



Limits on Hierarchical Black Hole Mergers from the Most Negative χ_{eff} Systems

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Abstract

It has been proposed that some black holes (BHs) in binary black hole (BBH) systems are born from “hierarchical mergers” (HMs), i.e., earlier mergers of smaller BHs. These HM products have spin magnitudes $\chi \sim 0.7$, and, if they are dynamically assembled into BBH systems, their spin orientations will sometimes be antialigned with the binary orbital angular momentum. In fact, as Baibhav et al. showed, $\sim 16\%$ of BBH systems that include HM products will have an effective inspiral spin parameter, $\chi_{\text{eff}} < -0.3$. Nevertheless, the LIGO–Virgo–KAGRA (LVK) gravitational-wave (GW) detectors have yet to observe a BBH system with $\chi_{\text{eff}} \lesssim -0.2$, leading to upper limits on the fraction of HM products in the population. We fit the astrophysical mass and spin distribution of BBH systems and measure the fraction of BBH systems with $\chi_{\text{eff}} < -0.3$, which implies an upper limit on the HM fraction. We find that fewer than 26% of systems in the underlying BBH population include HM products (90% credibility). Even among BBH systems with primary masses $m_1 = 60 M_{\odot}$, the HM fraction is less than 69%, which may constrain the location of the pair-instability mass gap. With 300 GW events (to be expected in the LVK’s next observing run), if we fail to observe a BBH with $\chi_{\text{eff}} < -0.3$, we can conclude that the HM fraction is smaller than $2.5^{+9.1}_{-2.2}\%$.

Unified Astronomy Thesaurus concepts: Gravitational wave astronomy (675); Astrophysical black holes (98); Stellar dynamics (1596); Massive stars (732); Gravitational waves (678); Gravitational wave sources (677); Astrostatistics (1882)

1. Introduction

The network of gravitational-wave (GW) detectors Advanced LIGO (Aasi et al. 2015), Advanced Virgo (Acernese et al. 2015), and KAGRA (Akutsu et al. 2021) has revealed a new population of binary black holes (BBHs), with 69 confident BBH events (false-alarm rate $< 1 \text{ yr}^{-1}$) in the latest LIGO–Virgo–KAGRA (LVK) catalog (Abbott et al. 2021a). Although the BBH discovery rate is accelerating, the formation history of these BBH systems remains unknown. Broadly, a BBH system may either evolve from a pair of binary stars (“isolated binary evolution”; Bethe & Brown 1998; Belczynski et al. 2002; Kalogera et al. 2007; Dominik et al. 2012) or be assembled dynamically in a dense stellar environment containing many black holes (BHs; “dynamical assembly”; Kulkarni et al. 1993; Sigurdsson & Hernquist 1993; Portegies Zwart & McMillan 2000; see Mapelli (2021) and Mandel & Farmer (2022) for recent reviews.

A popular way of distinguishing BBH formation channels is based on the statistics of their spin orientations (Rodríguez et al. 2016; Stevenson et al. 2017; Talbot & Thrane 2017; Vitale et al. 2017). To first order, dynamical assembly leads to random spin orientations (isotropic tilts; Portegies Zwart & McMillan 2000), while isolated binary evolution leads to preferential alignment between the BH spins and the orbital angular momentum axis (Kalogera 2000). The spin magnitudes and orientations of the component BHs in a BBH affect the GW signal primarily through a combination known as the effective inspiral spin parameter, which is approximately conserved during the GW inspiral

(Racine 2008; Santamaría et al. 2010; Ajith et al. 2011),

$$\chi_{\text{eff}} = \frac{m_1 \chi_{1,z} + m_2 \chi_{2,z}}{m_1 + m_2}. \quad (1)$$

Here $\chi_{1,z}$ ($\chi_{2,z}$) is the primary (secondary) dimensionless spin vector projected along the orbital angular momentum axis. The distribution of χ_{eff} in the BBH population provides a simple test of BBH formation channels. If spin tilts are isotropically distributed in BBH systems, the χ_{eff} distribution will be symmetric about zero, pointing to a dynamical origin, whereas if spins are always aligned within 90° , the χ_{eff} distribution will only have positive support, pointing to an isolated origin (Farr et al. 2017).

A population analysis of the latest GW observations suggests that the BBH χ_{eff} distribution is not symmetric about zero, with the majority of systems having $\chi_{\text{eff}} > 0$ (Abbott et al. 2021b). The observation of individual events with strictly positive $\chi_{\text{eff}} > 0.2$ requires the χ_{eff} distribution to extend to nonzero, positive χ_{eff} . In particular, the asymmetry in the χ_{eff} distribution implies that the spin tilt distribution prefers tilts smaller than 90° , indicating that dynamical assembly cannot account for all BBH systems. This asymmetry also suggests that some BHs in the isolated channel have nonzero spins, although the mechanism by which they acquire such spins remains poorly understood, depending on theoretically uncertain factors including angular momentum transport in massive stars (Tayler 1973; Spruit 2002; Fuller et al. 2019; Qin et al. 2019), tidal spin-up of stripped stars in binary systems (Kushnir et al. 2016; Qin et al. 2018; Bavera et al. 2020; Fuller & Lu 2022), BH spin-up during a supernova (Batta et al. 2017; Schröder et al. 2018; Janka et al. 2022), and BH spin-up by accretion from its binary companion (Podsiadlowski et al. 2002; Moreno Méndez et al. 2008; Fragos & McClintock 2015; van Son et al. 2020; Zevin & Bavera 2022).

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Meanwhile, it remains unclear whether the χ_{eff} distribution has support at negative χ_{eff} , which would suggest some contribution from dynamical assembly. The latest LVK catalog GWTC-3 does not contain any high-significance events with confidently negative χ_{eff} (at $>90\%$ credibility under the default parameter estimation priors). However, a few GWTC-3 events, GW191109_010717 and GW200225_060421, have $\chi_{\text{eff}} < 0$ with 90% and 85% credibility, respectively (Abbott et al. 2021a), and lower-significance candidate events with negative χ_{eff} have been reported in independent analyses (Venumadhav et al. 2020; Olsen et al. 2022). Combining information across multiple events, Abbott et al. (2021b, 2021c) and Callister et al. (2022) found that the BBH population likely contains systems with (small) negative χ_{eff} ($\chi_{\text{eff}} \lesssim -0.02$), but this evidence weakens under population models that include a correlation between χ_{eff} and mass ratio (Callister et al. 2021) or a nonspinning subpopulation that would create a narrow peak at zero in the χ_{eff} distribution (Galaudage et al. 2021; Roulet et al. 2021). Disentangling small negative χ_{eff} from nonspinning systems requires resolving the χ_{eff} distribution to very small scales, $\mathcal{O}(10^{-2})$ (Callister et al. 2022). If dynamically assembled BBHs always consist of slowly spinning, or even nonspinning, BHs (Fuller & Ma 2019), the ambiguity around the presence of systems with negative χ_{eff} will persist. However, there is probably a subpopulation of spinning BHs in dense stellar environments: BHs born from previous mergers.

In dynamical formation, as long as the escape speed of the environment is larger than the typical GW recoil kicks, some BBH merger products will be retained and dynamically assembled into new BBH systems in so-called hierarchical mergers (HMs; Antonini & Rasio 2016; Mapelli 2016; McKernan et al. 2018; Rodriguez et al. 2018; Fragione et al. 2022); see Gerosa & Fishbach (2021) for a review. For example, HMs are typically predicted to account for $\sim 10\%$ of mergers from globular clusters (Rodriguez et al. 2019), and this fraction increases for higher-density environments like nuclear star clusters (Mapelli et al. 2021).

Furthermore, HMs are a promising explanation for some of the most massive BBHs observed by the LVK (Abbott et al. 2020; Kimball et al. 2021), especially those with masses above $\sim 40\text{--}65 M_{\odot}$, where we expect few, if any, BHs born directly from stellar collapse due to the theorized pair-instability supernova mass gap (Fowler & Hoyle 1964; Bond et al. 1984; Heger & Woosley 2002; Belczynski et al. 2016). The GW observations find an excess of BH component masses between ~ 35 and $40 M_{\odot}$, followed by a steeper decrease in the merger rate, which may be a signature of pair instability (Fishbach & Holz 2017; Talbot & Thrane 2018; Abbott et al. 2021b, 2021c). The BBHs with primary masses above this excess may then be HM products. The merger rate of BBHs with primary masses between 50 and $100 M_{\odot}$ is $0.099\text{--}0.4 \text{ Gpc}^{-3} \text{ yr}^{-1}$ ($\lesssim 1\%$ of the total merger rate; Abbott et al. 2021b).

In the HM scenario, the binary’s orbital angular momentum contributes to the spin of the merger product, thereby producing BBHs with spinning components and, under the assumption of a classical star cluster,² isotropic spin tilts

(Fishbach et al. 2017; Gerosa & Berti 2017; Doctor et al. 2020). Therefore, BBH systems that contain an HM product have a broad distribution of χ_{eff} that is symmetric about zero (Baibhav et al. 2020; Kimball et al. 2020). In particular, some fraction of these BBH systems must exhibit significantly negative χ_{eff} .

In this Letter, we find that there is a threshold of $\chi_{\text{eff}} < -0.3$, below which we always expect to find the same fraction ($\sim 16\%$) of HMs, regardless of the initial BH spins, binary mass ratio, or number of previous mergers. We then use LVK observations to place upper limits on the fraction of BBH systems with $\chi_{\text{eff}} < -0.3$ in the underlying astrophysical population. Comparing this inferred upper limit against the expected fraction of HMs with $\chi_{\text{eff}} < -0.3$ constrains the HM contribution to the BBH population. As Baibhav et al. (2020) showed, if repeated mergers are the only way for BBHs to gain large spins, this upper limit (and a corresponding limit on observations with large positive χ_{eff}) becomes a direct measurement of the fraction of HMs in the population. However, as discussed earlier, both theory and observation suggest that BBHs can acquire spins in alternative scenarios, especially through binary interactions in the isolated formation channel, which may explain BBH observations with $\chi_{\text{eff}} > 0.2$. Among these spin-up scenarios, HMs are unique in that they also invariably produce large negative χ_{eff} . Thus, the absence of events with $\chi_{\text{eff}} < -0.3$ constrains the HM contribution more tightly than using the positive χ_{eff} fraction yet robustly enough to remain applicable across a wide range of HM scenarios.

The remainder of this paper is organized as follows. Section 2 derives theoretical χ_{eff} distributions for BBH systems that contain one or two HM products, showing that we generically expect 16% of these systems to have $\chi_{\text{eff}} < -0.3$. In Section 3, we infer the astrophysical χ_{eff} distribution from LVK observations and place constraints on the HM fraction, showing that even among BBHs with primary masses of $60 M_{\odot}$ (possibly within the pair-instability mass gap), HMs make up fewer than 69% of systems. We discuss expectations for future observations in Section 4 and conclude in Section 5.

2. Effective Inspiral Spins from HMs

A merger of two first-generation ($1g+1g$) BHs of comparable masses (mass ratio $0.5 < q_{1g} < 1$) produces a second-generation ($2g$) BH with dimensionless spin magnitude $\chi \sim 0.7$ (Berti & Volonteri 2008; Buonanno et al. 2008; Tichy & Marronetti 2008; Fishbach et al. 2017). If this $2g$ BH dynamically forms a binary and merges with another similarly formed $2g$ BH, the effective inspiral spin of this $2g+2g$ BBH merger (with mass ratio $q_{2g} = q_{1g}$) will be drawn from the probability distribution function (PDF) shown in the blue open histogram in the top panel of Figure 1, with the corresponding cumulative distribution function (CDF) shown in the bottom panel. Meanwhile, if the $2g$ BH instead merges with one of the original $1g$ BHs, the effective inspiral spin of the resulting $1g+2g$ BBH (with mass ratio $1/(1+q_{1g})$) will be drawn from the PDF shown by the orange filled histogram in Figure 1. The χ_{eff} distributions of $2g+2g$ and $1g+2g$ BBHs were previously derived in Baibhav et al. (2020). Here we build the χ_{eff} distribution with Monte Carlo samples. We first draw $1g+1g$ BBHs according to three different mass ratio q_{1g} and spin χ_{1g} distributions, as explained in the caption of Figure 1. These different $1g+1g$ distributions have mass ratios in the range

² We focus on classical star clusters in this work, but gas-rich stellar environments, such as the disks of AGNs, may produce preferentially aligned or misaligned spin tilts due to gas torques, as we discuss in the following sections (Bogdanović et al. 2007; McKernan et al. 2018).

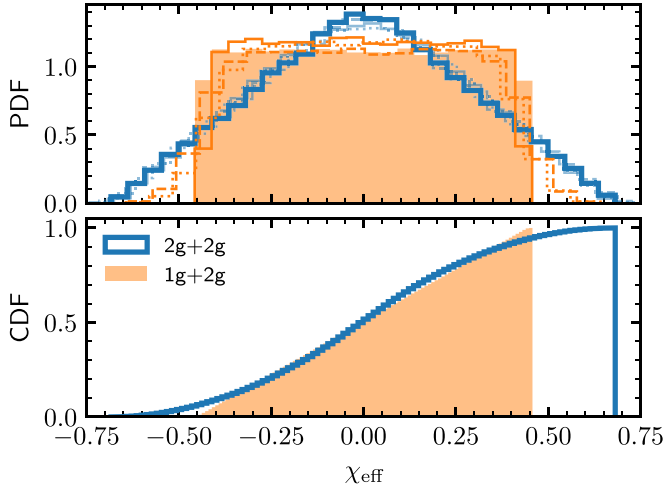


Figure 1. Theoretical distribution of χ_{eff} among BBHs in which one (orange) or both (blue) components are the products of previous mergers. Top panel: χ_{eff} PDF for four different assumptions about the initial spins χ_{1g} and mass ratios q_{1g} of 1g+1g BHs: $\chi_{1g} = 0$ and a uniform q_{1g} distribution $0.8 < q_{1g} < 1$ (default; thick blue and filled orange), $\chi_{1g} = 0$ and $0.5 < q_{1g} < 1$ (thin solid lines), uniform $0 < \chi_{1g} < 0.5$ and $0.8 < q_{1g} < 1$ (thin dashed lines), and $0 < \chi_{1g} < 0.5$ and $0.5 < q_{1g} < 1$ (thin dotted lines). The mass ratios of mixed-generation 1g+2g mergers are always assumed to be given by $1/(1 + q_{1g})$, where q_{1g} is drawn from the assumed mass ratio distribution for 1g+1g mergers. Bottom panel: corresponding CDFs, shown just for the default spin and mass ratio assumptions.

$0.5 < q_{1g} < 1$ and spins in the range $0 < \chi_{1g} < 0.5$, motivated by both theoretical expectations and BBH observations that suggest that asymmetric mass ratios and high spins are rare among merging systems (Abbott et al. 2021b, 2021c; Mandel & Farmer 2022). We then calculate 2g BH spins with PESUMMARY (Hoy & Raymond 2021), which averages over numerical relativity fits (Healy et al. 2014; Hofmann et al. 2016; Jiménez-Forteza et al. 2017). We assume that BBHs always merge with isotropic spin orientations.

Figure 1 shows that for both 2g+2g and 1g+2g mergers, χ_{eff} follows a broad distribution, with significant support at $|\chi_{\text{eff}}| > 0.3$. The χ_{eff} distribution depends only slightly on the mass ratios q_{1g} and initial spins of 1g+1g BHs. The near-universality of the HM χ_{eff} distribution stems from the fact that merger products always have $\chi_{2g} \sim 0.7$ for a wide range of initial spins and mass ratios (Fishbach et al. 2017). Our assumption of isotropic spin tilts implies that the χ_{eff} distribution is always symmetric about zero. In both 2g+2g and 1g+2g scenarios, approximately 16% of the systems assembled hierarchically have $\chi_{\text{eff}} < -0.3$ (ranging from 15% to 17% for the scenarios shown in Figure 1). This result agrees with C. Kimball et al. (2022, in preparation), who updated the results of Kimball et al. (2021) with GWTC-3 and fit the mass and spin distribution for 1g and HM BBH systems, together with the branching fraction between the two subpopulations. They found that 15.4% of BBHs among their HM subpopulation have $\chi_{\text{eff}} < -0.3$.

Higher generations of BHs will follow similar distributions because the final spin stays close to ~ 0.7 , even if the individual merging BHs have large spins. The exception is BH growth through repeated minor mergers ($q \ll 1$), which tend to spin the BH down (Hughes & Blandford 2003). As mentioned in footnote 2, our analysis also does not strictly apply to gas-rich environments for dynamical assembly, such as active galactic nucleus (AGN) disks, where the distribution of spin tilts may

not be isotropic, although misaligned and antialigned systems are expected to be common (Bogdanović et al. 2007; McKernan et al. 2018; Tagawa et al. 2020; Vajpeyi et al. 2022). In this case, if BHs grow through many generations of preferentially aligned (antialigned) repeated mergers, the spin magnitude of the HM products will converge to $\chi \sim 0.9$ ($\chi \sim 0.5$) rather than $\chi \sim 0.7$ (Fishbach et al. 2017). Additionally, the χ_{eff} distribution may not be symmetric. Nevertheless, we expect HMs in AGN disks to produce BBH systems with large negative χ_{eff} (Tagawa et al. 2020), and so our method broadly applies to this scenario, although the quantitative details depend on the highly uncertain spin tilt distribution of BBH systems in AGN disks.

We can convert the inferred fraction of systems with $\chi_{\text{eff}} < -0.3$, $f_{\chi_{\text{eff}} < -0.3}$, into an upper limit on the fraction of HM systems f_{HM} (including both 2g+2g and 1g+2g systems) in the astrophysical BBH population according to the simple rule

$$f_{\text{HM}} \leq \frac{1}{0.16} f_{\chi_{\text{eff}} < -0.3} = 6.25 f_{\chi_{\text{eff}} < -0.3}. \quad (2)$$

Meanwhile, the fraction of systems with $\chi_{\text{eff}} < 0$ can be used to infer an upper limit on the fraction of dynamically assembled BBHs f_{dyn} ,

$$f_{\text{dyn}} \leq \frac{1}{0.5} f_{\chi_{\text{eff}} < 0} = 2 f_{\chi_{\text{eff}} < 0}. \quad (3)$$

3. Observational Limits on the Negative χ_{eff} Fraction

The χ_{eff} CDF for the astrophysical BBH population, as inferred from the 69 confident BBH events in GWTC-3, is shown in the right panel of Figure 2. In blue, we show the χ_{eff} distribution inferred under a truncated Gaussian model (Roulet & Zaldarriaga 2019; Miller et al. 2020), with flat priors on the mean (between -0.5 and 0.5) and standard deviation (between 0.03 and 0.5). We assume that the χ_{eff} distribution is independent of other BBH properties, like mass or redshift, and we simultaneously fit the primary mass, mass ratio, and redshift distributions. We adopt a broken power-law model for the primary mass distribution, a power-law model for the mass ratio conditioned on the primary mass, and a power law in $(1+z)$ for the redshift distribution (Abbott et al. 2021c). While we expect the mass ratio distribution to have some effect on the χ_{eff} distribution because of the correlations between χ_{eff} and mass ratio in the parameter estimation of individual events, we do not expect the details of the primary mass or redshift distribution to noticeably affect our results. To perform the fit, we use the parameter estimation samples released in Abbott et al. (2019, 2021a, 2021d, 2021e), using the available “Overall_posterior” (for the first two observing runs), “NRSur7dq4,” “PrecessingSpinIMRHM,” or “C01:Mixed” (for the third observing run) set of samples. We use the detector sensitivity estimates from Abbott et al. (2021a) and LIGO Scientific Collaboration et al. (2021).

The inferred distribution is consistent with the results of Abbott et al. (2021b), who fit the truncated Gaussian model to the same set of events. The median χ_{eff} (a CDF value of 0.5, denoted by the dashed horizontal line) is greater than zero (denoted by the dashed vertical line) at $>99\%$ credibility, implying that the χ_{eff} distribution is not symmetric about zero, and not all binaries are dynamically assembled. In fact, following Equation (3), the inferred $\chi_{\text{eff}} < 0$ fraction implies

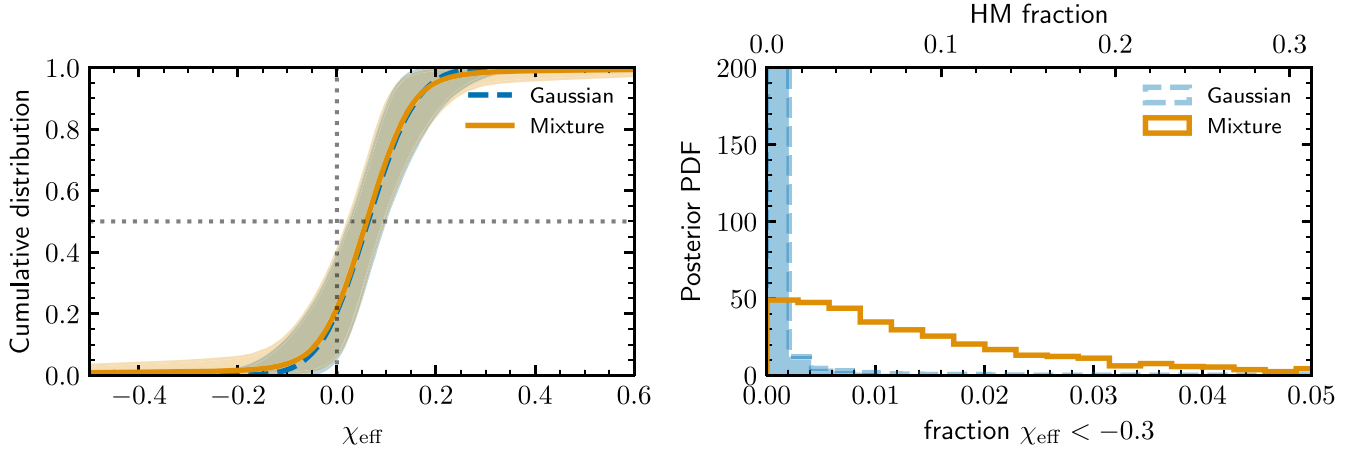


Figure 2. Inferred χ_{eff} CDF for the astrophysical BBH population (left) and posterior PDF on the fraction of $\chi_{\text{eff}} < -0.3$ (right). We infer the χ_{eff} distribution by fitting the confident BBH events in GWTC-3 to two models, a Gaussian (blue) and a mixture model between a Gaussian and a uniform distribution (orange), both truncated to the physically allowed range $-1 < \chi_{\text{eff}} < 1$. Solid lines show the median CDF value at each χ_{eff} , and shaded regions show 90% credible intervals. Both models assume that the χ_{eff} distribution does not vary with BBH mass. The mixture model is designed to allow for more support at the tails of the χ_{eff} distribution and so allows for a higher fraction of systems with $\chi_{\text{eff}} < -0.3$. The constraints on the $\chi_{\text{eff}} < -0.3$ fraction can be interpreted as limits on the contribution of HMs to the astrophysical BBH population (right panel, top axis).

a dynamically assembled fraction of at most $0.42^{+0.37}_{-0.31}$ (90% credibility).

Under the Gaussian model, we infer $f_{\chi_{\text{eff}} < -0.3} < 0.08\%$ (see right panel of Figure 2), which, according to Equation (2), suggests that HMs comprise fewer than 0.5% of the BBH population (90% credibility). Given that not all BBH systems are dynamically assembled, this upper limit corresponds to $f_{\text{HM}}/f_{\text{dyn}} = 0.7\%$ of the dynamically assembled population. However, this tight upper limit falsely assumes that we can resolve the tails of the χ_{eff} distribution to subpercent accuracy. In reality, because we only fit 69 events, we can only probe the distribution to a resolution of $1/69 \sim 1.4\%$ and conclude that fewer than 1.4% of BBH events in the detected population have $\chi_{\text{eff}} \lesssim -0.2$. (According to the inferred χ_{eff} distribution, the minimum observed χ_{eff} out of 69 events is $\chi_{\text{eff}} \gtrsim -0.2$.) Among detected HM events, assuming the HM fraction is the same at all BBH masses (an assumption that we revisit later), 11% have $\chi_{\text{eff}} < -0.3$. Thus, we expect that a more conservative upper limit on the HM fraction would be $0.014/0.11 \sim \mathcal{O}(10\%)$, which is much higher than the 0.5% inferred under the Gaussian model.

To derive a conservative upper limit on the HM fraction, we fit the χ_{eff} distribution to a model inspired by Roulet et al. (2021): a mixture model between a Gaussian and uniform component truncated to the range $-1 < \chi_{\text{eff}} < 1$,

$$p(\chi_{\text{eff}} | \mu, \sigma, f_U) = \frac{f_U}{2} + (1 - f_U) \mathcal{N}_{\mu, \sigma}^{T_{-1,1}}(\chi_{\text{eff}}). \quad (4)$$

We refer to this model as a Gaussian + Uniform Mixture. We take the same priors on μ , σ as the Gaussian model and a flat prior on the mixing fraction f_U in the range $0 < f_U < 1$. This model is introduced by Roulet et al. (2021) to allow for the presence of population outliers (which may or may not have already been observed) with respect to the “bulk” population. In our case, the bulk population, which contains most of the observations, is the Gaussian component. The broad uniform component can accommodate any outlier BBH systems, in particular, HM systems. The Gaussian + Uniform Mixture model therefore favors distributions with support at extreme

values of χ_{eff} far from the mean of the Gaussian. In other words, the induced prior on $f_{\chi_{\text{eff}} < -0.3}$ from the mixture model favors large values.

The χ_{eff} CDF inferred under the Gaussian + Uniform Mixture model is shown in orange in Figure 2 (left panel). The bulk 10%–90% of the inferred distribution agrees with the results under a Gaussian model, but as expected, the Gaussian + Uniform Mixture model puts more support at the tails of the distribution. As shown in the right panel of Figure 2, under the mixture model, we measure that up to 3.7% of BBH systems have $\chi_{\text{eff}} < -0.3$ (90% credibility). This implies an upper limit $f_{\text{HM}} \lesssim 23\%$ out of the total BBH population, or 63% of the dynamically assembled population, which, by design, is much more conservative than the result from the Gaussian model.

The results in Figure 2 assume that the χ_{eff} distribution does not depend on BBH mass. However, we expect the HM contribution to be larger among more massive BBH systems, causing a signature trend between χ_{eff} and BBH mass (Baibhav et al. 2020; Safarzadeh et al. 2020; Tagawa et al. 2021; Franciolini & Pani 2022). With GWTC-3, there is no conclusive evidence that the χ_{eff} distribution varies with mass (Abbott et al. 2021b; Biscoveanu et al. 2022), although it is not ruled out. To explore this possibility, we fit mass-dependent χ_{eff} distributions and measure $f_{\chi_{\text{eff}} < -0.3}(m_1)$ and the corresponding upper limit on $f_{\text{HM}}(m_1)$ as functions of the primary BH mass m_1 . We extend both the Gaussian and Gaussian + Uniform Mixture models introduced previously to allow for a possible dependence on m_1 . In the Gaussian model, we take the mean and standard deviations to be functions of m_1 . In the Gaussian + Uniform Mixture model, we take the mixture fraction f_U to be a function of m_1 . We assume the same models for the marginal primary mass, mass ratio, and redshift distributions as before.

In detail, in the m_1 -dependent Gaussian model, we assume that $\mu(m_1)$ and $\log \sigma(m_1)$ are linear. We define $\mu_{85} \equiv \mu(m_1 = 85 M_{\odot})$ and $\mu_5 \equiv \mu(m_1 = 5 M_{\odot})$ and take

$$\mu(m_1) = \frac{\mu_{85} - \mu_5}{80}(m_1 - 5) + \mu_5. \quad (5)$$

Similarly, defining $\log \sigma_{85} \equiv \log \sigma(m_1 = 85 M_\odot)$ and $\log \sigma_5 \equiv \log \sigma(m_1 = 5 M_\odot)$, we take

$$\log \sigma(m_1) = \frac{\log \sigma_{85} - \log \sigma_5}{80}(m_1 - 5) + \log \sigma_5. \quad (6)$$

In the m_1 -dependent Gaussian + Uniform Mixture model, we define $f_5 \equiv f_U(m_1 = 5 M_\odot)$ and $f_{85} \equiv f_U(m_1 = 85 M_\odot)$ and take

$$f_U(m_1) = (1 + ae^{bm_1})^{-1}, \quad (7)$$

where

$$b = 1/80[\log(1/f_{85} - 1) - \log(1/f_5 - 1)] \quad (8)$$

and

$$a = (1/f_5 - 1)e^{-5b}. \quad (9)$$

We take flat priors on all parameters.

The resulting constraints on $f_{\chi_{\text{eff}} < 0}(m_1)$, which places an upper limit on the dynamical fraction as a function of m_1 , and $f_{\chi_{\text{eff}} < -0.3}(m_1)$, which limits the HM fraction as a function of m_1 , are shown in the top and bottom panels of Figure 3. As seen in the top panel, the dynamical fraction must be less than 1, at least among systems with primary masses in the range ~ 10 – $45 M_\odot$, according to both the m_1 -varying Gaussian and m_1 -varying Gaussian + Uniform Mixture models. The m_1 -varying Gaussian permits the dynamical fraction to extend to 1 at lower and higher masses, where there are fewer events and the constraints are weaker.

The bottom panel of Figure 3 shows that $f_{\chi_{\text{eff}} < -0.3}$ is consistent with zero at all m_1 , in agreement with the results of Figure 2. Nevertheless, under the greater flexibility of the m_1 -varying models, it is possible that $f_{\chi_{\text{eff}} < -0.3}$, and therefore the HM contribution, increases with increasing primary mass, as we would expect in an HM scenario. The results are similar for both m_1 -varying models of Figure 3, although the mixture model a priori favors larger HM fractions and provides a more conservative upper limit. Our conclusions are consistent with the mass-dependent χ_{eff} fits of Franciolini & Pani (2022; see their Figure 3). However, in agreement with Biscoveanu et al. (2022), we do not find strong evidence that the χ_{eff} distribution varies with mass. Instead, the data are consistent with a constant (small) HM fraction at all masses but permit a larger HM fraction at higher masses where there are fewer events, leading to weaker upper limits.

We find that among BBHs with primary masses $m_1 = 60 M_\odot$, the HM fraction is less than 64% (according to the m_1 -varying Gaussian model) or 69% (according to the m_1 -varying Gaussian + Uniform Mixture model) at 90% credibility. The fact that this fraction is smaller than unity implies that, in the absence of a more exotic alternative formation channel, stellar collapse must make BHs with masses up to $\gtrsim 60 M_\odot$. This limit on the location of the pair-instability mass gap in turn has implications for stellar structure (Belczynski 2020; Renzo et al. 2020; Woosley & Heger 2021; Costa et al. 2021), nuclear physics (Farmer et al. 2019, 2020), new particles (Croon et al. 2020), and cosmology (Farr et al. 2019).

Although HMs may make up the majority of the high-mass population with $m_1 \gtrsim 60 M_\odot$, systems at these masses make up less than $\sim 1\%$ of the underlying population, as shown in the top axis of Figure 3. The small contribution of these high-mass systems, for which HMs may dominate, to the overall BBH population implies that HMs must make up no more than 26%

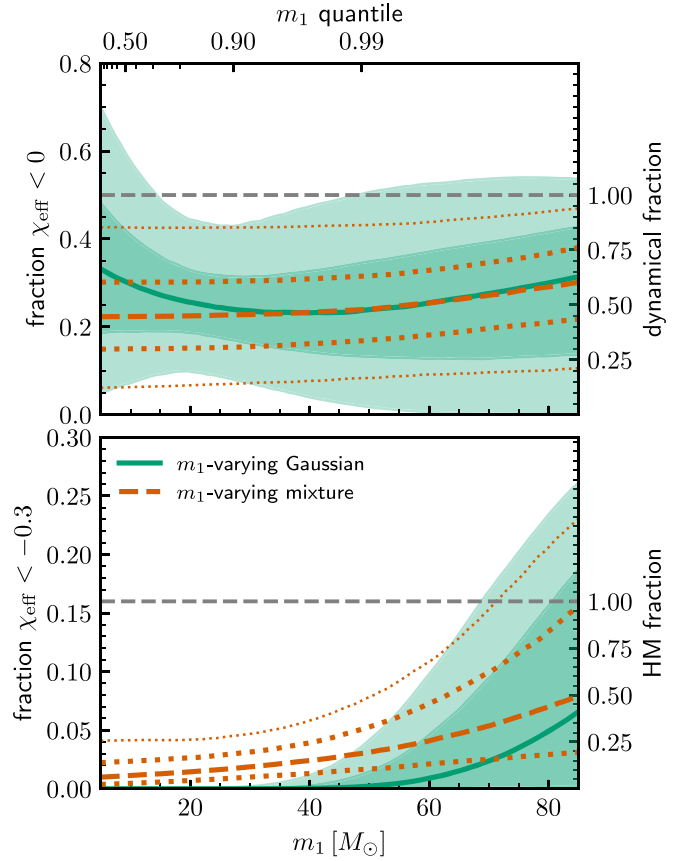


Figure 3. Fraction of systems with $\chi_{\text{eff}} < 0$ (top panel) and $\chi_{\text{eff}} < -0.3$ (bottom panel) at each primary mass m_1 , inferred by fitting a mass-dependent χ_{eff} distribution. The bottom axis shows primary mass in solar masses, while the top axis shows the corresponding quantile in the marginal m_1 posterior population distribution. In each panel, the orange dashed line shows the median value at each m_1 inferred under the m_1 -varying Gaussian + Uniform Mixture model, while the green solid line shows the median for the m_1 -varying Gaussian model. The inner orange dotted lines (dark green shaded region) enclose the symmetric 50% credible interval for the mixture (Gaussian) model. In the top panel, the outer orange dotted lines (light green shaded region) enclose the symmetric 90% credible interval for the two models, while in the bottom panel, we show the 90% upper limit, because the posterior peaks at HM fraction = 0 for all m_1 . Marginalizing over the mass distribution, the total HM fraction across all masses is less than 1.3% (Gaussian model) or 26% (mixture model; 90% credibility).

of the total BBH population (at 90% credibility), similar to the upper limit of 23% inferred under the m_1 -independent model. If the HM fraction is as high as 26%, the majority of HMs are hidden below 50 – $60 M_\odot$, because the high-mass BBH systems can only account for 1% of mergers.

We have presented results on the HM fraction using a Gaussian model and a Gaussian + Uniform Mixture model, but we have also verified that our upper limit of 26% holds with alternative models for the χ_{eff} distribution. For example, we considered a Student t distribution (which has more support at the tails compared to a Gaussian), as well as mixture models between a Gaussian and one of the predicted HM distributions of Figure 1. In this latter case, the inferred branching fraction for the HM component can be directly interpreted as the HM fraction and agrees with our upper limit of $f_{\text{HM}} \lesssim 26\%$ within a couple of percent regardless of the assumed HM distribution. Furthermore, this limit is consistent with but more conservative than the inferred upper limit on the HM fraction from Kimball et al. (2022, in preparation).

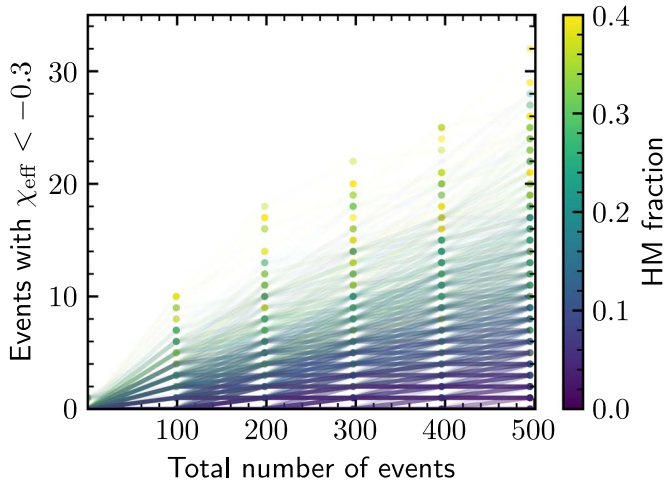


Figure 4. Expected number of BBH events with χ_{eff} as the total number of observations increases. We plot 2000 realizations, where each realization is drawn from the hyperposterior of the m_1 -varying Gaussian + Uniform Mixture model. Each realization is colored by the corresponding HM fraction. Larger HM fractions predict more events with $\chi_{\text{eff}} < -0.3$, especially as the total number of events approaches 500.

4. Predictions for Future Observations

With 69 BBH events, the absence of observations with $\chi_{\text{eff}} < -0.3$ limits the contribution of HMs to be most likely under $\mathcal{O}(10\%)$, but HM fractions as high as $\sim 30\%$ are still permitted by the data. As the BBH catalog grows, the number of future observations with $\chi_{\text{eff}} < -0.3$ will inform a tighter measurement of the HM contribution. Figure 4 shows the expected number of events with $\chi_{\text{eff}} < -0.3$ as a function of the total number of observations for 2000 different realizations, where each realization corresponds to a different hyperposterior draw under the m_1 -varying Gaussian + Uniform Mixture model. A hyperposterior draw specifies a mass–spin–redshift distribution from which we in turn draw 500 mock observations. After every 100 such mock observations, we track how many of them have $\chi_{\text{eff}} < -0.3$, as plotted in Figure 4. Each line is colored by the total HM fraction of its corresponding hyperposterior draw; there are more realizations with small HM fractions (dark purple curves) because we inferred that these values are more likely. When drawing mock observations, we incorporate selection effects matching the LIGO Scientific Collaboration et al. (2021). This is an approximation for the detector sensitivity in future observing runs, which will observe events to much higher distances but have similar mass- and χ_{eff} -dependent selection effects.

As seen in Figure 4, large HM fractions generally predict a higher fraction of events with $\chi_{\text{eff}} < -0.3$. We currently have fewer than 100 total events, so we would not expect more than a couple of events with $\chi_{\text{eff}} < -0.3$ according to any of the realizations (by construction, all of the allowed HM fractions are consistent with the current lack of events with $\chi_{\text{eff}} < -0.3$). However, by the time we observe 300 events, if the HM fraction is higher than 20%, we expect to observe 7^{+7}_{-5} events with a true $\chi_{\text{eff}} < -0.3$; we will observe at least one such event with 99.4% credibility. Meanwhile, if the HM fraction is smaller than 10%, unless there is another mechanism for producing events with large negative χ_{eff} , we will observe fewer than six such events (90% credibility).

A related question is whether we will observe events for which we can confidently identify that χ_{eff} is negative despite typical measurement uncertainties. A population analysis reveals the fraction of events with negative χ_{eff} but not necessarily which individual events have $\chi_{\text{eff}} < 0$, especially if $|\chi_{\text{eff}}|$ is small. However, events for which the true χ_{eff} is more negative than -0.3 will usually have very little likelihood support at positive χ_{eff} , allowing us to identify $\chi_{\text{eff}} < 0$ at high ($>99\%$) credibility (Ng et al. 2018). Figure 4 suggests that if the HM fraction is above 10%, we expect an observation with confidently negative χ_{eff} within the first 150 observations. If we have not seen a system with a confidently negative χ_{eff} within the first 300 events, we may conclude that the HM fraction is smaller than $2.5^{+9.1}_{-2.2}\%$.

5. Conclusion

The formation of BBH systems through dynamical interactions will generally lead to misaligned spins, with some systems having $\chi_{\text{eff}} < 0$. Furthermore, because repeated mergers produce BHs with spin magnitudes $\chi \sim 0.7$, HMs in dense stellar environments will produce BBH systems with $\chi_{\text{eff}} < -0.3$.

Among the current GW observations by the LVK Collaboration, some BBH events may have small negative χ_{eff} , but there are no observations with $\chi_{\text{eff}} \lesssim -0.2$. We argue that the lack of observations with moderately large, negative χ_{eff} values limits the possible contribution of 2g or higher-generation HM systems to the BBH population. This, in turn, limits the escape speeds of dense stellar environments because environments with high escape speeds will inevitably produce too many HMs to match the dearth of observations with $\chi_{\text{eff}} \lesssim -0.2$. Zevin & Holz (2022) made a similar argument using upper limits on the high-mass merger rate. They showed that environments with high escape speeds will overproduce BBH mergers with masses above $\sim 50 M_{\odot}$ if BBH formation is too efficient. Here we use upper limits on the negative χ_{eff} rate to place complementary limits on the HM fraction.

Our main conclusions are as follows.

1. As our most conservative upper limit, fewer than 4.2% of BBH systems in the astrophysical population have $\chi_{\text{eff}} < -0.3$. Assuming a classical star cluster in which BBHs merge with isotropic spin tilts, we expect 16% of 1g+2g or 2g+2g HMs to have $\chi_{\text{eff}} < -0.3$. Therefore, HMs make up no more than 26% of the BBH population (90% credibility).
2. In agreement with theoretical expectations, the data permit the HM fraction to increase with BBH primary mass. However, this is not required by the data. HMs may dominate the BBH population with primary masses above $\sim 60 M_{\odot}$, but such massive BBHs make up less than 1% of the population (see Figure 3). This implies that if the HM fraction is as high as 26%, the vast majority of these HMs have $m_1 < 60 M_{\odot}$.
3. Below primary masses of $\sim 60 M_{\odot}$, HMs make up the minority of systems. This suggests that stellar collapse can create BHs up to $\gtrsim 60 M_{\odot}$, implying a lower limit on the pair-instability mass gap. Alternatively, more exotic possibilities may populate the pair-instability gap, or our assumption of an isotropic spin tilt distribution may be invalid. However, if HMs between preferentially aligned BBHs populate the gap, the dearth of systems with $\chi_{\text{eff}} > 0.4$ limits their contribution.

4. If HMs contribute significantly to the BBH population, we expect to observe a handful of systems with $\chi_{\text{eff}} < -0.3$ within the LVK's next observing run (see Figure 4). We will probably be able to identify these events as confidently ($>99\%$ credibility) having negative χ_{eff} . If the HM fraction is above 10%, we expect to observe a BBH with $\chi_{\text{eff}} < -0.3$ within the first 150 events.

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