Identifying when precession can be measured in gravitational waveforms

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In binary-black-hole systems where the black-hole spins are misaligned with the orbital angular momentum, precession effects leave characteristic modulations in the emitted gravitational waveform. Here, we investigate where in the parameter space we will be able to accurately identify precession, for likely observations over coming LIGO-Virgo-KAGRA observing runs. Despite the large number of parameters that characterize a precessing binary, we perform a large scale systematic study to identify the impact of each source parameter on the measurement of precession. We simulate a fiducial binary at moderate mass-ratio, signal-to-noise ratio (SNR), and spins, such that precession will be clearly identifiable, then successively vary each parameter while holding the remaining parameters fixed. As expected, evidence for precession increases with signal-to-noise-ratio (SNR), higher in-plane spins, more unequal component masses, and higher inclination, but our study provides a quantitative illustration of each of these effects, and informs our intuition on parameter dependencies that have not yet been studied in detail, for example, the effect of varying the relative strength of the two polarizations, the total mass, and the aligned-spin components. We also measure the “precession SNR” \( \rho_p \), to quantify the signal power associated with precession. By comparing \( \rho_p \) with both Bayes factors and the recovered posterior distributions, we find it is a reliable metric for measurability that accurately predicts when the detected signal contains evidence for precession.

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I. INTRODUCTION

In September 2015, the first direct detection of gravitational-wave (GWs) marked the beginning of GW astronomy [1]. Another 14 detections have been announced by the LIGO Scientific and Virgo collaborations (LVC), the vast majority of which were due to black-hole (BH) mergers [2–7]. Additional events have also been reported by independent groups [8–11]. These GW observations have already provided significant insights into gravitational physics, cosmology, astronomy, nuclear physics and fundamental physics (see e.g., Refs. [12–19]). With an order of magnitude more observations expected over the next 5–10 years, as the sensitivities of the LIGO [20,21], Virgo [22] and KAGRA [23] detectors improve and additional detectors come online, GW astronomy from compact-binary mergers has the potential to transform our understanding of gravitational and fundamental physics [24–26].

Everything we learn from GW binary-black-hole (BBH) observations is a consequence of a detailed parameter estimation analysis that extracts the source parameters of the binary. While some parameters are extracted with good precision, inspiral dominated signals show strong correlations between certain parameters which means that they cannot be measured so accurately, for example correlations between the binary’s distance and inclination [27–29], the two masses [27,30], and the mass-ratio and spin components aligned to the binary’s orbital angular momentum [30–33]. As well as studies of the inspiral, work has been done to extract the source properties for high mass signals dominated by the merger ringdown, see e.g., [34–37].

Spin components misaligned with the binary’s orbital angular momentum, leading to a precession of the binary’s orbital plane and hence modulations of the amplitude and phase, have not yet been unambiguously measured in GW observations [2], see Fig. 1. Precession effects and correlations with other parameters are understood in principle [38,39] but since theoretical signal models of precessing binaries that include the merger and ringdown date from only shortly before the first detections [40,41], we have less experience of when precession will be measurable, and what the impact will be on other parameter measurements.

The purpose of this paper is to explore when precession will be measurable, and its impact on other parameter measurements, in the kind of configurations that are representative of expectations from binary populations based on LIGO-Virgo-KAGRA observations to date [2].
By utilizing the precession signal-to-noise ratio (SNR) $\rho_p$ [43,44] as a quantifier for the measurability of precession, we also verify that $\rho_p$ is indeed a good metric for the measurability of precession across the vast majority of the parameter space, and relate it to the standard means to identify the presence of precession, the Bayes factor. In doing so, we show that computationally expensive parameter estimation runs can be avoided by simply calculating the precession SNR.

Previous work has explored the general phenomenology of precession effects: its increased measurability with large in-plane spins [45–47], large mass ratios [45,46], high inclination [36,38,44,48–50], and of course high SNR [45,46,51]. Beyond these general expectations, the quantitative behavior of parameter measurements in the presence of precession has not been studied in great detail for typical LIGO-Virgo-KAGRA observations. The measurability of precession for high mass ratio LIGO-Virgo-KAGRA observations like GW190814 has been investigated in recent work [52].

In this paper, we focus on the region of parameter space most likely to yield binaries with observable precession: binaries of comparable mass, with moderate in-plane and aligned-spin components [44]. We perform a series of one-dimensional investigations of the parameter space, in which we vary one parameter at a time: total mass, mass ratio, spins (both in-plane as characterized by $\chi_p$, and the aligned spin combination $\chi_{\text{eff}}$), the binary orientation (both the inclination of the orbit and also binary polarization), and the sky location and show the impact of varying each of the binary parameters individually. These investigations serve to confirm that much of the known phenomenology is apparent even at relatively low SNR, while also demonstrating that the precession SNR can be effectively used across a significant fraction of the parameter space to predict the observable consequences of precession without the need for computationally costly parameter estimation analyses.

This paper is structured as follows: Sec. II provides an introduction to precession, a brief recap of the two-harmonic approximation that allows us to define $\rho_p$, and a summary of the importance of precession across the parameter space. Section III provides an introduction to the parameter estimation techniques used here, and parameter estimation results and interpretation for our fiducial system. In Sec. IV we perform a series of one-dimensional explorations of the parameter space. In Sec. VI we compare the predicted precession SNR with observations and in Sec. V we compare precession SNR with the Bayes factors between precessing and nonprecessing runs. We conclude with a summary and discussion of future directions.

II. BLACK HOLE SPIN INDUCED PRECESSION

A binary consisting of two compact objects will slowly inspiral due to the emission of GWs. Assuming that the binary is on a quasi-spherical orbit, it may be described by the individual component masses, $m_1$ and $m_2$ (where we define $m_1 > m_2$ and we denote the mass ratio to be $q = m_1/m_2 > 1$), and their spin angular momenta $S_1$ and $S_2$.

For the case where the total spin is misaligned with the total orbital angular momentum, $S_1 \times L \neq 0$, the system undergoes spin-induced precession. In most cases $L \ll J$, and the system undergoes “simple precession,” where the orbital angular momentum precesses around the
(approximately constant) total angular momentum, $\mathbf{J} = \mathbf{S}_1 + \mathbf{S}_2 + \mathbf{L}$, and the spins precess such that $\hat{\mathbf{S}} = -\mathbf{L}$, where $\mathbf{S} = \mathbf{S}_1 + \mathbf{S}_2$ [38,39].

The strength of precession is characterised by the tilt angle of the binary’s orbit, $\beta$, defined as the polar angle between $\mathbf{L}$ and $\mathbf{J}$; see Fig. 2. $\beta$ is determined primarily by the total spin in the plane, and the binary’s mass ratio and separation. At leading order the magnitude of the binary’s orbital angular momentum is given by $L = \mu \sqrt{M_r}$, and we can write,

$$\cos \beta = \frac{\mu \sqrt{M_r} + S_\parallel}{[(\mu \sqrt{M_r} + S_\parallel)^2 + S_\perp^2]^{1/2}},$$

where $M = m_1 + m_2$ is the total mass of the source, $\mu = m_1 m_2 / M$ is the reduced mass, $r$ is the separation and $S_\parallel$ and $S_\perp$ are the total spin parallel and perpendicular to $\mathbf{L}$ respectively. In general, the larger the “opening angle,” $\beta$, the more prominent the precession effects.

In simple precession cases, $\beta$ slowly increases during inspiral. This temporal variation is mostly dependent on the variation of $r$ as, throughout the portion of the binary’s inspiral that is visible in the GW detectors, $\mu$, $M$, $S_\parallel$ and $S_\perp$ are all approximately constant. Therefore, $\beta$ also typically varies very little over the duration that is visible in the GW detectors, and it is possible to make the simplifying assumption that $\beta$ remains constant. This assumption has been used to good effect in previous work, e.g., Ref. [48], although in this work we make no assumptions about $\beta$.

It is often convenient to quantify the binary’s in-plane spin by the scalar quantity $\chi_p$ [53] (see Refs. [54,55] for alternative measures). $\chi_p$ estimates a time-average of the in-plane spin components that drive precession, motivated by the leading-order PN precession equations [38,39], and is defined as,

$$\chi_p = \frac{1}{A_1} \max(A_1 S_{1\parallel}, A_2 S_{2\perp}),$$

where $A_1 = 2 + 3q/2$ and $A_2 = 2 + 3q/2$ and $S_{i\parallel}$ is the component of the spin perpendicular to $\mathbf{L}$ on the $i$th black hole. $\chi_p$ as defined above, takes values between 0 (non-precessing system) and +1 (maximally precessing system).

A. Two harmonic approximation

Non-zero values of $\beta$ and $\chi_p$ indicate that a binary is precessing, and, if all other parameters are kept constant, they parameterise the amount of precession in a system. However, they are not sufficient to tell us whether precession will be observable. Precession appears in the signal as modulations of the amplitude and phase, but these also depend on the binary orientation and signal polarization. References [43,44] introduce a method for decomposing a precessing waveform into a series of five non-precessing harmonics, where the characteristic modulations of a precessing signal are caused by the beating of these harmonics. The harmonics form a power series in the parameter,

$$b = \tan(\beta/2).$$

In most regions of parameter space, the two leading harmonics (a leading-order term independent of $b$, and a first-order term proportional to $b$) are sufficient to capture the significant precession features in the waveform, and the other harmonics can be neglected. As discussed in detail in Sec. V of Ref. [43]), these other precession harmonics can be ignored for binaries where $\beta \lesssim 45^\circ$. In general this is true for all binaries, apart from relatively extreme systems, for example, those that have $\theta_{\text{in}}$ close to edge-on and are either highly precessing or have very large negative spins.

Thus, for almost all signals we expect to observe, the waveform can faithfully be expressed as,

$$h(f) \approx A_0 h^0(f) + A_1 h^1(f),$$

where $A_0$ and $A_1$ are complex, orientation dependent amplitudes, and $h^0(f)$ and $h^1(f)$ are the waveforms of the two leading harmonics. For a detector with a one-sided noise spectral density of $S_n(f)$, the relative amplitude of the harmonics is given by

$$\tilde{b} = \frac{|h^1|}{|h^0|} = \sqrt{\frac{\int_0^f d\omega |h^1(\omega)|^2}{\int_0^f d\omega |h^0(\omega)|^2}}$$

which is the average value of $b$ over the frequency range defined by our starting frequency (20 Hz) to our Nyquist
At higher masses, the two harmonics have significant overlap and we discuss the impact of this when we vary the total mass of the binary below.

The quantity \( \rho_p \) parametrizes the observable precession, it is therefore the measured quantity in the data. By considering what we actually measure in the data we are able to understand many of the correlations and degeneracies in the physical parameters that are caused by the presence of (or lack of) measurable precession.

In the absence of precession, \( \rho_p^2 \) will be \( \chi^2 \) distributed with two degrees of freedom. Consequently, if there is no observable precession in the system, \( \rho_p \geq 2.1 \) in only 10% of cases. Thus far we have used \( \rho_p = 2.1 \) as a simple threshold to determine if there is any observable precession in the system. We revisit this in more detail in Sec. VI B.

### B. Observability of precession

The strength of the modulations in the GW signal depend primarily on the opening angle, \( \beta \), and this is reflected in the expansion parameter \( b \) in the two-harmonic approximation; the precession frequency \( \dot{\alpha} \) also plays a role. The strength of the modulations in the observed signal also depend on the binary’s inclination to the observer, \( \iota \), and the detector polarization \( \psi \), and these are all incorporated into the precession SNR \( \rho_p \), through Eqs. (7) and (8). From these we can draw immediate conclusions about the scenarios in which precession will be most easily measured. These observations are in general not new (see, as always, the pioneering discussions in Refs. [38,39]), but we summarize them here and, where salient, present them in terms of the two harmonic formalism, which highlights the insights and intuition that are simplified in this formulation. We then compare these expectations with the quantitative results that we find in our full parameter estimation study.

Our first basic picture of the strength of precession effects comes from Eq. (1), which gives the dominant effect on \( \beta \) during the inspiral. If we first consider cases where the spin is entirely in the orbital plane, i.e., \( S_{\perp} = 0 \), we see that the opening angle \( \beta \) will be zero if \( S_{\parallel} = 0 \) (as we would expect), and increases linearly for small \( S_{\parallel} \). The opening angle also increases as \( \mu \) decreases, i.e., as the mass ratio is increased. Equation (1) is no longer accurate near merger, and for equal-mass systems \( \beta \) does not become large, but for large mass ratios the opening angle can approach 90°.

If we now consider nonzero \( S_{\parallel} \), we see that the level of precession will be reduced for systems with a positive aligned-spin component, and will be increased for systems with a negative aligned-spin component. The importance of this effect will depend on the other terms, but we can see that for a high-mass-ratio system where \( \mu \) is very small, and close to merger, so \( rM \) is also small, the aligned-spin component will have a strong effect on \( \beta \), and therefore the measurability of precession. A negative \( S_{\parallel} \) is necessary to achieve \( \beta > 90° \), and for large mass-ratio systems near
merger (small $\mu$ and $rM$) and large negative $S_\|$, $\beta$ can approach 180°, but such systems will be rare.

The measurability of precession also depends on the orientation of the binary with respect to the detector, $\theta_{\text{IN}}$. As we see in Eq. (7), precession effects will be minimal if $\theta_{\text{IN}} \sim 0°$ or 180°, i.e., the observer views the system from the direction of $\mathbf{J}$. We expect precession to be strongest in the observed waveform for orientations close to $\theta_{\text{IN}} \sim 90°$. Additionally, when the detector, or network is primarily sensitive to the $\times$ polarization, precession effects will be more significant. The amplitude of the $k = 1$ harmonic vanishes in the + polarization for both face on $\theta_{\text{IN}} = 0°$ and 180° and edge-on $\theta_{\text{IN}} = 90°$ systems, while the $\times$ polarization is maximal for edge-on systems. Additionally, the $\times$ polarization for the $k = 0$ harmonic vanishes for edge on systems, while the $+$ polarization is only reduced by a factor of two. Thus, even when $b$ is small, there can be observable precession when the system is close to edge on and the network is preferentially sensitive to the $\times$ polarization. For a given choice of masses and spins, the maximum precession SNR is $\rho_p = \rho/\sqrt{2}$.

III. PARAMETER ESTIMATION RESULTS

A. Standard configuration

We begin by describing the results of the parameter recovery routine for a specific simulated signal. The details of the signal are given in Table I. These parameters were chosen so that precession effects would be significant in the observed waveform while still being consistent with the observed population of BBHs. In the following sections, we vary over the parameters of the signal one-by-one to investigate the impact of each parameter on the observability of precession and the accuracy of parameter recovery. For each parameter, we are able to both increase and decrease the significance of precession.

By taking the inferred properties of the BBHs observed in the first, second and third observing runs [56], it is predicted that 90% of detected binaries will have mass ratios $q < 4$ and $\sim 97%$ of BHs in these binaries will have masses less than 45 $M_\odot$ [57]. Our “standard” simulated signal was chosen to have total mass $M = 40 M_\odot$ and mass ratio $q = 2$ inclined at an angle of $\theta_{\text{IN}} = 60°$. This corresponds to component masses of 26.7 $M_\odot$ and 13.3 $M_\odot$. This mass ratio and inclination was chosen to increase the observability of precession.

Of the 50 events reported by the LIGO/Virgo, 13 exclude the aligned-spin measure $\chi_{\text{eff}} = 0$ at 90% confidence [2,56,58]. The other 37 observations peak at $\chi_{\text{eff}} = 0$ [2,56]. Based on this, studies have shown that it is likely BHs in binaries have low spin magnitudes [19,44,59,60]. For this reason, in our standard configuration the BH spins were chosen such that there is zero spin aligned with the binary’s orbital angular momentum, $\chi_{\text{eff}} = 0$. We introduce precession by giving the more massive BH a spin of 0.4 in-plane and leaving the second BH with zero spin; two-spin effects are generally far weaker than the dominant precession effect, which exhibits the same phenomenology as a single-spin system [53,61]. From Eq. (2) we see that this gives us a system with $\chi_p = 0.4$. The opening angle for the binary when the signal enters the detector’s sensitivity band is 10° and the average value of the parameter $b = \tan(\beta/2)$ is $\bar{b} = 0.11$, from Eq. (5). The signal is simulated using the IMRPhenomPv2 waveform model that incorporates precession effects, but not higher harmonics ($\ell > 2$) in the signal [40,62].

Our “standard” simulated signal was chosen to be more favorable to precession measurements than typical LIGO-Virgo observations. Assuming systems are distributed
uniformly in binary orientation, masses drawn from a power law distribution and spins drawn from a low isotropic distribution (see Ref. [44] for details), we expect that 4 in every 100 binaries detected by LIGO-Virgo will be inclined at angles greater than 60° and have $b > 0.11$.

The sky location of the binary was chosen to have RA = 1.88 rad, DEC = 1.19 rad. The coalescence time is $t = 1186741861$ GPS (corresponding to the merger time of GW170814 [63]). The polarization angle, defined by the orientation of the orbital plane when entering the sensitive band at 20 Hz, is $\psi = 40^\circ$. The two harmonic approximation is calculated in the J-aligned frame ($\hat{z} = \hat{J}$). In this frame, the polarization angle is $\psi_j = 120^\circ$, which gives antenna factors for H1 of $F_+ = 0.34$ and $F_x = 0.53$ and for L1 of $F_+ = -0.45$ and $F_x = -0.30$, thus both detectors are roughly equally sensitive to the two GW polarizations.

We injected the signals into zero noise. The zero-noise analysis results will be similar to those obtained from the average results of multiple identical injections in different Gaussian noise realizations. The simulated signal is recovered using the LIGO Livingston and Hanford detectors with sensitivities matching those achieved in the second observing run (O2) [2]. A low frequency cut-off of 20 Hz was used for likelihood evaluations, this frequency is also used as the reference frequency when defining all frequency dependent parameters such as $\theta_{JN}$. Both the LIGO Livingston and Hanford sensitivities improved prior to the third observing run (O3) [64] and are expected to improve further prior to the fourth observing run (O4) [65]. The results presented in this work are unlikely to be affected significantly by these changes and therefore we expect the main conclusions to be valid for O4 and beyond.

The SNR of the signal is fixed to be 20, corresponding to a moderately loud signal for aLIGO and AdV observations [65]. This sets the distance to $d_L = 223$ Mpc. The simulated SNR in the two detectors is 16.2 in L1 and 11.7 in H1. The simulated precession SNR in each of the detectors is 3.7 and 3.4 respectively, giving a network precession SNR of 5.0. Thus, we expect that precession will be clearly observable in this signal.

**B. Parameter estimation techniques**

We will adopt a parameter estimation methodology that uses matched filtering with phenomenological gravitational waveforms and Markov Chain Monte Carlo (MCMC) techniques to sample the posterior.

We begin by introducing the matched filtering formalization for parameter estimation. We assume that the time series received from the GW detectors can be decomposed as a sum of the GW signal, $h(t)$, plus noise, $n(t)$, which is assumed stationary and Gaussian with zero mean,

$$d(t) = h(t) + n(t).$$

Under the assumption of Gaussian noise, the probability of observing data $d$ given a signal $h(\lambda)$ parametrized by $\lambda = \{\lambda_1, \lambda_2, \ldots, \lambda_N\}$, otherwise known as the likelihood, is [66],

$$p(d|\lambda) \propto \exp \left( -\frac{1}{2} \langle d - h(\lambda) | d - h(\lambda) \rangle \right). \tag{10}$$

where $\langle a|b \rangle$ denotes the inner product between two waveforms $a$ and $b$ and is defined as,

$$\langle a|b \rangle = 4 \text{Re} \int_0^\infty \tilde{a}(f) \tilde{b}^*(f) \frac{S_n(f)}{S_n(f)} \, df, \quad \tag{11}$$

where $S_n(f)$ is the one-sided power-spectral density (PSD) and $\tilde{a}$ denotes the Fourier transform of the gravitational waveform $a$.

The posterior probability density function (PDF) can then be computed through a simple application of Bayes’ theorem,

$$p(\lambda|d) = \frac{p(\lambda) p(d|\lambda)}{p(d)} \propto p(\lambda) \exp \left( -\frac{1}{2} \langle d - h(\lambda) | d - h(\lambda) \rangle \right), \tag{12}$$

where $p(\lambda|d)$ is the posterior distribution for the parameters $\lambda$, $p(\lambda)$ is the prior probability distribution where $\int p(\lambda) \, d\lambda = 1$, and $p(d)$ is the marginalized likelihood where $p(d) = \int p(\lambda|d) \, d\lambda$. Posterior distributions for specific parameters can then be found by marginalizing over all other parameters,

$$p(\lambda_i|d) = \int p(\lambda|d) \, d\lambda_1 \ldots d\lambda_{i-1} \, d\lambda_{i+1} \ldots d\lambda_N. \tag{13}$$

In the idealized situation of zero noise, Eq. (10) has a maximum at $h(\lambda) = h(\lambda_0)$. However, as can be seen in Eq. (12) the posterior also includes priors, this means that, as well as effects due to noise, certain priors may cause the maxima to be deflected away from $h(\lambda) = h(\lambda_0)$. This would then lead to Eq. (13) recovering a biased posterior. In this work, we consider the effect of three closely related priors,

(i) **Global**: the prior used during the parameter estimation analysis. This reflects our prior belief before observing any data,

(ii) **Conditioned**: the global prior conditioned upon the posterior distributions of other parameters from the same analysis. For example since $\chi_{eff}$ and $\chi_p$ are correlated, any informative measurement of $\chi_{eff}$ modifies our prior beliefs about $\chi_p$. This prior has been used in previous LVC publications, see e.g., [2].
(iii) Informed: the global prior conditioned upon the posterior distributions from a different analysis. Here, we use this to inform our expectations of the degree of precession given the results from a nonprecessing analysis. See Sec. VI for details.

C. Parameter recovery

We performed parameter estimation on the signal using the LALInference [67] and LALSimulation libraries within LALSuite [68]. Parameter recovery was performed with the IMRPhenomPv2 model [40,62], which matches the simulated signal to remove any systematic error caused by waveform uncertainty, and the corresponding IMRPhenomD aligned-spin waveform model [69,70], which does not include any precession effects. Additionally, all analyses used exactly the same priors as those used in the LIGO-Virgo discovery papers, for details, see Appendix B.1 of [2]. All post-processing was handled by the PESummary PYTHON package [71].

Table I summarizes the key results for the standard configuration. All uncertainties are the 90% symmetric credible intervals.

We begin by comparing the overall differences between parameter recovery with the precessing, IMRPhenomPv2, and nonprecessing, IMRPhenomD, runs. From the table, we see that the maximum likelihood SNR for the nonprecessing model is, as expected, lower than for the precessing waveforms. This can be easily understood from the two-harmonic approximation. Since the precessing waveform is well approximated by the sum of two non-precessing harmonics, we would expect the nonprecessing recovery to accurately recover the more significant of these two. If that were the case, the we would expect that,

$$\rho_D^2 \approx \rho^2 - \rho_p^2,$$

and this is indeed the case, as $\rho_D = 19.52$, $\rho = 19.94$ and the recovered power in the second harmonic is $\rho_p = 4.6$. Furthermore, we see that the recovered waveforms confirm this expectation: the recovered waveform when we include precession matches well with the simulated signal, while the nonprecessing run recovers a waveform that matches the dominant harmonic, as show in Fig. 3.

We first consider the accuracy with which the masses and (aligned) spins are recovered. As expected, the chirp mass of the system is well recovered, in that it matches the simulated value with only a 2% uncertainty, which remains constant for both precessing and nonprecessing runs. As is well known, there is a degeneracy between mass-ratio and spin, particularly during the inspiral part of the waveform [30–33], which leads to significant uncertainty in both parameters. In Fig. 4 we show the recovery of the mass ratio and spin, for both precessing and nonprecessing runs. When the model used to recover includes precession effects, the peak of the posteriors is located close to the simulated value ($\chi_{\text{eff}} = 0$ and $q = 2.0$) and, while the degeneracy leads to significant uncertainty in both parameters, the mass-ratio distribution is clearly peaked away from $q = 1$. Interestingly, when we recover with a nonprecessing waveform model, the inferred aligned spin component is systematically offset, with a peak at $\chi_{\text{eff}} \approx 0.05$. This can be understood by recalling that precession induces a secular drift in the phase evolution of the binary, and this can be mimicked by a change in the value of the aligned spin [38,43]. This discrepancy has not been seen in LIGO/Virgo observations [2] as we have not observed any systems with significant $\rho_p$ (see Fig. 1). We investigate this further in Sec. IV D, where we study the effect of varying the mass ratio.

For nonprecessing binaries, it is generally not possible to accurately recover the distance and orientation of the source, due to a well known degeneracy (see e.g., Ref. [29] for details), although the observation of higher signal harmonics can break this degeneracy through an independent measurement of the source inclination [27,29,72]. Similarly, the observation of precession can break this degeneracy [73]. Precession causes an oscillation of the orbital plane leading to a time-dependence of the orientation of the orbital plane relative to the line of sight. Equivalently, in the two-harmonic picture, precession leads to the observation of a second harmonic and, consequently, additional constraints on the binary orientation as the amplitudes of the harmonics depend upon the viewing angle. In Fig. 4, we show the inferred two-dimensional distance and inclination posteriors for the precessing and nonprecessing runs. As expected, the precessing run constrains the source to be away from face-on, while the nonprecessing run simply returns the prior. However, even with observable precession, the simulated distance and orientation are not accurately recovered—a significant fraction of the posterior support is for a system at a greater distance and oriented closer to face-on. We will see

**FIG. 3.** Comparison of the simulated precessing (green), nonprecessing maximum likelihood (red), precessing maximum likelihood (black) and dominant precessing harmonic (blue) waveforms as a function of frequency. Waveforms are projected onto the LIGO Hanford detector.
how these measurements improve with stronger precession in Sec. IV B.

The sky location of the source is not well recovered. The analysis was performed with only the two LIGO detectors, and therefore we expect to recover the source restricted to a ring on the sky, which corresponds to a fixed time delay between the detectors [74,75]. The location along the ring cannot be well constrained and, as expected the inferred location is preferentially associated with sky positions where the detector network is more sensitive. Thus, while the simulated sky location is within the 90% region, it is not at or close to the peak. This impacts the recovery of the distance, with the signal being recovered at larger distances, although the simulated distance remains within the 90% range. In Sec. IV G, we show results from a set of runs with varying sky location, and verify that at sky locations where the network is more sensitive, the distance posterior is more consistent with the simulated value.

Lastly, we turn to measurement of precession. In Fig. 5 we show the recovered distributions for binary orientation, θ_JN, precessing spin χ_p, initial precession phase, ϕ_JL, and precession SNR, ρ_p. There is a clear correlation between the inferred orientation and χ_p, with binaries that are more inclined having lower values of χ_p. Neither of these quantities are directly observable, it is only the amount of observable precession in the system, encoded by ρ_p, that can be measured. Thus the orientation and spin must combine to give the right amount of power in precession, and we see that this is the case—there is little correlation between the recovered values of ρ_p and the precessing spin χ_p. The inferred value of the precessing spin χ_p and precession SNR ρ_p are both consistent with the simulated values. Specifically, the signal has χ_p = 0.4 and this is consistent with the recovered value, although the posterior distribution is broad, with support over essentially the entire range from 0 to 1. The precession SNR peaks well away from zero, giving clear indication of precession in the system. However, the peak of the distribution occurs at 3.5, while the simulated value is 5.0. We have deliberately chosen an event with significant observable precession. Only a small fraction of the parameter-space volume leads to such significant precession as shown by the informed prior on Fig. 5. This is calculated by estimating the allowed values of ρ_p conditioned on the measurements from a nonprecessing analysis. See Sec. VI for further details.

The precession phase, ϕ_JL, while not measured with great accuracy, does show two peaks, which are consistent with the simulated value of 45° (0.8 rad). The precession phase can be inferred from the relative phase of the two precessing harmonics using Eq. (7), provided the binary orientation is well measured. There is a clear dependence with the binary orientation: if θ_JN < 90° then the peak is in ϕ_JL at the simulated value and if it is greater then ϕ_JL is offset by 180°, to compensate for the change in sign of the cos θ_JN terms in Eq. (7).

**IV. IMPACT OF VARYING PARAMETERS**

We now look at the effect of varying individual parameters one at a time on the recovered posteriors, in particular
In the strong-signal limit, where the likelihood surface can be well approximated by a multivariate gaussian, it is well known that the accuracy with which parameters can be measured is generally inversely proportional to the SNR [27,30]. However, this is not always the case due to, for example, degeneracies between parameters (see Ref. [76] for a discussion of the limits of this approximation).

Fig. 6 shows that as the SNR of the simulated signal increases, the accuracy and precision of the inferred \( \chi_p \) posterior distribution improves. As expected the width of the 90\% credible interval decreases approximately linearly with increasing SNR. The improvement in the \( \chi_p \) posterior distribution can be mapped to a linear increase in \( \rho_p \).

When the simulated signal has low SNR (\( \rho = 10 \)), the recovered \( \chi_p \) posterior distribution resembles the prior, implying that there is no information about precession in the data. For this case, \( \rho_p \) matches the expected distribution in the absence of any measurable precession—a \( \chi \) distribution with 2 degrees of freedom. As the SNR increases (\( \rho = 20–30 \)), the 5\% percentile of the \( \rho_p \) distribution is comparable or greater than the \( \rho_p = 2.1 \) threshold. This maps to the \( \chi_p \) posterior distribution removing all support for near-zero \( \chi_p \) (\( \chi_p \lesssim 0.1 \)). For larger SNRs (\( \rho > 40 \)), the entire \( \rho_p \) distribution is greater than the 2.1 threshold. This implies significant power from precession. For these cases, we remove support for maximal precession \( \chi_p \sim 1 \).

As expected we find good agreement between \( \rho_p \) and a noncentral \( \chi \) distribution with 2 degrees of freedom and noncentrality equal to the inferred power in the second harmonic (median of the \( \rho_p \) distribution).

### B. In-plane spin components

We now look at the effect of varying the amount of precession in the system, varying \( \chi_p \) from 0 to 1 in steps of 0.25. At \( \chi_p = 1 \) we have maximal spin, all in the plane of the binary. The inferred values of precessing spin and precession SNR are shown in Fig. 7. We observe, as expected, that increasing the in-plane spin leads to an increase in the magnitude of precession effects observable in the system. With zero precessing spin, there is no evidence for precession in the system; the recovered \( \chi_p \) is consistent with zero.\(^3\) Similarly, there is no support for significant precession SNR, with \( \rho_p \) constrained near zero. As \( \chi_p \) increases, the amount of precession in the system grows and the measurement of \( \chi_p \) becomes both more accurate and more precise. Figure 7 shows the relationship between \( \rho_p \) and \( \chi_p \), and a larger value for \( \rho_p \) enables a better measurement for \( \chi_p \).

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\(^3\)We do not expect the \( \chi_p \) posterior to contain \( \chi_p = 0 \) as there is no prior support there, however the posterior is relatively well constrained at low precession.

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**FIG. 5.** A corner plot showing the recovered values of binary orientation \( \theta_{IN} \), precessing spin \( \chi_p \), precession phase \( \phi_{JL} \) and precession SNR \( \rho_p \). Shading shows the 1\%, 3\% and 5\% confidence intervals. Black dots show the simulated values. The grey histograms show the informed prior, see Sec. VI. There is a clear correlation between the binary orientation and inferred precession spin, with signals which are close to face on (\( \cos \theta \approx \pm 1 \)) having larger values of precessing spin, while those which are more inclined having less precessing spin. The precession SNR only weakly correlated with \( \chi_p \).

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**A. SNR**

We start with the fiducial run configuration described above and vary the SNR of the simulated signal.
Fig. 8 shows how the inferred mass ratio–aligned spin and distance–orientation contours change as the magnitude of the in-plane spins change. When there is no observable precession in the system, there is a clear degeneracy in both cases. However, as precession effects become stronger the degeneracy between both pairs of parameters is broken. If \( \rho \) is small then this can be explained by both a small amount of precession observed at almost any inclination angle, or a large \( \chi \) observed close to face on, as seen in Fig. 5. Since precession effects are not strong enough to provide an accurate measurement of the orientation, the degeneracy between distance and \( \theta_{JN} \) persists. When \( \rho \) clearly excludes small values, there is no support for close to face-on signals, allowing a more precise measurement of the inclination angle \( \theta_{JN} \), breaking the degeneracy with distance.

Stronger precession also allows for improved measurement of the mass ratio. The opening angle \( \beta \), and consequently the precession parameter \( \bar{b} \), increases as the mass-ratio is increased, as can be seen from Eq. (1). Thus, when strong precession effects are observed, the signal is inconsistent with an equal mass system. In addition, the difference in frequency between the two leading precession harmonics depends upon the mass-ratio [43], and this may also improve our measurement of \( q \). This can also be seen from the precession dynamics, where the precession rate of \( L \) around \( J \), \( \alpha \), depends the mass ratio, and the number of observable precession cycles corresponds to improved accuracy in the measurement of the mass ratio [77].

As \( \chi \) is increased, the peak of the recovered \( \rho \) distribution is closer to the simulated value. This is likely due to a better measurement of the binary orientation as shown in Fig. 8.

C. Inclination

It is well known that the inclination angle will affect our ability to measure precession, as outlined in the discussion in Sec. II. In particular, from Eq. (7) we see that in the
two-harmonic approximation the second harmonic vanishes when \( \theta_{\text{JN}} = 0^\circ \) or \( 180^\circ \). In this section we consider the effect of changing the orientation of our standard configuration, which allows us to quantify how it will manifest in realistic LIGO-Virgo signals. A related study has looked at the effect at higher mass ratios \[52\].

The effect of varying \( \theta_{\text{JN}} \) is shown in Fig. 9. For binaries where the total angular momentum is nearly aligned with the line of sight, precession effects are not observable, as is clear from both the \( \rho_p \) and \( \chi_p \) posteriors. It is not until \( \theta_{\text{JN}} \geq 40^\circ \) that we begin to be able to measure precession. Although the accuracy of the measurement clearly improves as we increase \( \theta_{\text{JN}} \), the uncertainty in the measurement of \( \chi_p \) remains large and even at \( \theta_{\text{JN}} = 90^\circ \) the posterior is very broad. This can be understood by considering the degeneracies shown in Fig. 5 for the standard signal and in Fig. 10 for the \( \theta_{\text{JN}} = 90^\circ \) signal. In both cases, the measured quantity, \( \rho_p \), is relatively well constrained but neither the binary orientation nor \( \chi_p \) are accurately measured. The observed precession is consistent with both a highly inclined system with lower precessing spin (i.e., low \( \chi_p \) and large \( \theta_{\text{JN}} \)) or by a less inclined system with higher precessing spin (i.e., high \( \chi_p \) and small \( \theta_{\text{JN}} \)). Both of these will produce similar observable effects in the waveform.

This allows us to explain the measured posterior for \( \chi_p \). At low inclination the posterior is consistent with small

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FIG. 8. Two dimensional posteriors for (left) mass ratio and aligned spin, \( \chi_{\text{eff}} \), (right) binary orientation and distance. Contours show the 90% confidence interval. Bounded two-dimensional KDEs are used for estimating the joint probability density. The black circle with corresponding horizontal and vertical lines indicates the simulated values. For the simulated distance, a solid horizontal band indicates the maximum and minimum simulated values.

FIG. 9. Violin plots showing the recovered posterior distributions for \( \chi_p \) compared to its prior (left) and \( \rho_p \) (right). Distributions are plotted for varying \( \theta_{\text{JN}} \). Parameters other than \( \theta_{\text{JN}} \) match the “standard injection” (see Table I).
values of $\chi_p$. While we are unable to rule out large $\chi_p$, there is limited support as it would require the system to be observed very close to face-on, otherwise precession effects become significant. At large values of $\theta_{\text{JN}}$, when precession is clearly observable in the signal, $\chi_p = 0$ is excluded but the distribution remains broad and extends to $\chi_p = 1$.

D. Mass ratio and aligned spin

Figure 11 shows how the inferred precessing spin and precession SNR varies with the mass ratio of the system. As expected from the general considerations presented in Sec. II, as the mass ratio increases, an in-plane spin on the larger BH leads to a larger opening angle and more significant precession effects. For near equal-mass systems ($q \lesssim 1.5$), the inferred $\chi_p$ posterior distribution resembles its prior, and there is not significant power in precession, as shown by the value of $\rho_p$. As the mass ratio increases, the inferred power in precession also increases and for $q \gtrsim 2.5$, the 90% credible interval of the inferred $\rho_p$ distribution is entirely above $\rho_p = 2.1$. At this stage, precession is clearly identified and $\chi_p \approx 0$ is clearly excluded. In addition, the maximum value of $\chi_p$ is also bounded away from maximal.

Fig. 12 shows how varying $\chi_{\text{eff}}$ affects our ability to measure precession. A system with a large negative $\chi_{\text{eff}}$ results in a larger opening angle compared to an equivalent system with a large positive $\chi_{\text{eff}}$. Thus, based upon Eq. (1), we expect the observable impact of precession to be greater for negative values of $\chi_{\text{eff}}$ and smaller for positive values. The results are consistent with this expectation, in that the precession SNR decreases with increasing $\chi_{\text{eff}}$ and the width of the recovered $\chi_p$ distribution increases. However, for the $\chi_{\text{eff}} = 0.4$ analysis, we find that the range of $\chi_p$ is restricted, with both $\chi_p = 0$ and $\chi_p = 1$ excluded. This is not due to the measurement of precession, but is actually due to the measured nonzero aligned-spin component. A nonzero measurement of $\chi_{\text{eff}}$ forces $\chi_p < 1$ as the primary and secondary spin magnitudes must be less than unity. For example, in the $\chi_{\text{eff}} = 0.4$ analysis, we measure $\chi_{\text{eff}} = 0.38^{+0.07}_{-0.07}$. Under the single spin assumption, this limits $\chi_p < 0.95$. Similarly, since we are using prior distributions that are uniform in spin magnitude and orientation, the observation of a large aligned spin component leads to greater support for a large in-plane spin component. This is shown in Fig. 13, where we plot both the uninformed prior on the primary spin as well as the
prior conditioned on $\chi_{\text{eff}} = 0.4$, which removes all support for $\chi_p \approx 0$.

The $\chi_p$ measurement for the $\chi_{\text{eff}} = 0.27$ and 0.4 analyses are similar to the conditional prior but do restrict the lower $\chi_p$ bound beyond prior effects. Although the distribution for $\rho_p$ does extend to zero, it still peaks above $\rho_p = 2.1$ indicating some evidence, although not particularly strong, for precession.

As we vary the mass ratio and aligned spin, the length of the waveform will change. In particular, the aligned spin and high mass ratio configurations produce longer waveforms than those with anti-aligned spins and equal masses [78]. In principle, this will impact the measurability of precession, as longer waveforms allow for a greater number of precession cycles in the detectors’ sensitive band. For very short signals, with less than one precession cycle in band, the two leading harmonics are no longer orthogonal (or even approximately so), which make it more challenging to unambiguously identify the second harmonic. This is not an issue for the signals considered here, but does become important when we vary the mass of the binary in Sec. IV E. With a greater number of precession cycles, we will also be able to more accurately measure the precession frequency (the frequency difference between the harmonics), which may improve the measurement of mass ratio [77]. However, it is still the precession SNR that determines the observability of precession. Finally, we note that changing the mass ratio and aligned spin will change the overall amplitude of the waveform. Since our study is performed at a fixed SNR, this simply leads to the signals being placed at a larger or smaller distance and therefore does not impact the results presented here.

E. Total mass

We now vary the total mass of the system, keeping all other parameters including mass ratio fixed, in steps of 20 $M_\odot$. As before, we keep the SNR of the system constant at 20, so the higher mass systems are generated at a greater distance. The inferred distributions for $\chi_p$ and $\rho_p$ are shown in Fig. 14.

As the total mass of the source increases, the length of the waveform decreases, as does the number of precession cycles, with the number scaling approximately inversely to the total mass (see Eq. (45) of [38]). From the two-harmonic perspective, a small number of precession cycles leads to a large overlap between the harmonics. Specifically, for the $M = 100 M_\odot$ system the overlap
between the normalized harmonics is \( \langle \hat{h}_0 | \hat{h}_1 \rangle = 0.77 \), where \( \hat{h} = h / |h| \) and the inner product is defined in Eq. (11). At \( M = 20 M_\odot \), the harmonics are close to orthogonal with \( \langle \hat{h}_0 | \hat{h}_1 \rangle = 0.15 \). The opening angle does not change significantly, with \( \bar{b} = 0.14 \) at \( M = 20 M_\odot \) and \( \bar{b} = 0.21 \) at \( M = 100 M_\odot \).

At lower masses, \( M \leq 40 M_\odot \), while the precessing spin is not tightly constrained, it is clearly restricted to be nonzero and the precession SNR has essentially no support for \( \rho_p = 0 \). For the 60 \( M_\odot \) and 80 \( M_\odot \) mergers, the precessing spin is still peaked close to the simulated value while \( \rho_p \) peaks above 2.1 showing evidence for observable precession, although both \( \rho_p \) and \( \chi_p \) distributions do extend to zero.

For the high-mass system, \( M = 100 M_\odot \), the \( \chi_p \) posterior more closely matches the prior and we are unable to exclude \( \chi_p = 0 \). The inferred \( \rho_p \) distribution peaks close to zero, and is consistent with no precession, even though the precession SNR in the simulated signal is similar to the lower mass signals. This is likely due to the breakdown of the two-harmonic approximation for this short signal. In particular, for a high-mass system, the power orthogonal to the leading harmonic will depend sensitively upon the initial precession phase \( \phi_{1L} \). The fact that the recovered value of \( \rho_p \) is inconsistent with the simulated value may be due to this fact: the value of \( \phi_{1L} = 45^\circ \) used in the simulation leads to maximal observable precession. Across the full parameter space there are very few configurations with significant precession, so this observation is disfavored by our priors. We explore the prior effects such as this in detail in Sec. VI B.

### F. Polarization

The effect of changing the relative sensitivity to the two GW polarizations is clear from Eq. (7). Recalling that \( \bar{b} = 0.11 \) and \( \theta_{\psi N} = 60^\circ \), we can express \( \zeta \) (the ratio of the amplitudes of the two harmonics) as

\[
|\zeta| = 0.15 \left| \frac{F_+ + 2iF_\times}{F_+ + 0.8iF_\times} \right|.
\]

Thus, \( \zeta \), and consequently the imprint of precession on the waveform, will be maximized when the detector network is primarily sensitive to the \( \times \) polarization and minimized when the network is sensitive to the \( + \) polarization. We can investigate this by varying the polarization angle of the simulated signal, in steps of 10° from the “standard” value of 40°. At \( \psi = 40^\circ \), the sensitivity to the two polarizations is approximately equal, \( |F_\times| / |F_+| = 0.9 \). It is largest for \( \psi = 20^\circ \) where \( |F_\times| / |F_+| = 25 \) and smallest for \( \psi = 60^\circ \) where \( |F_\times| / |F_+| = 0.04 \). This leads to a variation in the precession SNR from \( \rho_p \approx 3 \) to \( \rho_p \approx 7 \).

In Fig. 15 we show the recovered posteriors for \( \chi_p \) and \( \rho_p \) for a set of runs where the precession is varied. The precession SNR varies in accordance with expectation—it is largest at \( \psi = 20^\circ \), where the median of the posterior is at \( \rho_p = 6 \) and there is no support for nonprecessing systems, and smallest at 60° where the posterior extends down to \( \rho_p = 0 \). The amount of observable precession directly impacts the inferred distribution for \( \rho_p \). For the \( \psi = 60^\circ \) signal, the posterior for \( \chi_p \) is consistent with zero, or small in-plane spins, and large values are excluded. Meanwhile for \( \psi = 20^\circ \), \( \chi_p < 0.1 \) is excluded while extremal in-plane spins are consistent with the observation.

It is well known that precession leaves a stronger imprint upon the \( \times \) polarization. However, we are not aware of previous results showing how simply changing the polarization of the system can so dramatically change the observable consequences of precession—from being barely observable when the observed signal is primarily the \( + \) polarization to being strongly observed in \( \times \). Using the
two-harmonic approximation, we are able to straightforwardly predict this effect and then verify it with detailed parameter estimation studies.

G. Sky location

We performed a series of runs where we altered the sky location of the signal, keeping the masses and spins of the components fixed. We also maintained the binary orientation $\theta_{JN} = 60^{\circ}$, but varied the distance and polarization of the source to ensure that the SNR remained constant and that the relative contribution of the $+$ and $\times$ polarizations was consistent with the standard run. Furthermore, sky locations were restricted to those for which the relative time of arrival between the Hanford and Livingston detectors remains the same (i.e., we were sampling from the nearly degenerate ring on the sky of constant time delays). Details of the runs are given in Table II.

Table II shows that the inferred luminosity distance remains approximately constant despite the simulated luminosity distance varying by almost a factor of two. In addition, the recovered $\rho_p$ distribution remains consistent with the “standard” injection. Figure 16 shows that the inferred sky position of the source remains essentially unchanged, and consistent with locations of the detectors’ greatest sensitivity. We note here that for this study we only considered the two detector LIGO network. Including VIRGO would likely have considerably improved the precision of the inferred sky location. We do not expect that this would affect any of the inferred physical parameters or any of the main conclusions in this work.

V. RELATING $\rho_p$ POSTERIORS TO BAYES FACTORS

An alternative method for identifying evidence for precession can be calculated within the Bayesian framework. We can calculate the Bayes factor, $B$, by comparing the marginalized likelihoods [see Eq. (12)] from two competing hypotheses (A, B) [79],

$$\ln B = \ln p(d_A) - \ln p(d_B).$$

Bayes factors have thus far been the gold standard for identifying evidence for precession within the GW community and have been used extensively in previous works, see e.g., Ref. [52].

In the same way that Bayes factors can be used to quantify evidence for precession, it is also possible to quantify the significance of a GW signal by calculating the Bayes factor for signal versus noise [80]. It has been shown that the log Bayes factor for signal versus noise scales approximately with $\rho^2$ [81]. Here, we investigate the relationship between the Bayes factor in favor of precession and the precession SNR $\rho_p$. Both of these quantities have been used together in recent works when assessing the evidence for observable precession [4,5,52].

For a subset of the runs described in Sec. IV C, we reran the analysis using the aligned-spin waveform model IMRPhenomD. Bayes factors in favor of precession could then be calculated and compared to the derived $\rho_p$ posterior distributions.

![Violin plots showing the recovered posterior distributions distributions for $\chi_p$ compared to its prior (left) and $\rho_p$ (right). Distributions are plotted for varying $\psi_J$. Parameters other than $\psi_J$ match the “standard injection” (see Table I).](image-url)
Fig. 17 shows an approximately linear relationship between the log Bayes factor (\(\ln BF\)) and the square of the precession SNR (\(\rho_p^2\)). This is expected given that the likelihoods recovered from the precessing waveform model will be larger than the likelihoods recovered from the aligned-spin waveform model by a factor of \(\exp(\rho_p^2/2)\).

The commonly used heuristic when assessing the strength of evidence using Bayes factors is that \(1 \leq \ln BF \leq 3\) is marginal evidence and \(\ln BF > 3\) is strong evidence in favor of a hypothesis. From the plots above we conclude that if 90% (50%) of the \(\rho_p\) posterior distribution is above the \(\rho_p = 2.1\) threshold, this corresponds to a \(\ln BF \approx 3.5\) (\(\ln BF \approx 0.8\)) and is therefore very strong (marginal) evidence for precession. The posterior distribution on \(\rho_p\) can therefore be approximately mapped to the commonly used \(\ln BF\). Assessing the strength of evidence for precession using \(\rho_p\) would also reduce the need for additional parameter estimation runs using nonprecessing models, which are necessary to compute the Bayes factor. This reduction in computational cost will not be significant for a single event, but for population analyses and large scale PE studies this alternative metric could be extremely useful.

VI. PREDICTING THE PRECESSION SNR POSTERIOR

For the majority of simulations presented in this paper, the distribution for the precession SNR, \(\rho_p\), has been peaked significantly below the simulated value, although in nearly every case the simulated value does lie within the 90% confidence region. While the naive expectation is that the recovered posterior will peak at the simulated value, for complex parameter recovery where there are dependencies and degeneracies between the different parameters, this is often not the case. We have already seen that the distance is typically overestimated in the simulations we have performed—this is a well-known effect and arises for two reasons, first that the network is less sensitive to sources from the chosen sky location than from other locations consistent with the observed signal (as discussed in Sec. IV G), and second that the signal was simulated significantly inclined from face-on, yet preferentially recovered close to face-on (as discussed in Sec. IV C). Similarly, it seems likely that the signals we have simulated have more significant precession effects (deliberately, as we wish to understand the observability of precession) than the vast majority of possible sources. Thus, our conjecture is that the likelihood peaks at the simulated value of \(\rho_p\) but the posterior distribution will be biased to recover a smaller value owing to the much larger volume of parameter space consistent with low \(\rho_p\). To demonstrate this, we calculate a prior distribution for \(\rho_p\) which uses the information gleaned from a nonprecessing analysis to take into consideration...
the much larger volume of parameter space consistent with low \( \rho_p \). We then show that when multiplying the likelihood by the prior, the predicted posterior for \( \rho_p \) agrees well with the inferred posterior from a fully precessing parameter estimation analysis.

Let us first show that the likelihood peaks at the simulated value of \( \rho_p \). The two-harmonic approximation allows us to factorize the likelihood in Eq. (16) into two terms: a nonprecessing component (dependent on \( h_0 \)) \( \Lambda_{np}(\lambda) \) and precessing component (dependent on \( h_1 \)) \( \Lambda_p(\lambda) \).

Previously, in Ref. [43], we obtained a distribution for \( p(\rho_p|d) \) by maximizing the likelihood over \( A_1 \). This is equivalent to assuming uniform priors for the real and imaginary components of \( A_1 \), and leads to a prior \( p(\rho_p) \propto \rho_p \). It follows from Eq. (19) that this results in a \( \chi^2 \) distribution with 2 degrees of freedom. Here, we instead use a prior for \( \rho_p \) which is informed by the information obtained from a nonprecessing analysis, we refer to this as the informed prior. This informed prior better represents our prior knowledge about \( \rho_p \) before explicitly accounting for precession in our analysis.

The majority of parameters required to calculate the informed prior are already given in the nonprecessing results. The two exceptions are the amplitude of the precessing spin \( \chi_p \) and the initial precession phase \( \phi_{jL} \). As discussed in Sec. IV D, we can obtain a prior for \( \chi_p \) conditioned upon the other parameters, specifically the mass ratio and aligned spin \( \chi_{eff} \), and this can be used to generate the informed prior on \( \rho_p \). The initial precession phase is unconstrained by the non-precessing parameter recovery, this then allows us to assume it to be uniformly distributed. By calculating the predicted posterior distribution for \( \rho_p \) based upon a set of nonprecessing samples, we may examine the effect of other measured parameters on the final \( \rho_p \) distribution. For example, if the aligned-spin run favors a binary that is close to equal mass and an orientation consistent with a face-on system, then our prior belief will be that the precessing SNR will be low—it is only with unequal masses and systems misaligned with the line of sight that there are significant precession effects in the observed waveform. A prior belief of \( \rho_p \) peaking at low values will cause the predicted \( \rho_p \) to peak at values lower than the simulated one and consequently so too will the inferred posterior distribution for \( \rho_p \) inferred from a full 15-dimensional parameter estimation analysis.

A. Precessing signal

We now apply this conjecture to a precessing signal by attempting to predict the posterior distributions for \( \rho_p \).
This allows us to investigate how much our recovered posterior distributions may differ from the idealized case of a precession likelihood function distributed about the simulated (true) value. In Fig. 18 we show the results of this for the $q = 4$ simulation presented in Sec. IV D. This specific simulation was chosen since this case has the largest $\rho_p$ and corresponds to a simulation where a non-precessing analysis is less justified. It is therefore a good case to show how the combination of the informed prior and the additional likelihood from precession $\Lambda_p$ correctly estimates the large $\rho_p$. The prior strongly disfavors large observable precession and therefore pulls the posterior toward smaller values than the simulated value i.e., where the likelihood function peaks.

In Fig. 19, we show a comparison between the predicted and measured $\rho_p$ distributions for the set of runs with varying mass ratio presented in Sec. IV D. When we calculate the posterior, explicitly accounting for the parameter space weighting encoded in the informed prior on $\rho_p$, we find good agreement between the predicted and the inferred $\rho_p$ distributions and note that neither predicted nor inferred are centered around the true value for the set of signals that we have simulated. Of course, if we were to draw signals uniformly from the prior distribution, we would expect to observe the inferred distributions of $\rho_p$ matching with the simulated values.

### B. Nonprecessing signal

We now look at the expected posterior distribution for $\rho_p$ when there is no precession in the signal. As explained in Sec. VI, previously a $\chi^2$ distribution with two degrees of freedom was used to model the $\rho_p$ distribution in the absence of any precession (see Ref. [43]). This then led to the natural heuristic that $\rho_p = 2.1$ should be the threshold for observable precession. Using Eq. (19) we can now use a more informative prior on $\tilde{\rho}_p$ and obtain a more accurate estimate of the expected posterior distribution in the absence of precession. We do this by using parameter estimation samples from an aligned-spin model and setting the simulated precession SNR to be 0, this then allows us to account for the effects of priors and different noise realizations.

In Fig. 20 we show the predicted and observed distributions for the precession SNR $\rho_p$ for a nonprecessing signal. We use a nonprecessing equivalent of the “standard” injection as our simulated signal (i.e., we set $\chi_p = 0$ while ensuring all other parameters match those in Table 1). We inject with zero noise and use the IMRPhenomPv2 model for parameter recovery.

The inferred $\rho_p$ distribution is peaked at lower values that the $\chi^2$ distribution as shown in Fig. 20. However using the prediction from the likelihood [Eq. (16)] and the

![Figure 18](image1.png)

**FIG. 18.** The predicted distribution for the precession SNR $\rho_p$ (dashed orange) calculated as the product of the precessing contribution to the likelihood (black dotted line) and the informed prior of $\rho_p$ (blue) for the $q = 4$ simulation presented in Sec. IV D. For comparison, we show the inferred $\rho_p$ posterior distribution from the full 15 dimensional parameter estimation analysis (solid orange) and $\rho_p$ for the injection (red line). The informed prior is peaked at low values of $\rho_p$ causing the peak of the posterior to be smaller than the maximum likelihood value.

![Figure 19](image2.png)

**FIG. 19.** Violin plot comparing the observed $\rho_p$ distribution (colored) from a precessing analysis, and the predicted distribution (white) based on the aligned-spin results and simulated value of $\rho_p$ for the set of varying mass ratio simulations presented in Sec. IV D. The predicted and observed distributions for precession SNR are in good agreement, even though the $\rho_p$ in the simulated signal (red lines) lies above the peak of either distribution.
In most candidate astrophysical binary distributions, precession is likely to be first measured in a comparable-mass binary \[q = 2\]. We have considered a fiducial example of such a possible signal (mass-ratio \(q = 2\), SNR \(\rho = 20\), and in-plane spin \(\chi_p = 0.4\), such that the precession contribution to the total SNR is \(\rho_p = 5\)), and performed an extensive parameter-estimation study that has systematically explored the impact on parameter measurements of changes in each of the key source parameters: the SNR, the in-plane spin magnitude, binary inclination, the binary mass ratio and aligned-spin contribution, the binary’s total mass, the polarization, and sky location. These examples illustrate well-known features of precession signals [38,44–46,48–51], and quantify their effect on both the measurement of precession, and their impact on the measurement accuracy and precision of other parameters.

We have also verified that \(\rho_p\) provides a suitable and intuitive metric for determining whether or not we have measured precession, and shown that there is an approximate mapping between \(\rho_p\) and the use of the Bayes factor to assess the evidence of precession. We suggest that given these results, future large scale studies of precession can be made considerably computationally cheaper by computing \(\rho_p\), rather than a full Bayesian analysis.

We note that as \(\rho_p\) captures precession by identifying additional power beyond a simple nonprecessing waveform model, it could therefore be effected by phenomena such as eccentricity and higher order multipoles. As BFIs simply compare the evidence for two models, one precessing and one nonprecessing, using BFIs as the sole metric would also be biased by properties like eccentricity and higher order multipoles.

However, a similar approach to the 2-harmonic decomposition for precessing signals has recently been applied to GWs including the effects of higher harmonics [82]. In future work, we will combine these approaches and explore the measurability of precession in systems with significant evidence for higher harmonics, and the impact of the combination of higher modes and precession upon parameter accuracy. It may also be possible to account for eccentricity through a similar decomposition.

As highlighted in Sec. VI these decompositions provide powerful insights into how the addition of physical phenomena introduce information into the analysis. Here we show that the likelihood can be simply factored into precessing and nonprecessing contributions. This then allows us quantify the extra information that can be gained from a precessing analysis and even predict the recovered \(\rho_p\) distribution with or without these effects taken into consideration in the analysis.

The current study does not include higher harmonics, and uses a signal model (IMRPhenomPv2) that neglects two-spin precession effects, mode asymmetries that lead to out-of-plane recoil [83], and detailed modelling of precession effects through merger and ringdown. Although these effects are typically small, so is the imprint of precession on the signal, and it would be interesting in future to investigate the impact of these additional features on our results. We also emphasize that, although we consider it to be extremely useful to provide quantitative
examples of the effects of each of the binary parameters, these will necessarily depend on the location in parameter space of our fiducial example. However, having chosen a configuration from among what we expect to be the most likely signals, we hope that these examples will act as a useful guide in interpreting precession measurements when they arise in future gravitational-wave observations.

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