and identification problems, a very important aspect is the choice of the metric to be used as an identification tool. The metric we suggest here allows us to leverage the relative importance of the recorded energy in the time-frequency region of the SASI with respect to the overall energy of the candidate. Two visualizations that illustrate the time-frequency layout of the events (in particular, the location of the SASI and g-mode) are presented in Fig. 7. One is an example of the standard cWB scalogram and the other is the equivalent pixelization adopted for the estimates of the identification metrics in this paper. The difference in the scaling of the likelihood values is because of the presence of overlapping pixels in the left plot, whose likelihood values (which are all positive) add up, causing the upper limit of its likelihood color bar to be higher than the one in the right plot. In the second plot we

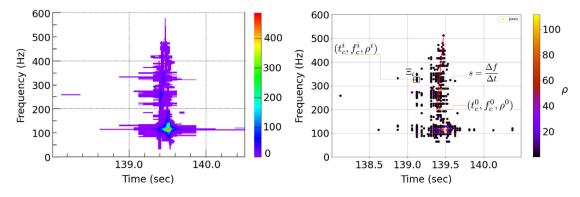


FIG. 7. Left: example of a standard cWB scalogram for a GW event. Each wavelet component involved in the reconstructed event is represented with a rectangle of sides equal to the frequency and temporal resolution. The darker blue regions highlight the reconstructed SASI and *g*-mode features. The duration resolution of tens of milliseconds of some of the wavelets below 200 Hz indicates that some impact on the duration reconstruction is to be expected. Right: the same event is displayed with a single dot for each wavelet, with the coordinates representing the central time and frequency of the wavelets. While the color bar to the right corresponds to the single-pixel likelihood, the color bar to the left (not used in the calculations of this paper) corresponds to the likelihood for a fixed-size timefrequency region. For each wavelet, we use/define different parameters for the χ^2 localization of the *g*-mode frequency evolution (red) with slope ($s = \frac{\Delta f}{\Delta t}$): the wavelet central time (t_c), wavelet central frequency (f_c), wavelet likelihood (ρ), and density of wavelets (Ξ) in a given time-frequency box. For details on these parameters and their usage, see Appendix B.

also illustrate the relevant parameters used in each pixel (pixel central time, pixel central frequency, and pixel SNR) as well as collective measures like the local density of pixels, estimated *g*-mode start time, and slope.

The performance of applying thresholds on a certain metric should be tested on a large number of test cases. Here we produce the test cases by injecting the simulated gravitational waves into different instances of real laser interferometric noise from the LIGO O3 scientific run. In each case, the cWB algorithm was used to process the data according to the configuration described in Appendix B. The normalized likelihood for a given trigger is defined as the sum of the likelihood values of all pixels surviving in the SASI region (see Appendix B) with respect to the total sum of likelihoods in the trigger:

$$\rho_{\text{norm}} = \frac{\sum_{i \in \text{SASI}} \rho^i}{\sum_{j \in \text{trigger}} \rho^j},$$
(16)

where the *i*th pixel \in SASI region and the *j*th pixel \in whole trigger. If there are no pixels in the SASI region, the normalized likelihood of the trigger is 0. In this case our recommendation is not to use the gravitational-wave data but instead use only the neutrino SASI-meter for the identification and parameter estimation of the SASI. In this analysis, the situation with no pixels in the SASI-region occurs with probability of $\lesssim 1\%$ at 1 Kpc and $\lesssim 20\%$ at 10 Kpc.

We prepare distributions of the normalized likelihood by repeating the analysis for injections at different times in the noise. The fraction of events above a given threshold can be considered as an estimate of the probability that the normalized likelihood is above the threshold. The detection probability ($P_{\rm D}^{\rm GW}$) and false identification probability

 $(P_{\rm FI}^{\rm GW})$ are calculated from the PDF of $\rho_{\rm norm}$ based on the simulated GW waveforms with and without SASI activities. $P_{\rm D}^{\rm GW}$ ($P_{\rm FI}^{\rm GW}$) is the ratio of the cumulative area under the SASI (no-SASI) PDF curve of $\rho_{\rm norm}$ with $\rho_{\rm norm} > \Lambda_{\rm GW}$ to the total area under the SASI (no-SASI) PDF curve (see the three distributions in the top row of Fig. 8). The ROC is the plot of $P_{\rm D}^{\rm GW}$ as a function of $P_{\rm FI}^{\rm GW}$, with varying $\Lambda_{\rm GW}$:

$$P_{\rm D}^{\rm GW} = \int_{\Lambda_{\rm GW}}^{\infty} d\rho \operatorname{Prob}_{\rm GW,S}(\rho), \qquad (17)$$

$$P_{\rm FI}^{\rm GW} = \int_{\Lambda_{\rm GW}}^{\infty} d\rho {\rm Prob}_{\rm GW,nS}(\rho). \tag{18}$$

The ROC with $P_{\rm D}^{\rm GW} = P_{\rm FI}^{\rm GW}$ corresponds to the 50/50 classification/detection scenario, which is equivalent to flipping a coin. The operating point depends on the maximum $P_{\rm FI}$ that we would like to use. Similarly to Ref. [5], we take $P_{\rm FI}^{\rm GW} = 0.1$.

The two SASI parameters that we calculate for each trigger are its central frequency and duration. The central frequency of the SASI, f_{GW} , is estimated as the weighted mean of the frequencies of the pixels belonging to the SASI region (f_c^i) in the time-frequency pixel map presented in Fig. 7. The weights are given by the likelihoods of corresponding pixels (ρ^i) , where the *i*th pixel \in SASI region. f_{GW} is

$$f_{\rm GW} = \frac{\sum_{i \in {\rm SASI}} \rho^i f_c^i}{\sum_{i \in {\rm SASI}} \rho^i},\tag{19}$$

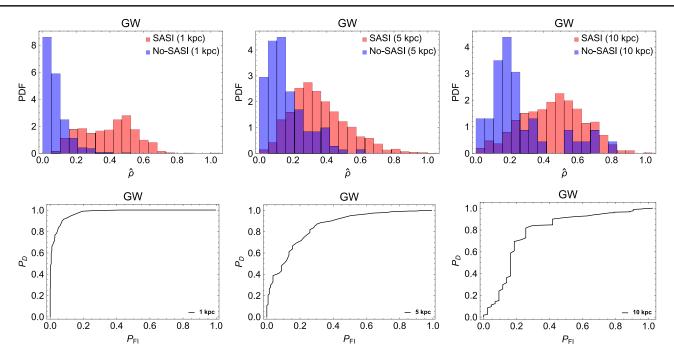


FIG. 8. Top row: examples of distributions of the test-statistics $\hat{\rho}$ obtained from simulated GW data, for distances D = 1, 5, 10 kpc to the supernova. Here, $\hat{\rho}$ is the rescaled logarithmic likelihood ratio of GW signals with its maximum in the SASI case being 1. Bottom row: corresponding receiver operating curves.

where the summation is done over all pixels remaining in the area of interest of the trigger (SASI region) and ρ^i is the likelihood value of the *i*th pixel.

The GW duration of the SASI (τ^{GW}) is estimated here by the difference in time coordinates of the two extreme pixels in the SASI region. Since the pixels correspond to wavelet components from multiresolution, the time resolutions of the two extreme pixels are also used in the estimate according to

$$\tau^{\rm GW} = t_{\rm max} + \frac{\delta t_{\rm max}}{2} - t_{\rm min} - \frac{\delta t_{\rm min}}{2}, \qquad (20)$$

where t_{max} and t_{min} are the time coordinates of the rightmost pixel (with δt_{max} as its time resolution) and the leftmost pixel (with δt_{min} as its time resolution), respectively.

2. Results: GW receiver operating curves

Figure 8 displays the PDFs as well as the ROCs of the test-statistics $\hat{\rho}$ for the SASI and no-SASI cases, for different values of CCSNe distance. The PDF of $\hat{\rho}$ with its maximum in the SASI case being one is obtained by rescaling the distribution of ρ_{norm} . As expected, the value of P_{D}^{GW} (for fixed $P_{\text{FI}}^{\text{GW}}$) decreases with increasing distance. The decline is noticeably slower than the one observed in the neutrino channel, reflecting the slower scaling of the GW signal with *D*. For D = 10 kpc and $P_{\text{FI}}^{\text{GW}} = 0.20$, we have $P_{\text{D}}^{\text{GW}} \sim 0.60-0.65$, which is larger than the corresponding neutrino result ($P_{\text{D}}^{\nu} \simeq 0.35$ for $P_{\text{FI}}^{\nu} = 0.20$).

In Fig. 9 and Table I, we present the parameter estimation results. Figure 9 shows the probability density distribution of the SASI starting time as well as the SASI duration in GW signals at various CCSNe distances.

In Table I we show the estimated values and uncertainties for the frequency $f_{\rm GW}$ and duration $\tau^{\rm GW}$ of the SASI. The results for $f_{\rm GW}$ can be directly compared to those obtained from the neutrino-only analysis (also shown in the table). Here we notice the different dependence on *D* in the neutrino and GW results: for D = 1 kpc the uncertainties are comparable in the two channels, with the performance being slightly better in the neutrino channel; however, as the distance grows, the performances in the two channels decrease at different rates.

The parameter estimation performance is poorer for τ^{GW} , for which the uncertainty is comparable or larger than the central value [44]. This result could be a limitation of our definition of duration. In the future, we might instead employ a definition based on an SNR-weighted duration.

The results discussed here show the possibility to identify the presence of the SASI for a galactic CCSN candidate at current interferometers. Our results can be used to forecast the performance of future interferometers that will have order-of-magnitude better sensitivity. Because GW signal amplitudes scale as D^{-1} , these detectors are expected to have ROCs similar to those shown here for larger distances (proportional to their sensitivity).

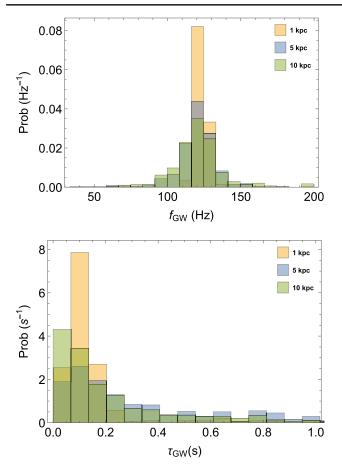


FIG. 9. Probability density distribution of the SASI central frequency (upper) and SASI duration (lower) estimations based on GW signals.

IV. MULTIMESSENGER ANALYSIS

In this section we discuss a novel methodology analyzing the SASI using *both* the neutrino and GW signals combined. To begin with, we present a flow chart illustrating the main procedure; see Fig. 4. In the following, the generation of combined receiver operating curves for SASI identification is discussed.

A. Combining probabilities

The identification of SASI activity can happen in two scenarios, where (usable) data are available either (i) only in a single messenger (either neutrino or GW signatures) or (ii) in multiple messengers. Each scenario is expected to happen with a specific probability. Realistically, the detection probability of neutrino signals from a galactic or nearby extragalactic CCSN is unity (when the detectors are active), and the neutrino data can always be used for the detection of the SASI activity. The probability of having usable data in the GW channel, P_{GW}^{D} , depends on the probability to detect the GW waveform itself (for a recent

discussion, see Ref. [45]) and the probability that the GW presents surviving wavelet components in the SASI time-frequency region.

In the case where only neutrino data are available, the detection probability of SASI activities is $P_{\rm D} = P_{\rm D}^{\nu}$. Similarly, $P_{\rm FI} = P_{\rm FI}^{\nu}$. In the second case, where both usable neutrino and GW data are present, we define joint detection and false identification probabilities as follows.

We can define the two-dimensional probability density distributions for the SASI and no-SASI cases as

$$\begin{aligned} &\operatorname{Prob}_{S}(\ln(\mathcal{L}),\rho) = \operatorname{Prob}_{\nu,S}(\ln(\mathcal{L}))\operatorname{Prob}_{GW,S}(\rho), \\ &\operatorname{Prob}_{nS}(\ln(\mathcal{L}),\rho) = \operatorname{Prob}_{\nu,nS}(\ln(\mathcal{L}))\operatorname{Prob}_{GW,nS}(\rho). \end{aligned} \tag{21}$$

Similarly to the single-messenger case, this distribution can be integrated to obtain detection and false identification probabilities. In this case, however, the integration threshold for SASI identification can be chosen in more than one way. The choice is driven by the goal of finding the optimal ROC curve (i.e., maximize the joint detection probability for a fixed false identification probability) for the identification of the SASI.

For example the identification of the presence of the SASI can be pursued in the three following ways:

(1) Logical And: SASI is established when $\mathcal{L} > \Lambda_{\nu}$ and $\rho > \Lambda_{GW}$. The combined probability of detection and the combined false identification probabilities are

$$P_{\rm D}^{\rm comb} = P_{\rm D}^{\nu} \times P_{\rm D}^{\rm GW}$$

= $\int_{\Lambda_{\nu}}^{\infty} \int_{\Lambda_{\rm GW}}^{\infty} d\ln(\mathcal{L}) d\rho \operatorname{Prob}_{\nu,\rm S}(\ln(\mathcal{L}))$
 $\times \operatorname{Prob}_{\rm GW,\rm S}(\rho)$ (22)

and

$$P_{\rm FI}^{\rm comb} = P_{\rm FI}^{\nu} \times P_{\rm FI}^{\rm GW}$$

= $\int_{\Lambda_{\nu}}^{\infty} \int_{\Lambda_{\rm GW}}^{\infty} d\ln(\mathcal{L}) d\rho \operatorname{Prob}_{\nu,\mathrm{nS}}(\ln(\mathcal{L}))$
 $\times \operatorname{Prob}_{\rm GW,\mathrm{nS}}(\rho).$ (23)

In a two-dimensional probability density distribution of (\mathcal{L}, ρ) , the SASI threshold defined above appears like a rectangle (see illustration in Fig. 10), with its accepted GW component lying between Λ_{GW} and ∞ and its accepted neutrino component lying between Λ_{ν} and ∞ . Only the (\mathcal{L}, ρ) lying inside of the defined rectangle would be identified as a detection of the SASI.

(2) Logical or: SASI is established when at least one of $\mathcal{L} > \Lambda_{\nu}$ or $\rho > \Lambda_{GW}$ is verified.

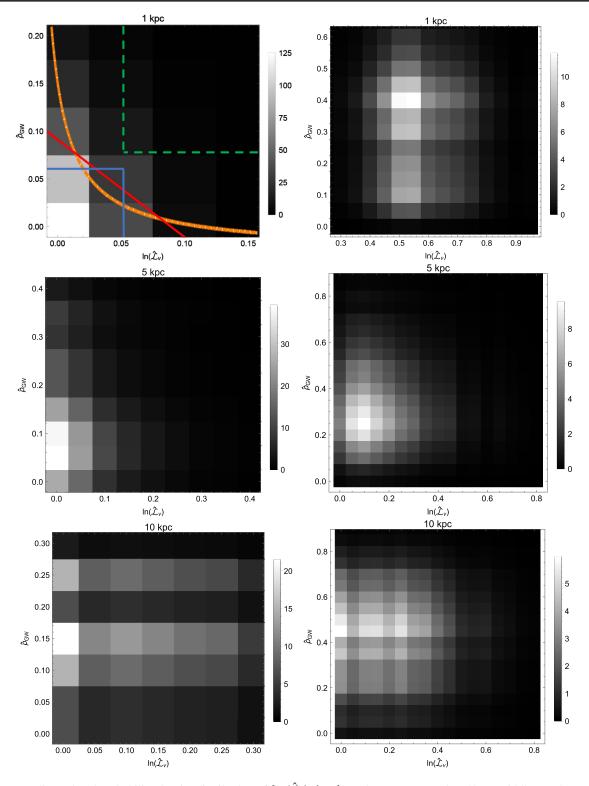


FIG. 10. Two-dimensional probability density distribution of $\{\ln(\hat{\mathcal{L}}_{\nu}), \hat{\rho}_{GW}\}$ at 1 kpc (upper panels), 5 kpc (middle panels), and 10 kpc (lower panels), where the distributions in the left (right) panels are based on neutrino and GW signals without (with) SASI activities. Here, $\ln(\hat{\mathcal{L}})$ is the rescaled logarithmic likelihood ratio with its maximum in the SASI case being 1, and $\hat{\rho}$ is the rescaled logarithmic likelihood ratio with its maximum in the SASI case being 1. In the upper left panel, the colored curves illustrate the different integration thresholds that can be used to identify the presence of the SASI, as discussed in Sec. IVA. Specifically, the region inside the green dashed rectangle is for "Logical And" case, and the region outside the blue solid rectangle is for "Logical or" case. The region outside the red solid triangle corresponds to the "x + y = const" case, and the region to the upper right of the orange dashed curve represents the " $x \times y = \text{const}$ " case.

The combined probability of detection and the combined false identification rate are then given by

$$P_{\rm D}^{\rm comb} = 1 - (1 - P_{\rm D}^{\nu}) \times (1 - P_{\rm D}^{\rm GW})$$
$$= 1 - \int_{0}^{\Lambda_{\nu}} \int_{0}^{\Lambda_{\rm GW}} d\ln(\mathcal{L}) d\rho \operatorname{Prob}_{\nu,S}(\ln(\mathcal{L}))$$
$$\times \operatorname{Prob}_{\rm GW,S}(\rho) \tag{24}$$

and

$$P_{\mathrm{FI}}^{\mathrm{comb}} = 1 - (1 - P_{\mathrm{FI}}^{\nu}) \times (1 - P_{\mathrm{FI}}^{\mathrm{GW}})$$

= $1 - \int_{0}^{\Lambda_{\nu}} \int_{0}^{\Lambda_{\mathrm{GW}}} d\ln(\mathcal{L}) d\rho \mathrm{Prob}_{\nu,\mathrm{nS}}(\ln(\mathcal{L}))$
 $\times \mathrm{Prob}_{\mathrm{GW,nS}}(\rho).$ (25)

Here, the SASI threshold appears like a "rectangular hole" (shown in Fig. 10), with its rejected GW component lying between 0 and Λ_{GW} and its rejected neutrino component lying between 0 and Λ_{ν} . Only the (\mathcal{L}, ρ) lying outside of the defined rectangle hole would be identified as a detection of the SASI imprints.

(3) *Mixed:* SASI is established when f(L, ρ) > Λ, where f is a function of L and ρ. Here, the SASI threshold can be defined in several ways, such as L × ρ > Λ, which is denoted as the "x × y = const" threshold, or as L + ρ > Λ, which is denoted as the "x + y = const" threshold (see Fig. 10 for an illustration). The P_D^{comb} based on the above thresholds are

$$P_{\rm D}^{\rm comb, x \times y} = \int_{\mathcal{L} \times \rho > \Lambda}^{\infty} d\ln(\mathcal{L}) d\rho \operatorname{Prob}_{\nu, \mathrm{S}}(\ln(\mathcal{L})) \\ \times \operatorname{Prob}_{\mathrm{GWS}}(\rho)$$
(26)

and

$$P_{\rm D}^{\rm comb, x+y} = \int_{\mathcal{L}+\rho>\Lambda}^{\infty} d\ln(\mathcal{L}) d\rho \operatorname{Prob}_{\nu, \rm S}(\ln(\mathcal{L})) \\ \times \operatorname{Prob}_{\rm GW, \rm S}(\rho).$$
(27)

The false alarm probability is defined similarly, with $\operatorname{Prob}_{\nu,S} \rightarrow \operatorname{Prob}_{\nu,nS}$ and $\operatorname{Prob}_{GW,S} \rightarrow \operatorname{Prob}_{GW,nS}$.

We will discuss the relationships between the above combined $P_{\rm D}$ in more detail in the next section. Note that the strength of the SASI imprints may depend on the distance *D* differently for neutrinos and GWs. So a combination of relatively high \mathcal{L} and relatively low ρ (or vice versa) from a CCSN is possible. The two-dimensional probability distribution map may provide a way to account for possible tensions between neutrino and GW SASI identifications, and provide a statistical interpretation for a combined data set including both neutrino and GW observations.

B. Results: The joint SASI-meter

Results are presented in Fig. 10, where the twodimensional probability density distributions of $\{\ln(\hat{\mathcal{L}}_{\nu}), \hat{\rho}_{GW}\}$ are shown for various CCSNe distances.

We first compare the two-dimensional probability density distributions of the joint indicator for SASI and no-SASI, for fixed *D*. As expected, for all three distances investigated here, the joint SASI indicator in the SASI scenario is likely to be located in the upper right region of the panel, while the indicator in the no-SASI scenario is likely to reside in the lower left corner. For example, for D = 1 kpc, the indicator is most likely to be located in the region {0.5, 0.4} when SASI appears, while it is most likely to be near {0.0, 0.0} when SASI is absent from the signatures.

We then consider how the two-dimensional probability density distributions of the joint SASI indicator vary with distance. As *D* increases, the PDF of the SASI indicator becomes broader, due to the increasing importance of the statistical fluctuations, in both neutrino and GW signatures, which results in larger uncertainties of both $\ln(\hat{\mathcal{L}}_{\nu})$ and $\hat{\rho}_{GW}$. Consequently, the PDFs for SASI and no-SASI have increasing overlap, meaning that the detectability of SASI activities in this joint analysis decreases with increasing distance. Figure 10 shows another interesting trend: as *D* increases, the maximum of the two-dimensional PDF for the SASI scenario moves from the $\ln(\hat{\mathcal{L}}_{\nu}) > \hat{\rho}_{GW}$ region to the $\ln(\hat{\mathcal{L}}_{\nu}) < \hat{\rho}_{GW}$ one. Such behavior reflects the faster decline of the sensitivity in neutrinos with distance compared to GW signatures.

Given the two-dimensional PDFs of $\{\ln(\hat{\mathcal{L}}_{\nu}), \hat{\rho}_{GW}\}$, we perform quantitative SASI-/no-SASI-scenario identification by constructing the receiver operating curves (P_D^{comb}) as a function of P_{FI}^{comb}) at different CCSNe distances, according to the prescriptions discussed in Sec. III [Eqs. (22)–(27)].

The resulting curves are shown in Fig. 11. Note that for the "Logical And" and "Logical Or" prescriptions the two-dimensional SASI thresholds are composed of two independent thresholds Λ_{ν} and Λ_{GW} . By varying Λ_{ν} and Λ_{GW} independently, multiple ROC curves are found (because one specific $P_{\rm D}^{\rm comb}$ corresponds to multiple $P_{\rm FI}^{\rm comb}$'s), which form a "receiver operating band" in the $P_{\rm D}^{\rm comb}$ - $P_{\rm FI}^{\rm comb}$ plane. For comparison, in Fig. 11 we also show the single-messenger ROCs (labeled ν and GW) from Eqs. (14) and (17).

From Fig. 11, one can see the dependence of the ROCs on *D*. For D = 1 kpc, the performance of the ν -only ROC is much better than the GW-only one, with the multimessenger ROCs being intermediate between the two. However, as the CCSNe distance increases, the decline of the ν -only ROC performance is much faster compared to the GW-only one. As *D* increases, interestingly, we find that the strategy of jointly using neutrino/GW information

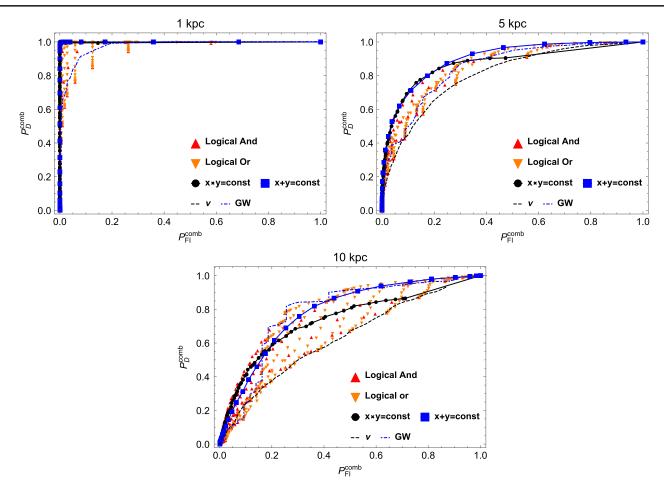


FIG. 11. Combined receiver operating curve at 1 kpc (upper left), 5 kpc (upper right), and 10 kpc (lower) based on Fig. 10, using different selection thresholds.

could provide ROCs that perform better than both singlemessenger ROCs. For example, at 5 kpc, we observe that $P_{\rm D} \approx 71\%$ at $P_{\rm FI} = 10\%$ using the " $x \times y = {\rm const}$ "

TABLE II. $P_{\rm D}$ corresponding to $P_{\rm FI}$ of 10%, which we take as the operating point, and 20%, where the difference between ν and GW channel becomes more pronounced, for the distances of 1 kpc, 5 kpc, and 10 kpc in the single-messenger (GW and ν) and combined cases.

$P_{\rm FI}$	Channel	$P_{\rm D}$ (10 kpc)	$P_{\rm D}~(5~{\rm kpc})$	$P_{\rm D}~(1~{\rm kpc})$
0.10	GW	≈0.15–0.25	≈0.40–0.50	0.90
0.10	ν	0.25	0.40	1.00
0.10	Logical And	≈0.19–0.47	$\approx 0.47 - 0.68$	≈0.80–1.00
0.10	Logical or	≈0.18–0.25	$\approx 0.48 - 0.60$	≈0.83-1.00
0.10	$x \times y = \text{const}$	0.41	0.71	1.00
0.10	x + y = const	0.35	0.70	1.00
0.20	GW	≈0.60–0.65	0.70	1.00
0.20	ν	0.35	0.65	1.00
0.20	Logical And	≈0.40–0.69	≈0.69–0.82	≈0.90–1.00
0.20	Logical or	≈0.36–0.65	≈0.66–0.76	≈0.92–1.00
0.20	$x \times y = \text{const}$	0.59	0.83	1.00
0.20	x + y = const	0.61	0.83	1.00

threshold in the multimessenger analysis, while $P_{\rm D} \approx 40-50\%$ at $P_{\rm FI} = 10\%$ in the single-messenger analysis.

We summarize the $P_{\rm D}$ based on single-messenger and multimessenger methods at $P_{\rm FI} = 0.1$ and $P_{\rm FI} = 0.2$ in Table II. As shown in Table II, when using the " $x \times y = \text{const}$ " threshold and the "x + y = const" threshold, $P_{\rm FI} = 0.1(0.2)$ corresponds to a range of $P_{\rm D}$'s, rather than a single $P_{\rm D}$. This is because when using these types of SASI-meter thresholds, we obtain a "receiver operating band," as explained above. Finally, the $P_{\rm D}$ at 10 kpc (with $P_{\rm FI} = 0.1, 0.2$) and at 5 kpc (with $P_{\rm D} = 0.1$) in the GW channel cannot be determined accurately because of the difficulty of obtaining enough GW triggers and the resulting "zigzags" on the GW ROC.

V. DISCUSSION AND CONCLUSION

We have introduced a novel multimessenger methodology ("SASI-meter") to identify and characterize the presence of SASI activities in a future core-collapse supernova event that is detected with neutrinos and gravitational waves.

For each messenger, the SASI-meter indicates the presence of SASI with a desired maximum false

identification probability. We studied the effectiveness of the procedure with ROCs, which give the probability of establishing the presence of SASI for a fixed false identification probability, that are tuned on generic properties of the GW and neutrino CCSNe signatures when the SASI is not present. The results were produced using numerical simulations with and without SASI-induced signatures at different distances from ground-based detectors.

The method also performs parameter estimation, by characterizing the features of the SASI oscillations such as the oscillation amplitude, frequency, duration, and starting time.

More explicitly, we characterized the pipeline with random realizations of reconstructed ν and GW signals for the test example. For the ν , this means adding Poissonian fluctuations to signals that have SASI as well as signals where the SASI was removed. In the case of future detections, we can achieve the same result by taking a smoothed-out version of the detected neutrino luminosities as the no-SASI ν signature, and randomize it with Poissonian fluctuations to identify the threshold for the desired $P_{\rm FI}$ used as a reference in this work for the single-channel or multiple-channel identification mode (here $P_{\rm FI} = 0.1$). For the GW channel, in the illustrative example used in this paper, we injected the GWs with and without SASI in real interferometric noise. In a realistic scenario, the no-SASI injections can be used to tune the threshold on the identification metric for the desired single- or multimessenger $P_{\rm FI}$ (here $P_{\rm FI} = 0.1$). The signals in the ν and GW channels with SASI were used in this paper to characterize the performance of the GW- ν SASI meter.

We anticipate and observe a different scaling with distance in the frequency estimation uncertainties in the GW and neutrino channels because of the different dependence of the signal amplitude with respect to the source distance. The method is capable of accounting for an intrinsic uncertainty of the SASI frequency due to, for example, the shock radius fluctuations of a progressive frequency drift like in the case of a spiral SASI. The frequency estimate in the neutrino channel uses the peak frequency in the spectrum. In the GW channel, the estimated SASI frequency is a weighted average frequency among the wavelet components in the SASI time frequency region. This concept of an average is well defined even when the frequency drifts because of shock radius fluctuations or other reasons. We have also shown estimates of the slope of the *g*-mode in the GW channel. The estimation of the SASI duration in the GW channel might not be optimized yet. We expect similar identification performances at order-of-magnitude larger distances for thirdgeneration GW detectors.

For neutrinos we have elaborated on a previously presented maximum likelihood method. Our singlemessenger results show that for a galactic event the SASI can be identified with high confidence, and its main parameters can be estimated.

The single-messenger methods have been combined into a fully consistent multimessenger SASI-meter, where a joint receiver operating curve is found. The results confirm the power of multimessenger astrophysics; they show that, for a typical galactic supernova (distance $D \gtrsim$ few kpc), a joint analysis can be more sensitive than each of the singlemessenger ones, depending on the degree of optimization of the integration domain of the multidimensional probability distribution curves.

Given future galactic CCSNe observations, the SASImeter calculates the quantities characterizing the strength of SASI-induced oscillations, namely, $\{\ln(\hat{\mathcal{L}}_{\nu}), \hat{\rho}_{GW}\}$, based on observed neutrino and GW data in the time-frequency domain. The $\{\ln(\hat{\mathcal{L}}_{\nu}), \hat{\rho}_{GW}\}$ based on observations can be compared with theoretical predictions. In this way, the results of the SASI-meter could be used to validate and interpret future CCSNe numerical simulations. The detailed calibration procedure was not discussed in this work and will be left for future investigations.

In this work we used the results of a representative but specific numerical simulation. As more simulations become available with both neutrino and GW signatures, we will repeat the SASI-meter analysis. We expect that the thresholds to achieve a given $P_{\rm FI}$ will change weakly when more models are included in the analysis, while the identification range will change depending on the relative amplitude of the SASI.

Future works may include 1) generalizing the GW SASImeter to other wavelets bases (in the future, the method could be tested in those directions as well), or 2) applying the SASI-meter method to a model with a fast-rotating progenitor (note that the KKHT model has a nonrotating progenitor). The PDF of $\{\ln(\hat{\mathcal{L}}_{\nu}), \hat{\rho}_{GW}\}$ in the SASI scenario might be sensitive to the observational direction with respect to the rotating axis of CCSNe. A directiondependent SASI activity indicator $\{\ln(\hat{\mathcal{L}}_{\nu}), \hat{\rho}_{GW}\}$ from simulations can then be compared with observations. In this way, the joint SASI-meter may help to identify the rotating axis of the CCSN in a quantitative way.

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APPENDIX A: NEUTRINO ANALYSIS: OPTIMIZATION OF THE SASI INITIAL TIME AND DURATION

In the SASI-meter method illustrated above, the duration τ and starting time t_0 are fixed before the process of identifying the existence and analyzing the features of the SASI. Let us now discuss a new aspect of this work, namely, that we do not assume prior values of t_0 and τ (as was done in Ref. [5]), but rather analyze the output of the KKHT simulation for their extremal values. Following the method suggested in Ref. [47], the process of SASI identification described above is performed repeatedly for various time series $[t_0, t_0 + \tau]$. The extremal interval $[t_0, t_0 + \tau]$ for the detection of the SASI can be found by exploring all the possible pairs (t_0, τ) . The starting time t_0 is varied in the range 50–210 ms, with a time step of 20 ms.

of 10 ms. For each pair (t_0, τ) , the interval $[t_0, t_0 + \tau]$ is used to analyze the detectability as well as the features of the SASI [48]. The detection probability P_D is calculated (for a fixed $P_{\rm FI}$). The optimal values of t_0 and τ are identified as those for which P_D is a maximum. Indeed, if the SASIinduced fluctuations predicted by the KKHT SASI model are present in $[t_0, t_0 + \tau]$, then the PDF of \mathcal{L} from the KKHT SASI model and that from the KKHT no-SASI model would deviate from each other. Thus, the resultant P_D would be high. Otherwise, the P_D would be low.

By applying the SASI-meter to neutrino signatures of $[t_0, t_0 + \tau]$, P_D and P_{FI} in a specific neutrino time series are found. In Fig. 12, at fixed $P_{FI} = 10\%$ we plot P_D in neutrino times series of $[t_0, t_0 + \tau]$, where t_0 (τ) varies from 80 (30) ms to 200 (70) ms. We find that P_D obviously increases when the neutrino time series of $[t_0, t_0 + \tau]$ (partly) overlaps the SASI region predicted by the

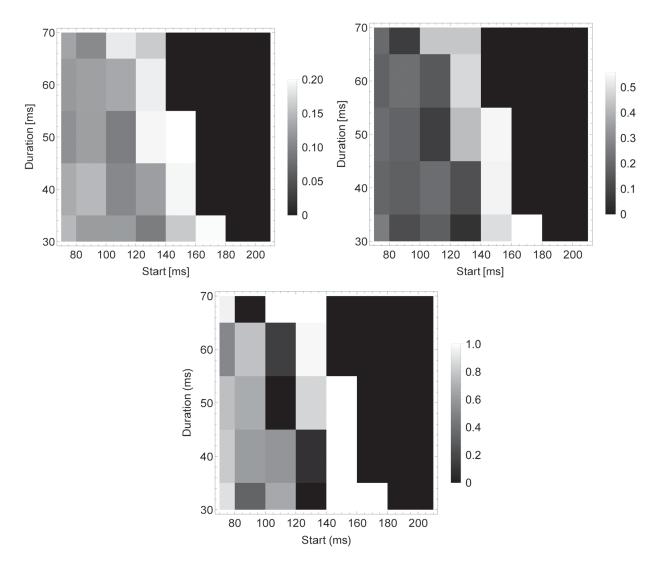


FIG. 12. Probability of detection P_D of SASI at a false alarm rate $P_{FI} = 10\%$. The P_D 's are evaluated in neutrino time series with different durations and starting times. The upper left, upper right, and lower panels are for P_D 's with CCSNe distances of 10, 5, and 1 kpc.

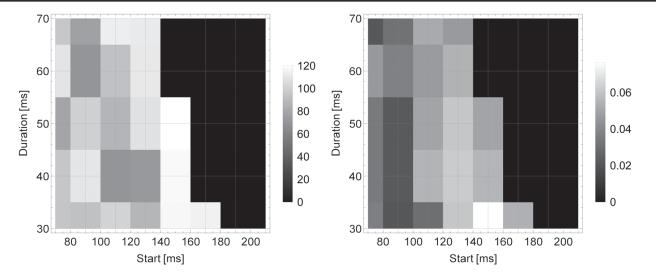


FIG. 13. Averaged optimal frequency (left) and amplitude (right) evaluated in neutrino time series with different durations and starting times. The corresponding distance to the CCSNe is 1 kpc.

KKHT model. For example, at a CCSNe distance D = 1 kpc, $P_D \approx 1.0$ in the neutrino time series of [110 ms, 110 + 70 ms], [130 ms, 130 + 60 ms], [130 ms, 130 + 70 ms], [150 ms, 150 + 30 ms], [150 ms, 150 + 40 ms], [150 ms, 150 + 50 ms], and [170 ms, 170 + 30 ms], and the SASI region predicted by the KKHT model is approximately in the range [150 ms, 150 + 50 ms]. In the neutrino time series ending after 200 ms, $P_D = 0.0$. This is because the simulation of CCSNe in KKHT model was truncated after 200 ms, $P_D \ll 1.0$, indicating that the SASI-induced neutrino oscillations cannot be identified in these time series. The P_D distributions at different distances are qualitatively similar.

As one may notice, the extremal (t_0, τ) defines the region where the SASI activity in the KKHT model has the highest probability to be verified against observations. Naturally, this interval may not adequately represent the starting time and the duration of the SASI activities residing in observed neutrino events. In fact, the comparison with the true, observed values of t_0 and τ is a test of the model. Such true values can be roughly measured using the SASI-meter method as follows. First, note that the SASI modulation on neutrino and GW emissions is predicted as a quasiperiodic signature and its oscillation frequency remains almost constant for tens of ms. Thus, when exploring the $\tilde{\Omega}_{SASI}$ for various points (t_0, τ) in the parameter space we find that for all time intervals $[t_0, t_0 + \tau]$ that at least partially include SASI-induced oscillations, the estimated SASI frequency \tilde{f}_{S} is approximately the same, representing the "monochromatic" feature of SASI-induced oscillations. By identifying the region of the parameter space where \tilde{f}_S stays constant, we can find the SASI region in the time domain and estimate t_0 and τ , without relying on a model. In this way, the SASI-meter can give an approximate estimation of the SASI duration and starting time. Such an estimation is necessarily rough and less accurate than the measurements for f_s and a, since the SASI duration and starting time are features in the time domain, while the SASI-meter is designed mainly for analyzing SASI features in the frequency domain.

In Fig. 13, the extremal frequencies that maximize the \mathcal{L} in various neutrino time series are plotted. Note that for time series of $[150 \,\mathrm{ms}, 150 + 30 \,\mathrm{ms}]$, $[150 \,\mathrm{ms}, 150 + 40 \,\mathrm{ms}]$, [150 ms, 150 + 50 ms], the corresponding optimal frequencies are approximately identical, indicating that a monochromatic SASI oscillation exists in these periods. Since the determination of optimal SASI oscillation parameters does not require a prior PDF of \mathcal{L} , we can model independently identify the neutrino time series with SASI-induced oscillations and measure the corresponding oscillation frequency given the observation of neutrino signatures. For example, at 1 kpc, the fact that the optimal frequency in $[t_0, t_0 + \tau]$ with $t_0 < 150$ ms is different from those in [150 ms, 150 + 30 ms], [150 ms, 150 + 40 ms], and [150 ms, 150 + 50 ms] suggests that the SASI activity happens no sooner than ≈ 150 ms. Additionally, the fact that the optimal frequencies in [150 ms, 150 + 30 ms], [150 ms, 150 + 40 ms], and [150 ms, 150 + 50 ms] are approximately the same indicates that the SASI activity lasts for $\gtrsim 50$ ms.

APPENDIX B: GW ANALYSIS: INITIAL PROCESSING AND g-MODE REMOVAL

In this appendix, we define the different GW parameters and the method used by the cWB algorithm to produce the likelihood time-frequency maps, describe the GW data we are using, the retuning of some cWB parameters that we perform, and the *g*-mode slope estimation for the removal of the *g*-mode region in order to define the SASI-dominant region (or the SASI region) where we perform the SASI parameter estimation.

1. Generation and processing of the time-frequency maps

For a single detector (with equal arms of length *l*), the response X(t) is the sum of detector noise n(t) and the GW signal contribution $\xi(t)$;

$$X(t) = n(t) + \xi(t), \tag{B1}$$

where $\xi(t)$ depends on the absolute difference in the change in length of the two arms $\delta l_x(t)$ and $\delta l_y(t)$ relative to the original length,

$$\xi(t) = \frac{|\delta l_x(t) - \delta l_y(t)|}{l} = F_+ h_+(t) + F_\times h_\times(t),$$

where $h_+(t)$ and $h_{\times}(t)$ are the plus and cross polarization components of the GW and F_+ and F_{\times} are the respective antenna patterns. ξ can also be expressed as

$$\xi = \zeta \cdot \tilde{A} + \tilde{\zeta} \cdot A, \tag{B2}$$

where $\zeta = h_+ + ih_{\times}$ and $A = \frac{1}{2}(F_+ + iF_{\times})$.

Since the response of the interferometer, or the detector data, is in the form of a time series $X = \{x[0], x[1], ..., x[I]\}$ that may or may not contain a GW signal, a decision has to be made regarding the presence (hypothesis H_1) or absence (hypothesis H_0) of the GW signal described by the two probability densities $p(x|H_1)$ and $p(x|H_0)$, respectively. Any decision rule would then be based on a threshold applied to these densities. For this, we define the likelihood ratio $\Lambda(x)$ as

$$\Lambda(x) = \frac{p(x|H_1)}{p(x|H_0)}.$$
(B3)

In GW data analysis, due to the standard whitening procedure performed by cWB, we can assume the noise to be a zero mean temporarily uncorrelated Gaussian process with standard deviation σ (non-Gaussian noise components are managed separately). The probability density functions associated with the two hypotheses H_0 and H_1 become

$$p(x|H_0) = \prod_{i=1}^{I} \frac{1}{\sqrt{2\pi\sigma}} e^{-\left(\frac{x^2[i]}{2\sigma^2}\right)},$$
 (B4)

$$p(x|H_1) = \prod_{i=1}^{I} \frac{1}{\sqrt{2\pi\sigma}} e^{-\left(\frac{(x[i]-\xi[i])^2}{2\sigma^2}\right)}.$$
 (B5)

Then, the logarithmic value of the likelihood ratio, which we simply call the likelihood, is

$$\rho = \ln(\Lambda(x)) = \ln\left(\prod_{i=1}^{I} e^{(\frac{1}{\sigma^2}(x[i]\xi[i] - \frac{1}{2}\xi^2[i]))}\right)$$
$$= \sum_{i=1}^{I} \frac{1}{\sigma^2} \left(x[i]\xi[i] - \frac{1}{2}\xi^2[i]\right). \quad (B6)$$

For *N* detectors, $\sigma = \{\sigma_1, \sigma_2, ..., \sigma_N\}, \xi = \{\xi_1, \xi_2, ..., \xi_N\}, F_+ = \{F_{+1}, F_{+2}, ..., F_{+N}\}, F_{\times} = \{F_{\times 1}, F_{\times 2}, ..., F_{\times N}\}, and A = \{A_1, A_2, ..., A_N\}$ and the total likelihood for the *N* detectors becomes

$$\rho = \sum_{k=1}^{N} \sum_{i=1}^{I} \frac{1}{\sigma_k^2} \left(x_k[i] \xi_k[i] - \frac{1}{2} \xi_k^2[i] \right).$$
(B7)

If we introduce

$$f_{+} = \left\{ \frac{F_{+1}}{\sigma_{1}}, \frac{F_{+2}}{\sigma_{2}}, \dots, \frac{F_{+N}}{\sigma_{N}} \right\},$$
(B8)

$$f_{\times} = \left\{ \frac{F_{\times 1}}{\sigma_1}, \frac{F_{\times 2}}{\sigma_2}, \dots, \frac{F_{\times N}}{\sigma_N} \right\},\tag{B9}$$

$$A_{\sigma} = \left\{ \frac{A_1}{\sigma_1}, \frac{A_2}{\sigma_2}, \dots, \frac{A_N}{\sigma_N} \right\},\tag{B10}$$

$$g_c = \sum_{k=1}^N \frac{A_k^2}{\sigma_k^2},\tag{B11}$$

where g_c is the network antenna pattern, the transformation $g_c \rightarrow g'_c$ which makes the imaginary part of g'_c vanish transforms to the dominant polarization frame (DPF). If $g_c = |g_c|e^{2i\gamma}$, the transformation of A_k is $A'_k = A_k e^{-i\gamma}$ and thus the normalized antenna patterns transform as

$$f'_{k+} = f_{k+}\cos(\gamma) + f_{k\times}\sin(\gamma), \qquad (B12)$$

$$f'_{k\times} = -f_{k+}\sin(\gamma) + f_{k\times}\cos(\gamma).$$
(B13)

The unitary vectors of the DPF are expressed as $e'_{+} = \frac{f'_{+}}{|f'_{+}|}$ and $e'_{\times} = \frac{f'_{\times}}{|f'_{\times}|}$. It can be shown, after transforming to the DPF, where the plus and cross antenna patterns are orthogonal, and assuming they have the same magnitude, that the maximum likelihood value, after applying the conditions $\frac{\delta\rho}{\delta h_{+}} = 0$ and $\frac{\delta\rho}{\delta h_{\times}} = 0$, is (see Ref. [49])

$$\rho_{\max} = (X \cdot e'_{+})^{2} + (X \cdot e'_{\times})^{2} = \sum_{k} \frac{\xi_{k}^{2}}{\sigma_{k}^{2}} = \sum_{k} \text{SNR}_{k}^{2}, \quad (B14)$$

where the SNR of the detector *k* calculated from the detector response ξ_k and noise variance σ_k is expressed as $\text{SNR}_k^2 = \frac{\xi_k^2}{\sigma_k^2}$. We can see that the maximum likelihood is the sum of squared SNR of the detectors.

2. GW data

For the processing of the interferometric data, we apply the CCSNe configuration of the cWB algorithm in the 16–2048 Hz band, to detect and reconstruct the CCSNe GW signals, after injecting those signals on LIGO noise from the O3 run [46] using data from detectors H1 and L1 (GPS times 1256652800 to 1269563392). In order to account for the time variability of the noise, we prepare hundreds of injected events (triggers) separately for each distance of 1, 5, and 10 Kpc. We use our prepared SASI and no-SASI waveforms (injections), with amplitudes scaled according to the distance, and the injections are added to the noise at different times.

3. Cluster formation, likelihood time-frequency maps, and cWB parameter tuning

cWB is an excess-power search algorithm for detecting and reconstructing GWs based on a constrained likelihood formalism [31]. The analysis of GW strain data is performed in a wavelet domain [50] using the wavelet transform, a tool that transforms the signal into the timefrequency domain. First, the cWB algorithm performs data conditioning on the calibrated strain data by applying a linear prediction error (LPE) filter to remove "predictable" components from the time series, such as lines of stationary noise. The LPE filter and whitening is applied in the wavelet domain individually for each wavelet layer [31].

The cWB algorithm uses Wilson-Daubechies wavelets, resulting in two-dimensional maps that are formed from wavelet components (pixels) with different time-frequency resolutions.

Since the wavelet decomposition is performed through seven wavelet bases in parallel, the time and frequency resolutions of each pixel are in general different, as seen in Fig. 7. The cWB algorithm performs wavelet/pixel selection. Wavelets with amplitudes above a threshold, designed to spare only a small percentage of the noiseinduced ones [black pixel probability (bpp)], are retained in each frequency. Each wavelet component is defined by its time, frequency, likelihood, and time-frequency resolution. The frequency resolution is inversely related to the time resolution.

For the time-frequency position *i*, *j* in the time-frequency plane (after wavelet decomposition), with time resolution δt , the resulting signal is expressed as $x(i, j, \delta t)$. From Eq. (B7), the likelihood thus needs to be defined over both time and frequency as (see Ref. [49])

$$\rho_{c} = \sum_{ij} \sum_{k=1}^{N} \frac{1}{\sigma_{k}^{2}(i,j)} \left(x_{k}(i,j,\delta t_{k})\xi_{k}(i,j,\theta,\phi) - \frac{1}{2}\xi_{k}^{2}(i,j,\theta,\phi) \right)$$
$$= \sum_{ij} \rho(i,j,\theta,\phi), \tag{B15}$$

where the likelihood functional $\rho(i, j, \theta, \phi)$, defined over time-frequency positions *i*, *j* and angles θ, ϕ referring to the source coordinates, is expressed as

$$\rho(i, j, \theta, \phi) = \sum_{k=1}^{N} \frac{1}{\sigma_k^2(i, j)} \left(x_k(i, j, \delta t_k) \xi_k(i, j, \theta, \phi) - \frac{1}{2} \xi_k^2(i, j, \theta, \phi) \right).$$
(B16)

The maximum likelihood statistic, for a given location (i, j), is then determined by maximizing the likelihood functional over the source coordinates θ, ϕ ($\rho_m(i, j)$). The likelihood time-frequency map, a pixel map, is then obtained from the set of these maximum likelihood values for each time and frequency. Coherent clusters are formed from these pixels, which are formed by selecting pixels with a maximum likelihood $\rho_m(i, j)$ greater than a chosen threshold, and they are composed of pixels belonging to all detectors involved in the network. Final clusters are used to reconstruct the gravitational-wave signal.

The cWB pipeline is divided into two stages: the coherent event generator and the post-production analysis. After the event generation, as explained above, the resulting data is stored as the output trigger files and the post-production stage deals with the selection of the optimal set of statistics. Some of the post-production metrics are explained below.

The likelihood is a quadratic form that can be expressed in a matrix form $[\rho_{mn}]$. And according to Eq. (B14) we have

$$\rho_{\rm max} = (X \cdot e'_+)^2 + (X \cdot e'_\times)^2. \tag{B17}$$

Expanding the dot product in N-dimensional space (where the m, n indices refer to the detector number), we get

$$\rho_{\max} = \sum_{mn} \{ (X_m e'_{+m}) (X_n e'_{+n}) + (X_m e'_{\times m}) (X_n e'_{\times n}) \} \\
= \sum_{mn} \rho_{mn}.$$
(B18)

We define the coherent network energy, or simply coherent energy (E_c) , as the sum of the off-diagonal terms of the likelihood matrix:

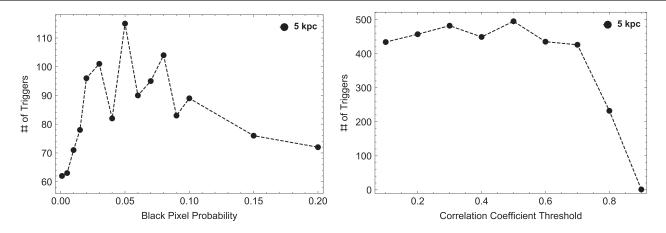


FIG. 14. Optimal choices of bpp (left) and the *CC* threshold (right) at 5 Kpc in terms of maximizing the number of triggers produced by the cWB algorithm, while other cWB parameters remain unchanged. The cWB algorithm arranges a number of triggers into job files that are executed individually. Here, the analysis is carried out for ten such jobs for each bpp value and 50 jobs for each *CC* value shown in the graph.

$$E_c = \sum_{m \neq n} \rho_{mn}.$$
 (B19)

The null energy (N_{ull}) , which is the total reconstructed energy of noise, is defined as

$$N_{\rm ull} = |X - \xi|^2.$$
(B20)

The network correlation coefficient (CC), which is an estimate of coherence among different interferometers, is defined as

$$CC = \frac{E_c}{E_c + N_{\text{ull}}}.$$
 (B21)

Real gravitational-wave events are expected to have a CC closer to 1, and noise events are expected to have a $CC \ll 1$.

Thresholds on the event metrics are applied. For example, we applied a threshold on the correlation coefficient CC and threshold on Z, which is the effective correlated amplitude or effective correlated SNR of a trigger which is based on E_c (see Ref. [31]). The relationship between Z and CC is given by

$$Z = \sqrt{\frac{e_c}{N}}CC,$$
 (B22)

where the reduced correlated energy e_c is defined as

$$e_c = \sum_{m \neq n} \rho_{mn} |r_{mn}|, \qquad (B23)$$

where m, n = 1, 2, ..., N refer to the detector number. The network correlation coefficient r_{mn} , obtained by cross-correlating detector data, is defined as

$$r_{mn} = \frac{\rho_{mn}}{2\sqrt{\rho_{mm}}\sqrt{\rho_{nn}}}.$$
 (B24)

The traditional tuning process of the cWB algorithm focused on producing the best ROC for the detection of the overall signal and not the identification of a specific feature in the signal, like the SASI. For this reason, we redo some of the tuning. Here, the focus is to maximize the chance an event is reconstructed since otherwise it is not possible to identify the presence of the SASI when it is present. For the cWB internal parameter optimization, we execute different cWB jobs at different bpp and *CC* thresholds to find the set of values that give us the maximum number of triggers. As we can see from Fig. 14, the optimal bpp is 0.05 and the optimal *CC* threshold is 0.5. However, exploring a systematic optimization of all cWB thresholds is beyond the scope of this study.

4. g-mode location, parameter estimation, and removal

The growth of the frequency of the fundamental mode of oscillation of the PNS in the GW spectrogram is the main feature that all CCSNe numerical simulations display. Estimating the q-mode slope has merit by itself, but for the detection of the SASI its contribution in the GW event is not necessarily useful as its contribution to the overall SNR does not make the total SNR a good indicator for the presence of a SASI. In this regard, the g-mode location and estimation in this work is considered sufficiently wellperformed in terms of benefits to the SASI ROC curves. The optimization of the *q*-mode parameters is the topic of future publications. We locate the g-mode region, in terms of start time and initial slope, and later, remove the pixels from that part of the event to focus on the SASI pixels. It is also possible that the statistically more significant wavelet components induced by noise are

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scattered in the time-frequency plane (but in proximity of the GW event). A second mechanism where energy could percolate in the SASI region is an impulsive stimulation of the PNS. This could happen, for example from an unusually large accretion funnel of material, inducing a broad band GW spike that also contains lower frequencies [25,35]. Also, there is a possibility that a CCSN explosion could have turbulent components containing some GW energy as well in the SASI region. Nevertheless, the metric introduced in Eq. (16) allows to produce probability distributions where the two scenarios can be distinguished (see, for example, Fig. 8).

We apply a (Python) code to process wavelet maps for all reconstructed triggers to remove the q-mode contribution in those triggers, detect SASI, and estimate its frequency and duration. For the *q*-mode parameter estimation, we remove pixels below 200 Hz, because there is no significant energy of the q-mode component there but the SASI component may be present which can affect our q-mode parameter estimation. Since the g-mode slope estimation is affected by the noise-induced pixels, we discard pixels with likelihoods below the event-dependent arithmetic mean of likelihood of the event. Next, as the g-mode is one of the most energetic features in the GW event, from the surviving pixels, we only choose pixels in an interval of 0.2 seconds (roughly twice the visible *q*-mode duration for most waveforms) identified as the 0.2-second interval containing the most energetic pixels for the *q*-mode parameter estimation. Such an interval is selected by comparing the sum of likelihood of the pixels, as the measure of total energy, within all 0.2-second intervals in the event and choosing the one with the maximum value.

If less than two pixels survive, we do not perform slope estimation and thus the following steps for *q*-mode slope estimation are bypassed and the next steps for SASI parameter estimation are implemented (see Sec. III B 1). The preliminary estimation of the slope and intercept of the g-mode line is performed using linear regression on the remaining pixels. For the triggers with estimated slopes outside the physical range (500, 5000) s^{-2} identified from the literature review, the slope-intercept optimization process mentioned below is bypassed, like before, and the next steps for SASI parameter estimation are implemented (see Sec. III B 1). The statistics of the number of triggers for which q-mode slope estimation is not bypassed are observed to be 807 out of 1081 at 1 Kpc, 671 out of 2867 at 5 Kpc, and 357 out of 2764 at 10 Kpc. The estimation of the slope is performed in two steps. The initial one involves a simple linear regression. The estimation is then refined by defining intervals around those values (200 points were taken around the initial estimates, with a relative variation of < 33% with respect to the initial points). In this way, we construct a grid in the slopeintercept plane and use this set of values to minimize the following weighted χ^2 function:

$$\chi^{2} = \sum_{i} \rho^{i} (f_{c}^{i})^{2} \Xi^{i} \left\{ t_{c}^{i} - \frac{(f_{c}^{i} - c)}{m} \right\}^{2}, \qquad (B25)$$

Here, t_c^i is the centre time of *i*th pixel. The f_c^i is the central frequency of the *i*th pixel. The $(f_c^i)^2$ is a weight function used to compensate for the LIGO noise curve. It is necessary because the LIGO noise increases with frequency, resulting in worse SNR at higher frequencies. The ρ^i is another weight function and is the likelihood value of the *i*th pixel. By using ρ^i , we take the energy of the pixels into consideration. This is because the pixels corresponding to signal would be more energetic in general than those corresponding to noise. Finally, since the pixels related to signals are most likely to be clustered together and the pixels related to noise are most likely to be isolated in the t-f maps, we introduce a quantity Ξ^i in the weights, which measures the density of pixels around the *i*th one. It is calculated as the sum of the likelihoods in the neighborhood of the pixel in a window of 0.01s time interval and 25 Hz frequency interval centered around the given pixel in the t-f map.

The summation is done over all pixels remaining in the area of interest of the trigger for all elements in the grid of possible slopes (i.e., m) and intercepts (i.e., c).

Minimization of the multivariate χ^2 function gives the optimal slope and intercept point (average of all slopes and intercepts with minimum chi-squared values) on the slope-intercept grid. Thus, the *g*-mode frequency (f^g) evaluation becomes

$$f^g = m_{\text{opt}}^{\text{GW}} t + c_{\text{opt}}^{\text{GW}}, \qquad (B26)$$

where $c_{\text{opt}}^{\text{GW}}$ and $m_{\text{opt}}^{\text{GW}}$ are the optimized intercept and slope, respectively. For the removal of the *g*-mode region, the

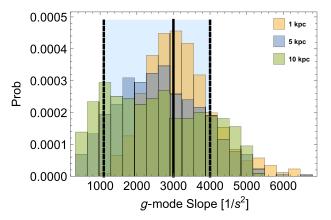


FIG. 15. *g*-mode slope histograms for D = 1, 5, 10 Kpc. The center solid vertical line identifies the slope of the *g*-mode evolution estimated from the spectrogram in Fig. 3, which is 3000 s⁻² (which is the *g*-mode slope of our illustrative example, estimated visually from Fig. 3). The left and right dashed vertical lines represent the range of slopes for slowly rotating progenitors observed in the literature (derived from Refs. [35,51]), which are 1100 and 4000 s⁻², respectively. The variance of the distribution shrinks with closer distances.

g-mode initial time is calculated from the fitted line at 200 Hz as $t_{ini}^g = \frac{(200-c_{opt}^{GW})}{m_{opt}^{GW}}$. When the slope estimation and/or the slope optimization steps are bypassed, we only select the pixels that are 50 ms after the earliest pixel in the time-frequency map of the whole event to be included in the SASI time-frequency region (or, simply, the SASI region). Otherwise, as the *g*-mode initial time t_{ini}^g would be available, only the pixels 50 ms after it are included in the SASI region. In this way, *g*-mode removal from the pixel map is performed. We then have a t-f map that only includes the SASI components. This map is further used for SASI parameter estimation described in Sec. III B 1. The results of the slope optimization at different distances are listed in Table I and shown in Fig. 15.

Using the pixels remaining in the SASI region, after the removal of the *g*-mode components, we estimate the SASI parameters—the central frequency and duration—using the statistical approaches described in Sec. III B 1. We find that with the decrease of the bpp value we use (0.05) with respect to the standard cWB configuration (bpp = 0.1), an extra threshold is needed to remove some noise-related pixels in the SASI time-frequency region, as lower bpp (to increase the number of triggers) would also allow more noise-related pixels in the triggers. We choose a threshold on the likelihood values of the pixels in the SASI region, as a pixel selection criteria for parameter estimation, which is set at 50% of the maximum likelihood value among all pixels in the given event (i.e., an event-dependent threshold).

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