Upper limits on the isotropic gravitational-wave background from Advanced LIGO and Advanced Virgo’s third observing run

R. Abbott et al.*
(LIGO Scientific Collaboration, Virgo Collaboration, and KAGRA Collaboration)

(Received 24 January 2021; accepted 9 June 2021; published 23 July 2021)

We report results of a search for an isotropic gravitational-wave background (GWB) using data from Advanced LIGO’s and Advanced Virgo’s third observing run (O3) combined with upper limits from the earlier O1 and O2 runs. Unlike in previous observing runs in the advanced detector era, we include Virgo in the search for the GWB. The results of the search are consistent with uncorrelated noise, and therefore we place upper limits on the strength of the GWB. We find that the dimensionless energy density $\Omega_{GW} \leq 5.8 \times 10^{-9}$ at the 95% credible level for a flat (frequency-independent) GWB, using a prior which is uniform in the log of the strength of the GWB, with 99% of the sensitivity coming from the band 20–76.6 Hz; $\Omega_{GW}(f) \leq 3.4 \times 10^{-9}$ at 25 Hz for a power-law GWB with a spectral index of 2/3 (consistent with expectations for compact binary coalescences), in the band 20–90.6 Hz; and $\Omega_{GW}(f) \leq 3.9 \times 10^{-10}$ at 25 Hz for a spectral index of 3, in the band 20–291.6 Hz. These upper limits improve over our previous results by a factor of 6.0 for a flat GWB, 8.8 for a spectral index of 2/3, and 13.1 for a spectral index of 3. We also search for a GWB arising from scalar and vector modes, which are predicted by alternative theories of gravity; we do not find evidence of these, and place upper limits on the strength of GWBs with these polarizations. We demonstrate that there is no evidence of correlated noise of magnetic origin by performing a Bayesian analysis that allows for the presence of both a GWB and an effective magnetic background arising from geophysical Schumann resonances. We compare our upper limits to a fiducial model for the GWB from the merger of compact binaries, updating the model to use the most recent data-driven population inference from the systems detected during O3a. Finally, we combine our results with observations of individual mergers and show that, at design sensitivity, this joint approach may yield stronger constraints on the merger rate of binary black holes at $z \gtrsim 2$ than can be achieved with individually resolved mergers alone.

DOI: 10.1103/PhysRevD.104.022004

I. INTRODUCTION

The gravitational-wave background (hereafter referred to as the GWB or the background) is a superposition of gravitational-wave (GW) sources that is best characterized statistically [1]. There are many possible astrophysical and cosmological contributions to the background, including distant compact binary coalescences (CBCs) that cannot be resolved individually [2–6], core collapse supernovae [7–11], rotating neutron stars [12–19], stellar core collapses [20,21], cosmic strings [22–26], primordial black holes [27–29], superradiance of axion clouds around black holes [30–33], phase transitions in the early universe [34–38], and GWs produced during inflation [39–41] or in a preheating phase at the end of inflation [42,43]. While some sources of the GWB, such as slow roll inflation, have a fundamentally stochastic character, others like the background from CBCs are a superposition of deterministic sources.

*Full author list given at the end of the article.
two advantages to using a log uniform prior. First, a log uniform prior gives equal weight to different orders of magnitude of the strength of the GWBs, which is appropriate given our current state of knowledge. Second, a log uniform prior is agnostic as to which power we raise the strain data. It is not clear whether one should put a uniform prior on the strain amplitude, or the strength of the GWB, which scales like the square of the strain. On the other hand, the log uniform prior does not depend on the exponent of the strain data. For completeness, we also present results with a uniform prior on the strength of the GWB in Sec. IV. Results with any other prior can be obtained by reweighing the posterior samples available at [58].

There are several new features in our analysis of the O3 data. First, we incorporate Virgo, by cross correlating the three independent baselines in the LIGO-Virgo network and combining them in an optimal way [59]. Second, in order to handle a large rate of loud glitches in O3, we analyze data where these artifacts have been removed via gating [60, 61]. Third, we perform a careful analysis of correlated magnetic noise that could impact the search. In addition to constructing a correlated magnetic noise budget, as in past runs, we use a Bayesian statistical framework developed in [62] to constrain the presence of magnetic noise.

Perhaps the most interesting source of an astrophysical GWB, given the current network sensitivity, is the GWB from CBCs. Previous studies have shown that this GWB may be detectable with Advanced LIGO and Advanced Virgo running at design sensitivity [63, 64], and the ability to detect such a background has been confirmed with mock data challenges [65–67]. Therefore in this work we carefully consider the implications of our results for the CBC population. We estimate the GWB using the most up-to-date information from observations during O3 [68–73] and compare with the sensitivity of the current and future detector networks. We show that an upgrade of the current Advanced LIGO facilities, known as A+ [74], could dig into a substantial part of the expected parameter space for the GWB at its target sensitivity. Furthermore, we apply the methods of [75] to constrain the merger rate as a function of redshift for binary black holes (BBHs) by combining the GWB upper limits with information about individually resolvable events. We find that the cross-correlation analysis can provide complementary information at large redshifts, compared to the population analysis using individually detectable events alone [76]. We make the results of our cross correlation analysis available [58], enabling further detailed studies of the GWB from CBCs and other models.

The rest of this work is organized as follows. In Sec. II, we review the method of the cross-correlation search. We discuss the data quality procedures and studies we performed in Sec. III. We present the main results of the search in Sec. IV: we derive upper limits on the GWB in Sec. IV A, putting constraints on the presence of scalar-and vector-polarized backgrounds in Sec. IV B, and in Sec. IV C we extend these results by simultaneously fitting for an astrophysical GWB and an effective GWB arising from magnetic correlations of terrestrial origin. We compare our upper limits with a fiducial model for the GWB from CBCs in Sec. VA, and derive constraints on the BBH merger rate using the upper limits on the GWB and observations of individual CBCs in Sec. VB. We conclude in Sec. VI.

II. METHODS

A GWB that is Gaussian, isotropic, unpolarized, and stationary is fully characterized by a spectral energy density. It is standard to express the spectrum in terms of the dimensionless quantity \( \Omega_{GW}(f) \), which is the GW energy density \( d\rho_{GW} \) contained in the frequency interval \( f \) to \( f + df \), multiplied by the GW frequency and divided by \( df \) times the critical energy density \( \rho_c \) needed to have a flat Universe

\[
\Omega_{GW}(f) = \frac{\int df d\rho_{GW}}{\rho_c}.
\]

where \( \rho_c = 3H_0^2c^2/(8\pi G) \), \( c \) is the speed of light, and \( G \) is Newton’s constant. For consistency with other GW measurements (for example those of [68]), we take the Hubble constant from Planck 2015 observations to be \( H_0 = 67.9 \, \text{km s}^{-1} \, \text{Mpc}^{-1} \) [77].

A. Cross correlation spectra

Let us label the GW detectors in the LIGO-Hanford, LIGO-Livingston, and Virgo (HLV) network by the index \( I \in \{H, L, V\} \). We denote the time-series output of the detectors by \( s_I(t) \), and the Fourier transform by \( \tilde{s}_I(f) \). Following [48, 59], we define the cross-correlation statistic for the baseline \( IJ \) as

\[
\hat{C}^{IJ}(f) = \frac{2}{T} \text{Re} \left[ \tilde{s}_I^*(f) \tilde{s}_J(f) \right] S_0(f),
\]

where \( \gamma^{IJ}(f) \) is the normalized overlap reduction function [59, 78, 79] for the baseline \( IJ \), the function \( S_0(f) \) is given by \( S_0(f) = (3H_0^2)/(10\pi^2 f^3) \), and \( T \) is the observation time. In practice, because the noise is nonstationary, we break the data into segments, and then take \( T \) to be the segment duration. We then average together segments using inverse noise weighting [59]. If the noise were stationary, this average would reproduce Eq. (2). This estimator is normalized so that \( \langle \hat{C}^{IJ}(f) \rangle = \Omega_{GW}(f) \) in the absence of correlated noise. In the small signal-to-noise ratio limit, the variance can be estimated as

\[
\sigma^2_{IJ}(f) \approx \frac{1}{2T\Delta f} \frac{P_I(f)P_J(f)}{\gamma^{IJ}(f)S_0^2(f)}.
\]
where $\Delta f$ is the frequency resolution, and $P_1(f)$ is the one-sided power spectral density in detector $I$. Note that $T \Delta f$ need not equal one if several frequency bins are coarse grained around the central frequency $f$ to produce the estimator in Eq. (2).

While we have expressed the cross-correlation estimator in terms of the GW strain channel, in fact this analysis can be applied to any pair of instruments. Following [62], in Sec. III D and Sec. IV C we will also employ these techniques to cross correlate magnetometer channels to search for correlated magnetic noise.

### B. Optimal filtering

Strictly speaking, the optimal estimator for a given signal includes both autocorrelation and cross-correlation terms [48]. We only use the cross correlation, and not autocorrelation, in the search because the noise power spectral density is not known precisely enough to be subtracted accurately, and therefore in practice the cross correlation is nearly optimal. With this caveat, we can construct an optimal estimator to search for a GWB of any spectral shape by combining the cross-correlation spectra from different frequency bins with appropriate weights

$$
\hat{C}_{IJ} = \sum_k w(f_k) \hat{C}_{IJ}(f_k) \sigma_{IJ}^2(f_k),
$$

$$
\sigma_{IJ}^2 = \sum_k w(f_k)^2 \sigma_{IJ}^2(f_k),
$$

where $f_k$ are a discrete set of frequencies, and the optimal weights for spectral shape $\Omega_{GW}(f)$ are given by

$$
w(f) = \frac{\Omega_{GW}(f)}{\Omega_{GW}(f_{\text{ref}})}.
$$

Here, $f_{\text{ref}}$ is a fixed reference frequency. For ease of comparison with previous observing runs, we choose the reference frequency to be $f_{\text{ref}} = 25$ Hz. This is approximately the start of the most sensitive frequency band for the isotropic search as described in [44]. This analysis is very flexible and can be applied to a GWB of any spectral shape. We will report results for a power law GWB of the form

$$
\Omega_{GW}(f) = \Omega_{\text{ref}} \left( \frac{f}{f_{\text{ref}}} \right)^\alpha.
$$

Our final estimator combines information from all baselines optimally using the sum

$$
\hat{C} = \frac{\sum_{IJ} \hat{C}_{IJ} \sigma_{IJ}^2}{\sum_{IJ} \sigma_{IJ}^2}, \quad \sigma^{-2} = \sum_{IJ} \sigma_{IJ}^{-2},
$$

where $\sum_{IJ}$ is a shorthand notation meaning a sum over all independent baselines $IJ$. We can also include cross correlation results from previous observing runs in a natural way by including them in this sum as separate baselines. More concretely, we combine HL-O1, HL-O2, HL-O3, HV-O3, and LV-O3.

### C. Parameter estimation

In order to estimate parameters of a specific model of the GWB, we combine the spectra from each baseline $IJ$ to form the likelihood [80]

$$
p(\hat{C}_{IJ}|\Theta) \propto \exp \left[ -\frac{1}{2} \sum_{IJ} \sum_k \left( \frac{\hat{C}_{IJ} - \Omega_{M}(f_k|\Theta)}{\sigma_{IJ}(f_k)} \right)^2 \right],
$$

where $\hat{C}_{IJ} \equiv \hat{C}_{IJ}(f_k)$, and where we assume that the $\hat{C}_{IJ}$ are Gaussian-distributed in the absence of a signal. The term $\Omega_{M}(f|\Theta)$ describes the model for the GWB, characterized by the set of parameters $\Theta$. This hybrid frequentist-Bayesian approach has been shown to be equivalent to a fully Bayesian analysis in [81].

Equation (8) assumes that cross-correlation spectra measured between different baselines are uncorrelated. This is not strictly true, as different baselines share detectors in common. Correlations between baselines, however, enter at $\mathcal{O}(\Omega^2)$ and so can be neglected in the small-signal limit [59].

In this work we shall consider several different models:

(i) **Noise (N):** $\Omega_N(f) = 0$. We implicitly include uncorrelated Gaussian noise as part of every model that follows.

(ii) **Power law (PL):** $\Omega_{PL}(f) = \Omega_{\text{ref}} \left( \frac{f}{f_{\text{ref}}} \right)^\alpha$. The parameters $\Theta_{PL}$ are the amplitude $\Omega_{\text{ref}}$ and spectral index $\alpha$. We will consider cases in which $\alpha$ is allowed to vary as well as those in which it is fixed.

(iii) **Scalar-vector-tensor power law (SVT-PL):** This model contains tensor polarizations, as allowed in general relativity (GR), and vector and scalar polarizations, which are forbidden in GR but generically appear in alternative theories of gravity. We define $p$ to be an index referring to polarization, $p = \{T, V, S\}$, where $T$, $V$, and $S$ refer to tensor, vector, and scalar polarized GWs, respectively. We assume the GWB for each polarization can be described by a power law, which may be different for each polarization. Thus there are six parameters $\Theta_{\text{SVT-PL}}$, given by the amplitudes $\Omega_{\text{ref}}^{[p]}$ and spectral indices $\alpha_{[p]}$ for each polarization. The model is given by the sum $\Omega_{\text{SVT-PL}}(f) = \sum_p \beta_{[p]}^{[p]}(f) \Omega_{\text{ref}}^{[p]} \left( \frac{f}{f_{\text{ref}}} \right)^{\alpha_{[p]}}$, where $\beta_{[p]}^{[p]}(f) = \gamma_{[p]}^{[p]}(f)/\gamma_{[T]}(f)$ is the ratio of the overlap reduction function for polarization $p$ and baseline $IJ$ to the standard (tensor) overlap reduction function for that baseline [46].

(iv) **Magnetic (MAG):** $\Omega_{\text{MAG}}(f)$ describes correlations between two detectors induced by large-scale
coherent magnetic fields, which can appear as an effective background. We model this effective background in terms of magnetometer correlations and a transfer function between the local magnetic field and the strain channel of the detectors. The free parameters $\Theta_{\text{MAG}}$ describe the coupling function, as described in Sec. IV C.

(v) CBC: $\Omega_{\text{CBC}}(f)$ is determined by an underlying parametrized model for the mass distribution of compact binaries and their merger rate $R(z)$ as a function of redshift. The parameters of this model are described in Sec. V.

We will also consider combinations of these models, for example $\Omega_M(f) = \Omega_{\text{PL}}(f) + \Omega_{\text{MAG}}(f)$. Given the likelihood, we form a posterior using Bayes theorem, $p(\Theta|C_k) \propto p(C_k|\Theta)p(\Theta)$, where $p(\Theta)$ is the prior distribution on the parameters $\Theta$. We will consider different prior choices for each model we consider below.

Finally, it is often of interest to combine upper limits on the amplitude of the GWB in the corresponding frequency band. While we use $\alpha = \{−5, 0, 3, 5\}$; each power law is sensitive to a different frequency band. We do not include it in the cut since it does not provide significantly new information for the nonstationarity cut, because the frequency range is very similar to the one probed by $\alpha = 0$. We remove Hanford data from April 1–April 16 2019 due to nonstationarity arising from calibration lines at 35.9 and 36.7 Hz. These lines were moved below 20 Hz on April 16, 2019.

In principle, the CBC signals known to be present in the data contribute to the integrated cross-correlation. Using the median values for the masses and redshifts of BBH events, we estimate that $\approx \Omega_{\text{CBC}} \approx 10^{-10}$ to the GWB, which is well below the O3 sensitivity. Therefore we do not remove the observed CBCs from the data.

After applying the category 1 vetoes and nonstationarity cut, we found that 17.9% of available livetime was lost in the HL baseline, 22.1% in the HV baseline, and 21.9% in the LV baseline.

B. Time and frequency domain cuts

For each baseline, we require that both detectors in the baseline are in observing mode, and that there are no critical issues with the detector hardware, as defined by category 1 vetoes described in [88,89]. As in previous runs, we apply a nonstationarity cut by removing times where the square root of the variance in Eq. (4) is found to vary by more than 20% between segments. We take the union of the cuts for $\alpha = \{−5, 0, 3, 5\}$; each power law is sensitive to a different frequency band. While we use $\alpha = 2/3$ for the search, we do not include it in the cut since it does not provide significantly new information for the nonstationarity cut, because the frequency range is very similar to the one probed by $\alpha = 0$. We remove Hanford data from April 1–April 16 2019 due to nonstationarity arising from calibration lines at 35.9 and 36.7 Hz. These lines were moved below 20 Hz on April 16, 2019.

In principle, the CBC signals known to be present in the data contribute to the integrated cross-correlation. Using the median values for the masses and redshifts of direct detections in O3a [68], the livetime for O3a, and the inspiral approximation in Eq. 16 of [67], we estimate that resolved sources contribute $\Omega_{\text{GW}} \approx 10^{-10}$ to the GWB, which is well below the O3 sensitivity. Therefore we do not remove the observed CBCs from the data.

After applying the category 1 vetoes and nonstationarity cut, we found that 17.9% of available livetime was lost in the HL baseline, 22.1% in the HV baseline, and 21.9% in the LV baseline.
We do not analyze frequency bins where there is evidence of coherence between instruments that is determined to be caused by the instruments themselves. The coherence between two channels,

\[ \Gamma_{IJ}(f) = \frac{\langle |\tilde{s}_I(f)|^2 \rangle}{\langle |\tilde{s}_J(f)|^2 \rangle} \],

is a useful measure to determine when correlations in a given frequency bin exceed what is expected from uncorrelated data. In the above expression, the angle brackets \( \langle \rangle \) refer to an average over analysis segments. The coherence between the strain and auxiliary channels at a given site can also be used to identify an instrumental source of contamination\[90\].

We removed 13.3% of the frequency band in the HL baseline, 21.5% of the frequency band in the HV baseline, and 18.9% of the frequency band in the LV baseline. However, we only removed 3.2% from HL, 9.3% from HV, and 5.9% from LV below 300 Hz, where the search is most sensitive. In O3, we found many 1 Hz harmonics which were coherent between Hanford and Virgo. We also observed a large coherent line in the HL baseline at 33.2 Hz, which was likely due to the beating of two different calibration lines at Hanford and Livingston, and therefore did not appear in linear coherences between the strain and auxiliary channels. Generally speaking, line mitigation efforts were particularly effective at the LIGO-Livingston detector, and the HL and LV baselines had many fewer coherent lines. The full list of frequencies removed from the analysis is available online\[58\].

### C. Gating

In O3, we found a much higher rate of loud glitches compared to previous observing runs\[89\]. A naive application of the standard nonstationarity cut used in previous searches led to losing >50% of the data when running with 192-s data segments. In order to reduce the amount of data lost to the nonstationarity cut, and thus improve the sensitivity of the search, we pre-conditioned the data by applying a gating procedure. This procedure involves first identifying data from the Hanford and Livingston baselines that contain a glitch, and then zeroing out these data. We defined segments containing a glitch when the root-mean-square (RMS) value of the whitened strain channel in the 25–50 Hz band or 70–110 Hz band exceeded a threshold value. We then removed the glitches from the time series by multiplying the data in these segments by an inverse Tukey window. We found that a total of 0.4% of Hanford data was gated in the data that we analyzed, and 1% of Livingston data for each baseline. We refer the interested reader to\[60\] for further details of the procedure, including the whitened channels and precise thresholds used. This was not necessary for Virgo data due to the lower rate of large glitches.

The impact of gating can be clearly seen on the coherence spectra, as we show in Fig. 1. Compared to the nongated data, many more segments are analyzed after applying nonstationarity cuts, and the spectrum is much closer to what is expected from uncorrelated Gaussian noise. It was discovered that from April 20–25 a 1/120-Hz comb was visible in the Livingston data area around large calibration lines.

---

**FIG. 1.** Coherence spectra for the HL baseline without (left panel) and with (right panel) gating applied. In both cases, we have applied the nonstationarity cut described in the main text. We show the coherence spectrum at 1/32 Hz (the same frequency resolution used for the isotropic search) as a gray line, a rolling average with a 1 Hz resolution as a blue line, and the average value expected for uncorrelated, Gaussian noise which is given by 1 divided by the number of averages used to make the coherence spectrum as a black dashed line. The gated coherence is consistent with the expectation from Gaussian noise, while the spectrum without gating is not. Additionally, we see that more segments are used for averages in the gated spectrum after applying the nonstationarity cut described in the text. This is because without gating, many segments are removed due to a single glitch biasing the power spectrum estimate of the segment. Note that in this figure, we have not applied any frequency notching; the large line visible at 33.2 Hz is due to a beat note in the calibration lines at both H1 and L1 arising from nonlinear couplings.
The comb was caused by an inadvertently running diagnostic camera clicking at regular two minute intervals. To be cautious, we removed this period of time from the analysis. We have verified with a mock data challenge that applying this gating procedure to simulated data did not affect our ability to recover a GWB. This check is described further in [61].

D. Correlated magnetic noise budget

In order to be able to claim detection of a GWB, one must understand and control environmental sources of correlated noise [91]. Some magnetic fields are expected to be correlated between sites and are monitored with sensitive magnetometers placed away from the buildings. For example, Schumann resonances are electromagnetic modes of the Earth-ionosphere resonant cavity [92]. They are coherent on a global scale [93], so if they couple to the interferometer and produce noise in the GW channel, they will cause correlations between the outputs of detectors on different continents [94,95]. If these effects are large enough, they can be a source of confusion noise for cross-correlation searches. In this section we show that there is no evidence for correlated magnetic noise in the O3 GW strain data.

As in past runs [44,96], following [94,95] we create a budget for the magnetic correlations

$$\hat{C}_{mag,II}(f) = \frac{2}{T} \frac{|T_I(f)||T_J(f)|}{\gamma_{IJ}(f)S_0(f)} \text{Re}[\hat{m}_{I}(f)\hat{m}_{J}(f)], \quad (11)$$

where $\hat{m}_{I}(f)$ are Fourier transforms of the magnetometer channels. The coupling functions $T_I(f)$ are estimated by injecting an oscillating magnetic field of a known frequency and amplitude at different locations near each detector, and measuring the resulting output in the GW strain channel. Weekly injections were performed to study the time-dependence of the magnetic coupling [97].

Potential differences in the strength of the magnetic field at the magnetometers located around the detector versus the strength of the field at the “true” coupling location mean that these measurements are only rough estimates, and are susceptible to large uncertainties. This uncertainty is estimated by comparing injections at different locations at each site; to account for this, we include a factor of two uncertainty in the coupling function of each detector [98].

Another possible source of error in the coupling function measurement is that the low-noise magnetometers are located outside, far from the local magnetic noise associated with the buildings, but the weekly injections described above are performed inside. One may worry that ferromagnetic material in the buildings can amplify the outside-to-inside magnetic coupling. However, additional measurements at Hanford suggest that the coupling function from outside to inside the building is less than one. Injections were performed around the corner station using seven frequencies ranging from 11 to 444 Hz, and the magnetic field was measured inside and outside the building at the same distance from the injection coil. A power-law fit to the ratio of the magnetic field measured inside to the field measured outside as a function of frequency indicates that the magnetic coupling is suppressed by up to a factor of 2 in the frequency range 10–100 Hz, however with variation depending on the orientation of the field. To be conservative, we assume the inside-to-outside magnetic coupling is equal to one.

To construct the budget, we first compute a linear interpolation for the coupling function as a function of frequency measured at each detector in each week. For weeks where a coupling function was not measured, we use the coupling function that was nearest in time. For each baseline, and each week, we then multiply the coupling functions for each detector by the magnetic cross-correlation spectrum for that baseline, to form a budget. We use the pair of directions that gives the largest coherence. Studies based on shorter stretches of data indicate that the coherence of the magnitude of the magnetic field can be up to a factor of two larger than the coherence of the worst-case components; therefore to be conservative we multiply the coherence in each detector baseline by a factor of two. We combine the budgets across baselines by using the error bars from the GW channels as weights to account for the relative sensitivity of each baseline, $\hat{C}_{mag} = \sum_{II} w_{II}(f) \hat{C}_{mag,II}(f)$, where $w_{II}(f) = (\sigma_{II}(f)/\sigma(f))^2$. We show an estimate of the correlated magnetic noise compared to the O3 sensitivity curve in Fig. 2, combining all three baselines. The red band shows the range of budgets we obtain accounting for the combined weekly magnetic coupling function measurements, as well as the overall factor of two uncertainty in each detector’s coupling function described above. The overall trend of the red band should be compared with the O3 power-law integrated (PI) curve [99], which shows the sensitivity of our search to power law backgrounds, accounting for integration over frequency. The black dotted line shows the upper range of the budget. Narrowband features should be compared with $\sigma(f)$, shown as a black solid line, which shows the sensitivity to a GWB in every frequency bin. The measurements at Hanford were sampled at a fine frequency resolution due to the use of broadband injections with a large coil [100]. This allowed us to see fine-grained features in the coupling function, such as the broad resonances visible between 80 Hz and 100 Hz Fig. 2. While the exact origin of these resonances is presently unknown, they are correlated with excess motion of test masses in the power recycling cavity [101]. The final budget indicates that the nonobservation of correlated magnetic noise is expected given the coupling function measurements.
Figure 2. Correlated magnetic noise budget, as described in the main text. The band shows the expected range of magnetic contamination, using weekly measurements of the coupling function at each site, and accounting for the uncertainty in the coupling function measurements. We display the O3 sensitivity in two ways. First, we plot the power-law integrated (PI) curve as a red line. This indicates the sensitivity of the search to power-law backgrounds and includes the effect of integrating over frequencies, and should be compared with the overall trend of the red band. Second, we plot the square root of the variance, $\sigma(f)$, as a black line, which gives the sensitivity of the search to narrowband features. This can be compared with narrow features in the upper range of the noise budget, which we show as a black dashed line. We conclude that the O3 sensitivity is well above the level of correlated magnetic noise estimated in O3.

IV. RESULTS

A. Upper limits on the GWB

In Table I we report the point estimate and 1-$\sigma$ error bar from O3 obtained from each baseline independently, as well as combining all three baselines together with the HL baseline results available from O1 and O2, using an optimal filter for three different power law models:

(i) $\alpha = 0$ approximately characterizes cosmic string [22–26], and slow-roll inflation GWBs [39–41] in the LIGO-Virgo frequency band.

(ii) $\alpha = 2/3$ describes the CBC GWB when contributions from the inspiral dominate the GWB, which is a very good approximation in the LIGO-Virgo frequency band [102]. However, this approximation may not be valid for mergers of binaries arising from Population III stars [103], or from heavy BBH mergers with masses above the pair-instability mass gap [104].

(iii) $\alpha = 3$ is a fiducial choice used in past searches which approximately describes some astrophysical sources such as supernovae [10–37,39–88,90,92–105], and corresponds to a GWB that is flat in the strain power, $S_h(f) \propto f^{-3} \Omega_{GW}(f)$ [59].

While we use the entire band 20–1726 Hz to compute the point estimate and error bar, we also show $f_{99\%}^{IL}$, which is the upper frequency of the band starting at 20 Hz that contains 99% of the sensitivity in baseline $IJ$.

The HL baseline contributes most to the sensitivity. The contributions from the baselines that include Virgo are relatively more important at higher frequencies and especially relevant to searches for larger power laws. We note that the point estimates for HV and LV are approximately 2$\sigma$ away from zero, however we do not interpret this as evidence of a signal given that the point estimate of the much more sensitive HL baseline is consistent with zero to within 1$\sigma$. The combined spectrum is shown in Fig. 3. From this figure, one can see that the point estimate fluctuates roughly symmetrically around zero, consistent with expectations from Gaussian noise. Additionally, by comparing with Fig. 1 of [44], it is clear that the addition of Virgo data compensates for a zero in the HL overlap reduction function at around 64 Hz. After having applied the data quality cuts described in Sec. III, data are consistent with uncorrelated, Gaussian noise. The spectra have a $\chi^2$-per-degree-of-freedom value of 0.98.

Since we do not find evidence of a GWB, we place upper limits on the PL model, combining the O3 spectra with the results from previous runs. We report upper limits using both a prior that is uniform in the log of the strength of the GWB, and a prior that is uniform in the strength. We choose to report the upper limit obtained with the log uniform prior as our headline result, because a log uniform prior is a more

Table I. Search results for an isotropic GWB, using the optimal filter method for power law GWBs with $\alpha = \{0, 2/3, 3\}$. For each of the three baselines $IJ$, we show the point estimate and $1\sigma$ uncertainty for the cross-correlation estimate $C_{IJ}$, along with the frequency band from 20 Hz to $f_{99\%}^{IL}$ containing 99% of the sensitivity. We see that the HL baseline is the most sensitive, and the HV and LV baselines are more sensitive at higher frequencies, and for larger spectral indices, due to the longer baseline. In the last two columns, we also present the search result combining all three baselines from O3, as well as the O1 and O2 data. As noted in the main text, the point estimates for the HV and LV are approximately 2$\sigma$ away from zero, however this is not consistent with a GWB given the result of the much more sensitive HL baseline.

| Power law | $f_{99\%}^{IL}$ [Hz] | $\hat{C}_{IL}^{HL}/10^{-9}$ | $f_{99\%}^{HV}$ [Hz] | $\hat{C}_{IL}^{HV}/10^{-9}$ | $f_{99\%}^{LV}$ [Hz] | $\hat{C}_{IL}^{LV}/10^{-9}$ | $f_{99\%}^{O1+O2+O3}$ [Hz] | $\hat{C}_{IL}^{O1+O2+O3}/10^{-9}$ |
|-----------|----------------------|-----------------------------|----------------------|-----------------------------|----------------------|-----------------------------|-----------------------------|-----------------------------|
| 0         | 76.1                 | $-2.1 \pm 8.2$              | 97.7                 | $229 \pm 98$               | 88.0                 | $-134 \pm 63$              | 76.6                        | $1.1 \pm 7.5$               |
| 2/3       | 90.2                 | $-3.4 \pm 6.1$              | 117.8                | $145 \pm 60$               | 107.3                | $-82 \pm 40$               | 90.6                        | $-0.2 \pm 5.6$              |
| 3         | 282.8                | $-1.3 \pm 0.9$              | 375.8                | $9.1 \pm 4.1$              | 388.0                | $-4.9 \pm 3.1$             | 291.6                       | $-0.6 \pm 0.8$              |
natural choice for a scaling parameter, and also is more sensitive to small signals. However, since upper limits computed with a uniform prior are more conservative, we present results for the uniform prior as well. For both cases, we choose the upper bound of the prior to be large enough that there is no posterior support at the upper end of the prior range. For the log uniform prior, the upper limit depends mildly on the lower bound of the prior range, which cannot be taken to be zero. Following [44], we choose the lower bound to be $\Omega_{\text{ref}} \geq 10^{-13}$. This choice enables a direct comparison with previous upper limits, and is the same order of magnitude as the expected reach of next-generation ground-based detectors [106–108].

For the spectral index, we compute upper limits by fixing $\alpha$ to the three values $\{0, 2/3, 3\}$ discussed earlier, as well as allowing $\alpha$ to vary. For the latter case, we assume a Gaussian prior on $\alpha$ with zero mean and standard deviation 3.5. This prior on $\alpha$ is very similar to the triangular prior on $\alpha$ we used in the O2 analysis [44], however it does not vanish for large values of $|\alpha|$. Therefore in principle, this prior allows us to probe extreme power laws if the data support them. We have checked that the Gaussian prior gives posterior distributions that are nearly identical to those produced using the triangular prior.

We marginalize over calibration uncertainty following the methods in [109]. We use an amplitude calibration uncertainty of 7.0% for Hanford, 6.4% for Livingston, and 5% for Virgo [110]; this is a conservative choice describing the worst case over the entire run. We use the same amplitude calibration uncertainty factors for O1 and O2 as in previous analyses [44]. In all cases, phase uncertainty is negligible. The results are given in Table II. We also show the posterior in the $\Omega_{\text{ref}}$-$\alpha$ plane in Fig. 4.

At the 95% credible level, using a log-uniform (uniform) prior, we find that $\Omega_{\text{GW}}(25 \text{ Hz})$ is less than $5.8 \times 10^{-9}$ ($1.7 \times 10^{-8}$) for $\alpha = 0$, $3.4 \times 10^{-9}$ ($1.2 \times 10^{-8}$) for $\alpha = 2/3$, $3.9 \times 10^{-10}$ ($1.3 \times 10^{-9}$) for $\alpha = 3$, and $6.6 \times 10^{-9}$ ($2.7 \times 10^{-8}$) when marginalizing over $\alpha$. This represents an improvement by a factor of about 6.0 (3.6) for a flat power law, 8.8 (4.0) for a power law of $\alpha = 2/3$, and 13.1 (5.9) for a power law of $\alpha = 3$. The improvement for large $\alpha$ is due in part to the improved high-frequency sensitivity of Advanced LIGO in O3; to the addition of the baselines involving Virgo; and to the specific noise realization, in particular the negative point estimate $\alpha = 3$ in O3, as seen in Table I. We find a log$_{10}$ Bayes Factor of $-0.3$ when comparing the hypotheses of signal and noise to noise-only when marginalizing over $\alpha$.

### B. Non-GR polarizations

We can use our results to constrain modifications to GR by using the SVT-PL model defined in Sec. II C. This analysis benefits from the inclusion of Virgo data, since adding more detectors to the network can help distinguish between different polarizations, as shown in [46]. We note that $\Omega_{\text{GW}}$ does not necessarily have the interpretation of an energy density in modified theories of gravity, and it is in general more appropriate to think of these quantities as a measure of the strain power in each polarization [111].

We use the log-uniform prior on each strength $\Omega_{\text{ref}}^{(p)}$, and the Gaussian prior for each spectral index $\alpha_p$, as described

| $\alpha$ | Uniform prior | Log-uniform prior |
|---------|---------------|-------------------|
|         | O3 | O2 [44] | Improvement | O3 | O2 [44] | Improvement |
| 0       | $1.7 \times 10^{-8}$ | $6.0 \times 10^{-8}$ | 3.6 | $5.8 \times 10^{-9}$ | $3.5 \times 10^{-8}$ | 6.0 |
| 2/3     | $1.2 \times 10^{-8}$ | $4.8 \times 10^{-8}$ | 4.0 | $3.4 \times 10^{-9}$ | $3.0 \times 10^{-8}$ | 8.8 |
| 3       | $1.3 \times 10^{-9}$ | $7.9 \times 10^{-9}$ | 5.9 | $3.9 \times 10^{-10}$ | $5.1 \times 10^{-9}$ | 13.1 |
| Marg.   | $2.7 \times 10^{-8}$ | $1.1 \times 10^{-7}$ | 4.1 | $6.6 \times 10^{-9}$ | $3.4 \times 10^{-8}$ | 5.1 |
in the previous section. We show the results in Table III. Marginalizing over the spectral indices for each polarization, we find that the upper limit on a scalar-polarized GWB in this model is $\Omega_{GW}^{(S)} (25 \, \text{Hz}) \leq 2.1 \times 10^{-8}$, the limit on a vector GWB is $\Omega_{GW}^{(V)} (25 \, \text{Hz}) \leq 7.9 \times 10^{-9}$, and the limit on a tensor GWB is $\Omega_{GW}^{(T)} (25 \, \text{Hz}) \leq 6.4 \times 10^{-9}$. Note that the upper limit on tensor modes in this analysis is slightly different from the upper limit when we consider only GR modes given in the previous section, because of the inclusion of additional parameters under this current model.

We compute two Bayesian odds ratios $O$: the odds $O_{N}^{SIG}$ that a stochastic background of any polarization is present (SIG) versus Gaussian noise (N), and the odds $O_{N}^{NGR}$ that alternative polarization modes are present (NGR) versus standard tensor polarizations obeying GR. Note that the signal hypothesis is formally the union of seven distinct sub-hypotheses, corresponding to the seven possible combinations of tensor, vector, and scalar modes (T, TV, TS, TVS, etc.) When computing $O_{N}^{SIG}$, we adopt equal prior odds between the signal and noise hypotheses, and within

\begin{table}[h]
\centering
\begin{tabular}{|c|c|c|}
\hline
(Polarization) & O3 & O2 \cite{44} & Improvement \\
\hline
Tensor & $6.4 \times 10^{-9}$ & $3.2 \times 10^{-8}$ & 5.0 \\
Vector & $7.9 \times 10^{-9}$ & $2.9 \times 10^{-8}$ & 3.7 \\
Scalar & $2.1 \times 10^{-8}$ & $6.1 \times 10^{-8}$ & 2.9 \\
\hline
\end{tabular}
\caption{Upper limits at the 95\% credible level on $\Omega_{ref}$ for scalar, vector, and tensor polarizations, along with the improvement of the O3 result over the previous result from O2. We use the log-uniform prior for $\Omega_{ref}$ and a Gaussian prior on the spectral index for each polarization, as described in the main text.}
\end{table}

the signal hypothesis assign equal priors among the various signal sub-hypotheses. Similarly, in computing $O_{N}^{NGR}$ we choose equal priors between NGR and GR hypotheses, and within NGR assign equal priors to the six distinct ways that non-standard polarizations might be present; see Refs. \cite{46} and \cite{47} for more details. We find $\log_{10} O_{N}^{SIG} = -0.4$ and $\log_{10} O_{N}^{NGR} = -0.2$, confirming that we have no evidence for a stochastic background alternative gravitational-wave polarizations.

C. Joint fit for GWB and magnetic noise

We extend the standard analysis to do a joint fit allowing for both a GWB with an arbitrary power-law index, as well as an apparent GWB arising from correlated magnetic noise. While we have already seen that correlated magnetic noise is below the O3 sensitivity in Sec. III D, the analysis presented here is complementary because it allows us to simultaneously fit for the presence of both a GWB of astrophysical origin and a correlated magnetic noise component. In future runs, this kind of joint fit will become increasingly important. We use the method described in \cite{62}.

We evaluate whether correlated magnetic noise is detected by first constructing a likelihood function that includes a model for both the correlated magnetic noise and a power-law GWB, $\Omega_{M}(f, \Theta) = \Omega_{PL}(f, \Theta_{PL}) + \Omega_{MAG}(f, \Theta_{MAG})$. Our model $\Omega_{MAG}(f, \Theta_{MAG})$ takes the same form as Eq. (11). However, rather than use the coupling functions measured using magnetic-field injections, we model the coupling functions as power laws, which approximate the frequency dependence of the measurements. The vector $\Theta_{MAG}$ contains the parameters of the model for the coupling functions $T_{I,J}(f)$, which we take to be a simple power law

$$|T_{I,J}(f)| = \kappa_{I} \left(\frac{f}{10 \, \text{Hz}}\right)^{-\beta_{I}}.$$  \hspace{1cm} (12)

The parameters for the power law GWB are the strength $\Omega_{ref}$ and spectral index $\alpha$. We use nested sampling to estimate the model evidences for three separate models: N, MAG, and PL + MAG, using the notation defined in Sec. II C.

Our prior distribution for the magnitude $\kappa_{I}$ is log uniform from $10^{-25}$ to $10^{-22}$ pT$^{-1}$ for all of the detectors. Our prior on the spectral index $\beta_{I}$ is uniform from $\beta_{I}^{\min}$ to $\beta_{I}^{\max}$, the minimum and maximum values of the spectral index for the magnetic coupling measured at detector $I$ during the O3 run. For Hanford, Livingston and Virgo, the $\beta$ priors chosen for the study are (0, 12), (1, 10) and (0, 7), respectively. The chosen prior range is large enough to encompass all measured coupling function measurements in O3, including the uncertainties mentioned in Sec. III. We find $\log_{10} O_{N}^{MAG} = -0.03$, which indicates that there is no preference for a model with correlated magnetic noise.
compared to a model with only uncorrelated Gaussian noise. We also consider a model with a power-law GWB present, using the log-uniform prior on $\Omega_{\text{ref}}$ and Gaussian prior on $\alpha$ as in Sec. IV.A. We find that the Bayes factor between a model with correlated GWB and magnetic noise, to a model with only uncorrelated Gaussian noise, is $\log_{10} F_{\text{mag+pl}}^{\text{R}} = -0.3$, confirming that there is no evidence of a GWB in the data.

V. IMPLICATIONS FOR COMPACT BINARIES

With upper limits on the GWB in hand, we now explore the implications of these results for the GWB due to CBCs. We first compare our upper limits to updated predictions for the energy-density due to CBC sources. We then combine our limits with the direct detections of CBCs in the local Universe to constrain the merger rate of compact binaries at large redshifts.

A. Fiducial model

Observations from O3a have significantly increased our knowledge of the compact binary population [68,69,71–73,76]. Here, we update the fiducial model of the GWB due to compact binaries [44,63,64,96] in accordance with the latest observational and theoretical advances. The energy-density spectrum due to a particular source class $k$ is

$$\Omega_k(f) = \frac{f}{\rho_c} \int_{0}^{z_{\text{max}}} dz \frac{R_k(z) \langle (dE_s / df_s)_{|f_s} \rangle_k}{(1 + z) H(z)},$$

(13)

where $R_k(z)$ is the source-frame merger rate per comoving volume of objects of class $k$ and $H(z) = H_0 \sqrt{\Omega_m (1 + z)^3 + \Omega_{\Lambda}}$ is the Hubble parameter, where $\Omega_m$ is the fraction of the critical energy density $\rho_c$ contained in matter and $\Omega_{\Lambda}$ the fraction contained in the cosmological constant; we take $\Omega_m = 1 - \Omega_{\Lambda} = 0.3065$ [77]. The quantity $\langle dE_s / df_s \rangle_{|f_s}$ is the source-frame energy radiated by a single source, evaluated at the source frequency $f_s = f(1 + z)$ and averaged over the ensemble properties of the given class $k$:

$$\left\langle \frac{dE_s}{df_s} \right\rangle_k = \int d\phi p_k(\phi) \frac{dE_s}{df_s}(\phi),$$

(14)

where $p_k(\phi)$ is the probability distribution of source parameters $\phi$ (e.g., masses, spins, etc.) across class $k$.

We consider here three classes of compact binaries: binary black holes (BBHs), binary neutron stars (BNSs), and neutron-star–black-holes (NSBHs). Except where otherwise stated, we use the same choices for $dE_s / df_s$, $p_k(\phi)$, and $R_k(z)$ as in [44]. We note that there are several important astrophysical uncertainties which are not included in our fiducial model, which could potentially have an impact on our predictions. These include the possibility that the initial mass function can lead to a lower number of neutron stars than what we assume [112]; indications that the star formation rate may peak at a smaller redshift [113]; and uncertainty in the metallicity evolution.

1. Binary black holes.

We assume that BBH formation follows a metallicity-weighted star formation rate (SFR) with a distribution $\rho(t_d) \sim t_d^{-1}$ of time delays $t_d$ between binary formation and merger, where $50 \text{ Myr} \leq t_d \leq 13.5 \text{ Gyr}$. We take the SFR from Ref. [114], and multiply it by the fraction of stellar formation occurring at metallicities $Z < Z_{\text{thresh}}$ [115]. In Ref. [44], we adopted $Z_{\text{thresh}} = 0.5 Z_\odot$, and applied this threshold only to black holes above $30 M_\odot$. Here, we adopt a more stringent cutoff $Z_{\text{thresh}} = 0.1 Z_\odot$ [116,117]. Moreover, we apply this weighting across the entire mass spectrum, as recent population synthesis studies suggest that the mass spectrum of BH mergers does not evolve appreciably with redshift [117].

We additionally update our assumptions regarding the mass and spin distributions of BBHs. In Ref. [44], we assumed that BBHs had aligned dimensionless spin magnitudes distributed uniformly between $-1$ to $1$. It now appears, though, that the BBH population exhibits small effective spins [76,118], and so when computing $\Omega_{\text{BBH}}(f)$ we now assume that BBHs had negligibly small spin. We also adopt a close variant of the broken power law model of Ref. [76] to describe the mass distribution of BBHs (for convenience we assume a sharp low-mass cutoff in the BBH mass spectrum, corresponding to $\delta_m \to 0$ in Eq. (B6) of [76]). We do not assume fixed values for the parameters of this model, but include our uncertainty on the BBH mass spectrum as an additional systematic uncertainty in our estimate of $\Omega_{\text{BBH}}(f)$. To achieve this, we use GWTC-2 [68] to hierarchically compute a joint posterior on the mass distribution and local merger rate of BBHs, given the assumed redshift distribution described above. Hierarchical inference is performed following the method discussed in Ref. [76]. By evaluating Eq. (13) across the resulting ensemble of posterior samples, we subsequently obtain a probability distribution on the energy-density spectrum $\Omega_{\text{BBH}}(f)$ due to BBH mergers, given our knowledge of the local population.

Our updated estimate of $\Omega_{\text{BBH}}(f)$ is shown in green in Fig. 5. We find $\Omega_{\text{BBH}}(25 \text{ Hz}) = 4.7^{+1.6}_{-1.4} \times 10^{-10}$. This uncertainty includes the standard Poisson uncertainty on the local merger rate, which we find to be $R_{\text{BBH}}(0) = 19_{-8}^{+18} \text{ Gpc}^{-3} \text{ yr}^{-1}$ (median and symmetric 90% credible interval) given our fiducial redshift distribution above. This rate estimate matches that obtained in Ref. [76] when agnostically allowing the merger rate to evolve with redshift, although in general estimates of $R_{\text{BBH}}(0)$ may differ under different presumed redshift distributions. Our
estimate of $\Omega_{\text{BBH}}(25\;\text{Hz})$ also reflects, though, the additional systematic uncertainty on $\Omega_{\text{BBH}}(f)$ due to imperfect knowledge of the BBH mass distribution. This uncertainty on the mass distribution is, for example, responsible for the larger uncertainty in $\Omega_{\text{BBH}}(f)$ at high frequencies.

### 2. Binary neutron stars.

As in [44], we assume that the rate of BNS progenitor formation is proportional to the rate of star formation [114] and that the distribution of time delays $t_d$ between their formation and merger is of the form $p(t_d) \propto t_d^{-1}$ between 20 Myr $\leq t_d \leq 13.5$ Gyr. The detection of a second binary neutron star merger in O3a, GW190425 [70], has decreased uncertainty on the BNS merger rate and demonstrated that at least some neutron star mergers contain significantly heavier masses than expected. Following [76], we assume a uniform distribution of component masses between 1–2.5 $M_\odot$, which yields an estimated present-day merger rate of $R_{\text{BNS}}(0) = 320^{+490}_{-240} \; \text{Gpc}^{-3} \; \text{yr}^{-1}$. When modeling $\Omega_{\text{BNS}}(f)$, we consider the energy radiated during the inspiral phase only, truncating the BNS energy spectra $\frac{dE}{df}(\phi)$ at frequencies corresponding to the innermost stable circular orbit. Our estimate of the BNS GWB is shown in red in Fig. 5. We find $\Omega_{\text{BNS}}(25\;\text{Hz}) = 2.0^{+3.2}_{-1.4} \times 10^{-10}$.

### 3. Neutron star-black hole binaries.

To date, Advanced LIGO and Virgo have made no confirmed detections of neutron star-black hole (NSBH) mergers. Two events, GW190814 and the low-significance candidate GW190426_152155, have secondary masses constrained below 3 $M_\odot$ with primary masses above 3 $M_\odot$ and so are possibly consistent with NSBH systems, but their true physical natures remain unknown [68,73]. In order to forecast the possible contribution of NSBH mergers to the GWB, we therefore use the upper limit on the NSBH merger rate previously adopted in Ref. [44], again assuming a delta-function mass distribution at $10 \; M_\odot + 1.4 \; M_\odot$. We estimate $\Omega_{\text{NSBH}}(f)$ using the same redshift distribution as adopted for BBH mergers, and include contributions from the complete inspiral, merger, and ringdown. This likely results in an overestimate of $\Omega_{\text{NSBH}}(f)$ at high frequencies, since some fraction of NSBH inspirals are expected to end in tidal disruption of the neutron star companion [119–121]. The resulting upper limit on $\Omega_{\text{NSBH}}(f)$ is shown as a dashed black line in Fig. 5, with $\Omega_{\text{NSBH}}(25\;\text{Hz}) \leq 8.4 \times 10^{-10}$.

### 4. Total CBC GWB.

In the right-hand side of Fig. 5 we present an updated estimate of the combined GWB due to BBH and BNS mergers. Under our model, we predict this combined background to be $\Omega_{\text{BBH+BNS}}(25\;\text{Hz}) = 6.8^{+3.6}_{-2.2} \times 10^{-10}$. Combining the upper limit on $\Omega_{\text{NSBH}}(f)$ with the upper 95% credible bound on the contributions from BBH and BNS mergers, we bound the total expected GWB to be $\Omega_{\text{Total}}(25\;\text{Hz}) \leq 1.9 \times 10^{-9}$. We also show the $2\sigma$ power-law integrated (PI) curves [99] indicating the integrated sensitivity of the O3 search [99], along with projections for...
2 years of the Advanced LIGO-Virgo network at design sensitivity, and the envisioned A + design sensitivity after 2 years, assuming a 50% duty cycle. We use the power spectra available from [122,123]. Previous work has shown that the residual background obtained after subtracting resolvable signals is expected to be within 10% of the total background for Advanced LIGO and Virgo at design sensitivity, and approximately a factor of 2 smaller for the A + detectors [63,106]. These curves indicate that by the time the detectors reach the A + design sensitivity, much of the expected parameter space of the compact binary GWB will be accessible by ground-based detectors. The continued addition of new instruments to the worldwide detector network, like KAGRA [124] and LIGO-India [125], is expected to further improve upon our projected sensitivity. For coincident and co-aligned detectors with the same power spectral density, the improvement in sensitivity should scale like the square root of the number of baselines, or equivalently with the number of detectors in the network. However, in practice one must account for the different overlap reduction functions for each baseline, as well as for different sensitivities. Reference [99] gives a prescription for a general set of detectors. Figure 9 of that paper indicates that the improvement is likely to be minimal for small power laws, and becomes larger for large power laws.

### B. Constraining the BBH merger rate

The energy-density spectra in Fig. 5 show our current best estimates for the GWB under an astrophysically plausible model for the rate density $R_{\text{BBH}}(z)$ of BBH mergers of stellar origin. By combining direct detections of compact binaries with upper limits on the GWB, however, we can alternatively seek to directly measure $R_{\text{BBH}}(z)$. Knowledge of the BBH redshift distribution, and in particular the redshift at which $R_{\text{BBH}}(z)$ is at a maximum, offers a potential measure of the mean evolutionary time delay between binary formation and merger, the branching ratios between competing formation channels, or even the underlying star formation rate [126,127]. Although the measurement of $R_{\text{BBH}}(z)$ is made difficult by the limited range of present ground-based detectors, we can nevertheless make progress by combining direct BBH detections with upper limits on the GWB [75].

Here, we update the constraints on the rate evolution of BBHs from [75], using our latest O3 limits on the GWB and the expanded GWTC-2 catalog of BBH detections. We again assume a broken power law form for the mass distribution of BBH mergers, but now adopt a phenomenologically-parametrized form

$$R_{\text{BBH}}(z) = C(\lambda_1, \lambda_2, z_{\text{peak}}) R_0 (1 + z)^{\lambda_1} \left[ 1 + \left( \frac{z}{z_{\text{peak}}} \right)^{\lambda_2} \right]$$  \hspace{1cm} (15)$$

for their merger rate density. Under this form, the merger rate evolves as $R_{\text{BBH}}(z) \propto (1 + z)^{\lambda_1}$ at $z \lesssim z_{\text{peak}}$ and $R_{\text{BBH}} \approx (1 + z)^{-\lambda_2}$ at $z \gtrsim z_{\text{peak}}$, and at low redshifts $\lambda_1$ can be identified with the parameter $\kappa$ of Ref. [76]. The normalization constant $C(\lambda_1, \lambda_2, z_{\text{peak}})$ is defined such that $R_0$ is the local merger rate density of BBHs at $z = 0$.

Using the direct BBH detections from GWTC-2 along with the updated GWB search results presented here, we jointly infer the parameters governing both the mass and redshift distributions of BBH mergers. We adopt the factorized likelihood from Eq. (9), given by the product between the standard GWB likelihood $p(C^{(\ell)}(f_s) | \Theta_{\text{BBH}})$ under our model for the BBH background, and the likelihood $p(d | \Theta_{\text{BBH}})$ of having measured data $d$ associated with the 44 direct BBH detections in GWTC-2 with false alarm rates $< 1$ yr$^{-1}$. This likelihood $p(d | \Theta_{\text{BBH}})$ for direct detections is evaluated using posterior samples on the parameters of each individual event, as described further in Sec. 4 of [76]. The direct detection likelihood also corrects for selection biases, such as LIGO and Virgo’s higher detection efficiency for higher-mass systems; we evaluate selection effects using the same injection campaign discussed in [76]. Our priors are uniform on $\lambda_1$, $\lambda_2$, and $z_{\text{peak}}$, and log-uniform on $R_0$.

The resulting constraints on the BBH merger rate as a function of redshift are shown in Fig. 6. Each blue trace represents a single draw from our posterior on the BBH mass distribution and merger rate history. The black curve marks the median estimated merger rate at a given redshift, while solid grey curves mark our central 90% credible bound. From O1 and O2 data, the nondetection of the GWB served to constrain the BBH merger rate to less than $\sim 10^3$ Gpc$^{-3}$ yr$^{-1}$ beyond $z \approx 2$ at 90% credibility [75].
With continued data collection, however, the nondetection (or eventual detection) of the GWB may again offer informative constraints on $\lambda_1$ and $z_{\text{peak}}$. As additional direct BBH detections are made, our knowledge of $\lambda_1$ will continue to improve, identifying an increasingly narrow, nearly-vertical contour in the $\lambda_1 - z_{\text{peak}}$ plane. Continued time integration in searches for the GWB, meanwhile, will exclude a growing fraction of this plane, ruling out large values of both $\lambda_1$ and $z_{\text{peak}}$. In Fig. 7, for example, we show projected exclusion contours corresponding to one year of integration with Advanced LIGO and Virgo, at both their design sensitivity and A+ configurations; both exclusion curves extend into the presently allowed values of $\lambda_1$, where they may again be informative and break the degeneracy between $\lambda_1$ and $z_{\text{peak}}$.

**VI. CONCLUSIONS**

In this work, we have performed a search for an isotropic GWB using data from Advanced LIGO’s and Virgo’s first three observing runs. Since we did not find evidence for a background of astrophysical origin, we placed upper limits, improving previous bounds by about a factor of 6.0 for a flat background.

We considered the implications of the results, and by combining the upper limits with measurements from GWTC-2 we have constrained the BBH merger rate as a function of redshift. Our results can be used to constrain additional models such as cosmic strings [128] or phase transitions [129], using the cross correlation spectra we have made publicly available [58]. Our results can also be combined with other measurements of the GWB at other frequencies [82]. Although in this work we focused on searching for an isotropic GWB, in a companion paper we present a search for an anisotropic GWB using data from LIGO and Virgo’s third observing run [130].

Moving forward, we expect currently proposed ground-based facilities such as A+ have the potential to probe a large range of the model space for CBC backgrounds. In order to make full use of the data and confidently claim a detection, it will be important to further develop the methods to handle correlated terrestrial noise sources, such as the magnetic couplings described here.

**ACKNOWLEDGMENTS**

This material is based upon work supported by NSF’s LIGO Laboratory which is a major facility fully funded by the National Science Foundation. The authors also gratefully acknowledge the support of the Science and Technology Facilities Council (STFC) of the United Kingdom, the Max-Planck-Society (MPS), and the State of Niedersachsen/Germany for support of the construction of Advanced LIGO and construction and operation of the GEO600 detector. Additional support for Advanced LIGO was provided by the Australian Research Council.
authors gratefully acknowledge the Italian Istituto Nazionale di Fisica Nucleare (INFN), the French Centre National de la Recherche Scientifique (CNRS) and the Netherlands Organization for Scientific Research, for the construction and operation of the Virgo detector and the creation and support of the EGO consortium. The authors also gratefully acknowledge research support from these agencies as well as by the Council of Scientific and Industrial Research of India, the Department of Science and Technology, India, the Science & Engineering Research Board (SERB), India, the Ministry of Human Resource Development, India, the Spanish Agencia Estatal de Investigación, the Vicepresidéncia i Conselleria d’Innovació, Recerca i Turisme and the Conselleria d’Educació i Universitat del Govern de les Illes Balears, the Conselleria d’Innovació, Universitats, Ciència i Societat Digital de la Generalitat Valenciana and the CERCA Programme Generalitat de Catalunya, Spain, the National Science Centre of Poland and the Foundation for Polish Science (FNP), the Swiss National Science Foundation (SNSF), the Russian Foundation for Basic Research, the Russian Science Foundation, the European Commission, the European Regional Development Funds (ERDF), the Royal Society, the Scottish Funding Council, the Scottish Universities Physics Alliance, the Hungarian Scientific Research Fund (OTKA), the French Lyon Institute of Origins (LIO), the Belgian Fonds de la Recherche Scientifique (FRS-FNRS), Actions de Recherche Concertées (ARC) and Fonds Wetenschappelijk Onderzoek Vlaanderen (FWO), Belgium, the Paris Île-de-France Region, the National Research, Development and Innovation Office Hungary (NKFIH), the National Research Foundation of Korea, the Natural Science and Engineering Research Council Canada, Canadian Foundation for Innovation (CFI), the Brazilian Ministry of Science, Technology, and Innovations, the International Center for Theoretical Physics South American Institute for Fundamental Research (ICTP-SAIIFR), the Research Grants Council of Hong Kong, the National Natural Science Foundation of China (NSFC), the Leverhulme Trust, the Research Corporation, the Ministry of Science and Technology (MOST), Taiwan, the United States Department of Energy, and the Kavli Foundation. The authors gratefully acknowledge the support of the NSF, STFC, INFN and CNRS for provision of computational resources. This work was supported by MEXT, JSPS Leading-edge Research Infrastructure Program, JSPS Grant-in-Aid for Specially Promoted Research 26000005 (Kajita 2014-2018), JSPS Grant-in-Aid for Scientific Research on Innovative Areas 2905: JP17H06358, JP17H06361 and JP17H06364, JSPS Core-to-Core Program A. Advanced Research Networks, JSPS Grant-in-Aid for Scientific Research (S) 17H06133 and 20H05639, JSPS Grant-in-Aid for Transformative Research Areas (A) 20A203: JP20H05854, the joint research program of the Institute for Cosmic Ray Research, University of Tokyo, National Research Foundation (NRF) and Computing Infrastructure Project of KISTI-OSDC in Korea, Academia Sinica (AS), AS Grid Center (ASGC) and the Ministry of Science and Technology (MoST) in Taiwan under grants including AS-CDA-105-M06, Advanced Technology Center (ATC) of NAOJ, Mechanical Engineering Center of KEK. All plots have been prepared using Matplotlib [131]. We would like to thank all of the essential workers who put their health at risk during the COVID-19 pandemic, without whom we would not have been able to complete this work. This document has been assigned the number LIGO-DCC-P2000314.

[1] N. J. Cornish and J. D. Romano, Phys. Rev. D 92, 042001 (2015).
[2] P. A. Rosado, Phys. Rev. D 84, 084004 (2011).
[3] X.-J. Zhu, E. Howell, T. Regimbau, D. Blair, and Z.-H. Zhu, Astrophys. J. 739, 86 (2011).
[4] S. Marassi, R. Schneider, G. Corvino, V. Ferrari, and S. P. Zwart, Phys. Rev. D 84, 124037 (2011).
[5] C. Wu, V. Mandic, and T. Regimbau, Phys. Rev. D 85, 104024 (2012).
[6] X.-J. Zhu, E. J. Howell, D. G. Blair, and Z.-H. Zhu, Mon. Not. R. Astron. Soc. 431, 882 (2013).
[7] A. Buonanno, G. Sigl, G. G. Raffelt, H.-T. Janka, and E. Muller, Phys. Rev. D 72, 084001 (2005).
[8] E. Howell, D. Coward, R. Burman, D. Blair, and J. Gilmore, Mon. Not. R. Astron. Soc. 351, 1237 (2004).
[9] P. Sandick, K. A. Olive, F. Daigne, and E. Vangioni, Phys. Rev. D 73, 104024 (2006).
[10] S. Marassi, R. Schneider, and V. Ferrari, Mon. Not. R. Astron. Soc. 398, 293 (2009).
[11] X.-J. Zhu, E. Howell, and D. Blair, Mon. Not. R. Astron. Soc. 409, L132 (2010).
[12] V. Ferrari, S. Matarrese, and R. Schneider, Mon. Not. R. Astron. Soc. 303, 258 (1999).
[13] T. Regimbau and J. A. de Freitas Pacheco, Astron. Astrophys. 376, 381 (2001).
[14] E. Howell, T. Regimbau, A. Corsi, D. Coward, and R. Burman, Mon. Not. R. Astron. Soc. 410, 2123 (2011).
[15] X.-J. Zhu, X.-L. Fan, and Z.-H. Zhu, Astrophys. J. 729, 59 (2011).
[16] S. Marassi, R. Ciolfi, R. Schneider, L. Stella, and V. Ferrari, Mon. Not. R. Astron. Soc. 411, 2549 (2011).
[17] P. A. Rosado, Phys. Rev. D 86, 104007 (2012).
[18] C.-J. Wu, V. Mandic, and T. Regimbau, Phys. Rev. D 87, 042002 (2013).
R. ABBOTT et al.

PHYS. REV. D 104, 022004 (2021)

[83] http://www.lemisensors.com.
[84] https://www.geo-metrонi.de/mtxgeo/index.php/mfs-06e-overview.
[85] https://git.ligo.org/stochastic-public/stochastic/.
[86] MATLAB, 9.8.0.1323502 (R2020a) (The MathWorks Inc., Natick, Massachusetts, 2020).
[87] C. Biwer et al., Phys. Rev. D 95, 062002 (2017).
[88] B. P. Abbott et al. (LIGO Scientific and Virgo Collaborations), Classical Quantum Gravity 35, 065010 (2018).
[89] D. Davis et al., arXiv:2101.11673.
[90] P. Covas et al. (LSC Instrument Authors), Phys. Rev. D 97, 082002 (2018).
[91] P. Nguyen et al., Environmental noise in Advanced LIGO detectors, Classical Quantum Gravity 38, 145001 (2021).
[92] W. Schumann, Z. Naturforsch. 7A, 250 (1952).
[93] M. C. Araya,1 J. M. Ezquiaga and D. E. Holz, Astrophys. J. Lett.
[94] T. Regimbau, Res. Astron. Astrophys.
[95] E. Thrane, N. Christensen, R. Schofield, Phys. Rev. D 87, 123009 (2013).
[96] B. P. Abbott et al. (LIGO Scientific and Virgo Collaborations), Phys. Rev. Lett. 118, 121101 (2017).
[97] K. Merfeld et al., aLIGO LHO Logbook, https://alog.ligo-wa.caltech.edu/aLOG/index.php?callRep=48212.
[98] P. Nguyen et al., aLIGO LHO Logbook, https://alog.ligo-wa.caltech.edu/aLOG/index.php?callRep=57672 ().
[99] E. Thrane and J. D. Romano, Phys. Rev. D 88, 124032 (2013).
[100] P. Nguyen et al., aLIGO LHO Logbook, https://alog.ligo-wa.caltech.edu/aLOG/index.php?callRep=43406 ().
[101] I. Michaloliakos et al., aLIGO LHO Logbook, https://alog.ligo-wa.caltech.edu/aLOG/index.php?callRep=56295.
[102] T. Regimbau, Res. Astron. Astrophys. 11, 369 (2011).
[103] C. Périgois, C. Belczynski, T. Bulik, and T. Regimbau, Phys. Rev. D 103, 043002 (2021).
[104] J. M. Ezquiaga and D. E. Holz, Astrophys. J. Lett. 909, L23 (2021).
[105] P. Sandick, K. A. Olive, F. Daigne, and E. Vangioni, Phys. Rev. D 73, 104024 (2006).
[106] T. Regimbau, M. Evans, N. Christensen, E. Katsavounidis, B. Sathyaprakash, and S. Vitale, Phys. Rev. Lett. 118, 151105 (2017).
[107] S. Sachdev, T. Regimbau, and B. Sathyaprakash, Phys. Rev. D 102, 024051 (2020).
[108] K. Martinovic, P. M. Meyers, M. Sakellariadou, and N. Christensen, Phys. Rev. D 103, 043023 (2021).
[109] J. T. Whelan, E. L. Robinson, J. D. Romano, and E. H. Thrane, J. Phys. Conf. Ser. 484, 012027 (2014).
[110] L. Sun et al., Classical Quantum Gravity 37, 225008 (2020).
[111] M. Isi and L. C. Stein, Phys. Rev. D 98, 104025 (2018).
[112] M. Chruslinska, T. Jerábková, G. Nelemans, and Z. Yan, Astron. Astrophys. 636, A10 (2020).
[113] R. López Fernández, R. M. González Delgado, E. Pérez, R. García-Benito, R. Cid Fernandes, W. Schoenell, S. F. Sánchez, A. Gallazzi, P. Sánchez-Blázquez, N. Vale Asari, and C. J. Wacher, Astron. Astrophys. 615, A27 (2018).
[114] E. Vangioli, K. A. Olive, T. Prestegard, J. Silk, P. Petitjean, and V. Mandic, Mon. Not. R. Astron. Soc. 447, 2575 (2015).
[115] N. Langer and C. A. Norman, Astrophys. J. Lett. 638, L60 (2006).
[116] M. Chruslinska, G. Nelemans, and K. Belczynski, Mon. Not. R. Astron. Soc. 482, 5012 (2019).
[117] M. Mapelli, N. Giacobbo, F. Santoliquido, and M. C. Artale, Mon. Not. R. Astron. Soc. 487, 2 (2019).
[118] S. Miller, T. A. Callister, and W. Farr, Astrophys. J. 895, 128 (2020).
[119] F. Foucart, M. D. Duez, L. E. Kidder, and S. A. Teukolsky, Phys. Rev. D 83, 024005 (2011).
[120] K. Kyutoku, M. Shibata, and K. Taniguchi, Phys. Rev. D 82, 044049 (2010).
[121] K. Kawaguchi, K. Kyutoku, H. Nakano, H. Okawa, M. Shibata, and K. Taniguchi, Phys. Rev. D 92, 024014 (2015).
[122] B. Abbott et al. (KAGRA, LIGO Scientific, and VIRGO Collaborations), Living Rev. Relativity 21, 3 (2018).
[123] L. Barsotti, P. Fritschel, M. Evans, and S. Gras, https://dcc.ligo.org/T1800044-v5/public ().
[124] T. Akutsu, M. Ando, K. Arai, Y. Arai, S. Araki et al., arXiv:2005.05574.
[125] B. Iyer, T. Souradeep, C. S. Unnikrishnan, S. Dhurandhar, S. Raja, A. Kumar, and A. Sengupta, LIGO-India Technical Report No. LIGO-M1100296, 2011.
[126] S. Vitale, W. M. Farr, K. K. Y. Ng, and C. L. Rodriguez, Astrophys. J. 886, L1 (2019).
[127] K. K. Y. Ng, S. Vitale, W. M. Farr, and C. L. Rodriguez, Astrophys. J. Lett. 913, L5 (2021).
[128] R. Abbott et al. (LIGO Scientific, Virgo, and KAGRA Collaborations), Phys. Rev. Lett. 126, 241102 (2021).
[129] R. Romero, K. Martinovic, T. Callister, H.-K. Guo, M. Martínez, M. Sakellariadou, F.-W. Yang, and Y. Zhao, Phys. Rev. Lett. 126, 151301 (2021).
[130] R. Abbott et al. (LIGO Scientific, Virgo, and KAGRA Collaborations), arXiv:2103.08520.
[131] J. D. Hunter, Comput. Sci. Eng. 9, 90 (2007).

R. Abbott,1 T. D. Abbott,2 S. Abraham,3 F. Acernese,4 K. Ackley,6 A. Adams,7 C. Adams,8 R. X. Adhikari,1 V. B. Adya,9 C. Affeldt,10,11 D. Agarwal,3 M. Agathos,12 K. Agatsuma,14 N. Aggarwal,15 O. D. Aguiar,16 L. Aiello,17,18,19 A. Ain,20,21 T. Akutsu,22,23 K. M. Aleman,24 G. Allen,26 A. Allocca,26,5 P. A. Altin,9 A. Amato,27 S. Anand,1 A. Ananyeva,1 S. B. Anderson,1 W. G. Anderson,25 M. Ando,29,30 S. V. Angelova,31 S. Ansoldi,32,33 J. M. Antelis,34 S. Antier,35 S. Appert,1 Koya Arai,36 Koji Arai,1 Y. Arai,36 S. Araki,37 A. Araya,38 M. C. Araya,1 J. S. Areeda,24 M. Arène,35 N. Aritomi,29 N. Arnaud,39,40 S. M. Aronson,25 H. Asada,43 Y. Asali,43 G. Ashton,6 Y. Aso,44 S. M. Aston,34 P. Astone,46 F. Aubin,47 P. Aufmuth,10,11 K. AultONeal,34 C. Austin,2 S. Babak,35 F. Badaracco,18,19 M. K. M. Bader,8 S. Bae,49 Y. Bae,50

022004-16
R. ABBOTT et al.  PHYS. REV. D 104, 022004 (2021)

151 Dipartimento di Scienze Matematiche, Fisiche e Informatiche, Università di Parma, I-43124 Parma, Italy
152 INFN, Sezione di Milano Bicocca, Gruppo Collegato di Parma, I-43124 Parma, Italy
153 Institut d’Astrophysique de Paris, Sorbonne Université & CNRS, UMR 7095, 98 bis bd Arago, F-75014 Paris, France
154 Physik-Institut, University of Zurich, Winterthurerstrasse 190, 8057 Zurich, Switzerland
155 Université de Strasbourg, CNRS, IPHC UMR 7178, F-67000 Strasbourg, France
156 West Virginia University, Morgantown, West Virginia 26506, USA
157 Montclair State University, Montclair, New Jersey 07043, USA
158 Colorado State University, Fort Collins, Colorado 80523, USA
159 Institute for Nuclear Research, Hungarian Academy of Sciences, Bem t’er 18/c, H-4026 Debrecen, Hungary
160 CNR-SPIN, c/o Università di Salerno, I-84084 Fisciano, Salerno, Italy
161 Scuola di Ingegneria, Università della Basilicata, I-85100 Potenza, Italy
162 Observatori Astronòmic, Universitat de València, E-46980 Paterna, València, Spain
163 The University of Utah, Salt Lake City, Utah 84112, USA
164 Kenyon College, Gambier, Ohio 43022, USA
165 Vrije Universiteit Amsterdam, 1081 HV, Amsterdam, Netherlands
166 Department of Astronomy, The University of Tokyo, Mitaka City, Tokyo 181-8588, Japan
167 Faculty of Engineering, Niigata University, Nishi-ku, Niigata City, Niigata 950-2181, Japan
168 International Centre for Theoretical Sciences, Tata Institute of Fundamental Research, Bengaluru 560089, India
169 State Key Laboratory of Magnetic Resonance and Atomic and Molecular Physics, Innovation Academy for Precision Measurement Science and Technology (APM), Chinese Academy of Sciences, Xiao Hong Shan, Wuhan 430071, China
170 University of Szeged, Dóm tér 9, Szeged 6720, Hungary
171 Universiteit Gent, B-9000 Gent, Belgium
172 University of British Columbia, Vancouver, BC V6T 1Z1, Canada
173 Tata Institute of Fundamental Research, Mumbai 400005, India
174 INAF, Osservatorio Astronomico di Capodimonte, I-80131 Napoli, Italy
175 Università di Trento, Dipartimento di Fisica, I-38123 Povo, Trento, Italy
176 INFN, Trento Institute for Fundamental Physics and Applications, I-38123 Povo, Trento, Italy
177 The University of Mississippi, University, Mississippi 38677, USA
178 University of Michigan, Ann Arbor, Michigan 48109, USA
179 Department of Physics, School of Natural Science, Ulsan National Institute of Science and Technology (UNIST), Ulju-gun, Ulsan 44919, Korea
180 Applied Research Laboratory, High Energy Accelerator Research Organization (KEK), Tsukuba City, Ibaraki 305-0801, Japan
181 Dipartimento di Fisica, Università di Trieste, I-34127 Trieste, Italy
182 Shanghai Astronomical Observatory, Chinese Academy of Sciences, Shanghai 200030, China
183 American University, Washington, D.C. 20016, USA
184 Faculty of Science, University of Toyama, Toyama City, Toyama 930-8555, Japan
185 Institute for Cosmic Ray Research (ICRR), KAGRA Observatory, The University of Tokyo, Kamioka-cho, Hida City, Gifu 506-1205, Japan
186 Carleton College, Northfield, Minnesota 55057, USA
187 University of California, Berkeley, California 94720, USA
188 College of Industrial Technology, Nihon University, Narashino City, Chiba 275-8575, Japan
189 Graduate School of Science and Technology, Niigata University, Nishi-ku, Niigata City, Niigata 950-2181, Japan
190 Department of Physics, National Taiwan Normal University, sec. IV, Taipei 116, Taiwan
191 Astronomy & Space Science, Chungnam National University, Yuseong-gu, Daejeon 34134, Korea, Korea
192 Department of Physics and Mathematics, Aoyama Gakuin University, Sagamihara City, Kanagawa 252-5258, Japan
193 Kavli Institute for Astronomy and Astrophysics, Peking University, Haidian District, Beijing 100871, China
194 Yukawa Institute for Theoretical Physics (YITP), Kyoto University, Sakyou-ku, Kyoto City, Kyoto 606-8502, Japan
195 Graduate School of Science and Engineering, University of Toyama, Toyama City, Toyama 930-8555, Japan

022004-24
196 Department of Physics, Graduate School of Science, Osaka City University, Sumiyoshi-ku, Osaka City, Osaka 558-8585, Japan
197 Nambu Yoichiro Institute of Theoretical and Experimental Physics (NITEP), Osaka City University, Sumiyoshi-ku, Osaka City, Osaka 558-8585, Japan
198 Institute of Space and Astronautical Science (JAXA), Chuo-ku, Sagamihara City, Kanagawa 252-0222, Japan
199 Directorate of Construction, Services & Estate Management, Mumbai 400094 India
200 Universiteit Antwerpen, Prinsstraat 13, 2000 Antwerpen, Belgium
201 University of Białystok, 15-424 Białystok, Poland
202 Department of Physics, Ewha Womans University, Seodaemun-gu, Seoul 03760, Korea
203 National Astronomical Observatories, Chinese Academy of Sciences, Chaoyang District, Beijing, China
204 School of Astronomy and Space Science, University of Chinese Academy of Sciences, Chaoyang District, Beijing, China
205 University of Southampton, Southampton SO17 1BJ, United Kingdom
206 Institute for Cosmic Ray Research (ICRR), The University of Tokyo, Kashiwa City, Chiba 277-8582, Japan
207 Institut de Física d’Altes Energies (IFAE), Barcelona Institute of Science and Technology, and ICREA, E-08193 Barcelona, Spain
208 Graduate School of Science and Technology, Tokyo Institute of Technology, Meguro-ku, Tokyo 152-8551, Japan
209 University of Washington Bothell, Bothell, Washington 98011, USA
210 Institute of Applied Physics, Nizhny Novgorod, 603950, Russia
211 Ewha Womans University, Seoul 03760, South Korea
212 Inje University Gimhae, South Gyeongsang 50834, South Korea
213 Department of Physics, Myongji University, Yongin 17058, Korea
214 Korea Astronomy and Space Science Institute (KASI), Yuseong-gu, Daejeon 34055, Korea
215 Department of Physical Science, Hiroshima University, Higashihiroshima City, Hiroshima 903-0213, Japan
216 Bard College, 30 Campus Rd, Annandale-On-Hudson, New York 12504, USA
217 Institute for Cosmic Ray Research (ICRR), Research Center for Cosmic Neutrinos (RCCN), The University of Tokyo, Kamioka-cho, Hida City, Gifu 506-1205, Japan
218 Institute of Mathematics, Polish Academy of Sciences, 00656 Warsaw, Poland
219 National Center for Nuclear Research, 05-400 Swierk-Otwock, Poland
220 Cornell University, Ithaca, New York 14850, USA
221 Institute for Advanced Research, Nagoya University, Furocho, Chikusa-ku, Nagoya City, Aichi 464-8602, Japan
222 Université de Montréal/Polytechnique, Montreal, Quebec H3T 1J4, Canada
223 Laboratoire Lagrange, Université Côte d’Azur, Observatoire Côte d’Azur, CNRS, F-06034 Nice, France
224 Department of Physics, University of Texas, Austin, Texas 78712, USA
225 Department of Physics, Hanyang University, Seoul 04763, Korea
226 NAVIER, École des Ponts, Univ Gustave Eiffel, CNRS, Marne-la-Vallée, France
227 National Center for High-performance computing, National Applied Research Laboratories, Hsinchu Science Park, Hsinchu City 30076, Taiwan
228 Institute for High-Energy Physics, University of Amsterdam, Science Park 904, 1098 XH Amsterdam, Netherlands
229 NASA Marshall Space Flight Center, Huntsville, Alabama 35811, USA
230 University of Washington, Seattle, Washington 98195, USA
231 Dipartimento di Matematica e Fisica, Università degli Studi Roma Tre, I-00146 Roma, Italy
232 INFN, Sezione di Roma Tre, I-00146 Roma, Italy
233 ESPCI, CNRS, F-75005 Paris, France
234 Concordia University Wisconsin, Mequon, Wisconsin 53097, USA
235 Università di Camerino, Dipartimento di Fisica, I-62032 Camerino, Italy
236 Southern University and A&M College, Baton Rouge, Louisiana 70813, USA
237 Centre Scientifique de Monaco, 8 quai Antoine Ier, MC-98000, Monaco
238 Institute for Photon Science and Technology, The University of Tokyo, Bunkyo-ku, Tokyo 113-8656, Japan
239 Indian Institute of Technology Madras, Chennai 600036, India
240 Saha Institute of Nuclear Physics, Bidhannagar, West Bengal 700064, India

UPPER LIMITS ON THE ISOTROPIC GRAVITATIONAL-WAVE ...

PHYS. REV. D 104, 022004 (2021)
Deceased, August 2020.