Parametrized tests of the strong-field dynamics of general relativity using gravitational wave signals from coalescing binary black holes: Fast likelihood calculations and sensitivity of the method

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Thanks to the recent discoveries of gravitational wave signals from binary black hole mergers by Advanced Laser Interferometer Gravitational Wave Observatory and Advanced Virgo, the genuinely strong-field dynamics of general relativity (GR) can now be probed, allowing for stringent tests of general relativity (GR). One set of tests consists of allowing for parametrized deformations away from GR in the template waveform models and then constraining the size of the deviations, as was done for the detected signals in previous work. In this paper, we construct reduced-order quadratures so as to speed up likelihood calculations for parameter estimation on future events. Next, we explicitly demonstrate the robustness of the parametrized tests by showing that they will correctly indicate consistency with GR if the theory is valid. We also check to what extent deviations from GR can be constrained as information from an increasing number of detections is combined. Finally, we evaluate the sensitivity of the method to possible violations of GR.

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I. INTRODUCTION

Since 2015, the twin Advanced Laser Interferometer Gravitational Wave Observatories (Advanced LIGOs) [1] have routinely been detecting gravitational wave signals from coalescing binary black holes [2–6], recently also in conjunction with Advanced Virgo [7,8]. Later in the decade, the worldwide gravitational wave detector network will be extended with the Japanese KAGRA [9], to be followed by LIGO-India [10]. In the course of the next several years, tens to hundreds more binary black hole detections are expected to be made [4].

Coalescences of stellar mass binary black holes (BBHs) are ideal laboratories for testing the genuinely strong-field dynamics of general relativity (GR) [11,12]: they are likely to be pure spacetime events, involving stronger curvatures and shorter dynamical time scales than in any other experiment or observation, by many orders of magnitude [13]. The process starts with two black holes that are orbiting each other, gradually losing orbital energy and
Given these restrictions, at the present time, it is expedient to devise tests of the theory of general relativity itself, which to the largest extent possible are generic and as accurate as we can make them. Following the recent binary black hole merger detections, a battery of such tests were deployed [4,5,12]: looking for coherent excess signal power in the data after subtraction of the best-fitting GR waveform [38,39], checking for consistency with GR between the pre- and postmerger signals in terms of masses and spins [40,41], evaluating consistency of the postmerger signal with the presence of a least-damped ringdown mode [12], constraining anomalous GW propagation with a view on bounding the mass of the graviton as well as violations of local Lorentz invariance [42,43] (the latter also using the binary neutron star detection [44–46]), looking for evidence of nonstandard polarization states [47], and measuring a series of judiciously chosen coefficients associated with parametrized deformations of IMR waveforms away from GR [4,5,12,36,48–54]. This paper deals primarily with the latter tests.

As mentioned above, a number of IMR waveform models have been developed. For parametrized tests of GR, we use the phenomenological models, which have a closed expression in the frequency domain and hence can be generated fast on a computer (which is important for data analysis purposes when exploring high-dimensional parameter spaces), capture the essential physics of the problem (including, e.g., spin-induced precession), and allow for some amount of analytic insight into the meaning of the induced deformations. In particular, we use the model which in the LIGO Algorithm Library is designated as IMRPhenomPv2 [31–33]. The IMRPhenomPv2 waveform form is characterized by a number of parameters \( \{p_i\} \): (i) in the adiabatic inspiral regime, PN coefficients \( \{q_{\phi_0}, \ldots, q_{\phi_7}\} \) and \( \{q_{\phi_5}, q_{\phi_6}\} \); (ii) in the intermediate regime between adiabatic inspiral and merger, phenomenological coefficients \( \{\beta_0, \ldots, \beta_3\} \); and (iii) in the merger-ringdown regime \( \{a_0, \ldots, a_5\} \). In the most relevant of these coefficients, parametrized deformations are introduced by allowing for relative deviations: \( p_i \rightarrow (1 + \delta p_i) p_i \). The \( \delta p_i \) will be referred to as our testing parameters.

We then perform a series of tests, in each of which some testing parameter \( \delta p_j \) is allowed to vary freely along with all other parameters entering the phase (component masses and spins, which enter through the GR expressions for the \( p_i \) themselves), but \( \delta p_k = 0 \) for \( k \neq j \). In principle, multiple \( \delta p_i \) could be allowed to vary at the same time, but this will lead to a degradation in the measurement accuracy for all of them [12]; statistical errors will be much smaller when the \( \delta p_i \) are varied one at a time. Note that, in most alternative theories of gravity, a violation will likely show up in more than one coefficient. However, as already demonstrated in [55] in a PN context, looking for a deviation from zero in a single testing parameter is an efficient way to search for GR violations that occur at multiple PN orders, and one can
even find violations at powers of frequency that are distinct from the one that the testing parameter is associated with [52,53]. Of course, if a deviation is present, then the individual measurements of the $\delta \hat{p}_i$ will not necessarily reflect the predicted values of the correct alternative theory. Should one want to measure or constrain, e.g., extra charges or coupling constants that may be present in one’s favorite alternative theory using an IMR signal, then an accurate IMR waveform model would need to be constructed for that particular theory. However, this is not the aim of the framework presented here; what we want to do is test Einstein’s theory itself by constraining deviations from the theory.

Even though the IMRPhenomPv2 waveform model has an explicit analytic expression, in the case of low-mass binary mergers, which leave a long signal in the detectors’ sensitive band, the analyses are computationally costly and can take more than a month of time to complete, due to the large number of likelihood evaluations $[O(10^8)]$ that must be performed. Given the large number of detections that are expected to be made in the coming LIGO-Virgo observing runs, ways must be found to reduce the computational burden. One solution is to speed up the likelihood calculation by constructing reduced-order quadratures (ROQs) [56–58], which in turn are based on reduced-order models [59–66]. In the method of reduced-order quadratures, the discrete overlap calculation involved in the likelihood evaluation is split up into a data dependent sum, which only needs to be evaluated once for each detection, and a much shorter sum that takes care of the parameter dependent part of the overlap calculation that must be performed many times during the sampling over parameter space. In line with the method outlined above, a series of ROQs is created, in each of which a single testing parameter $\delta \hat{p}_i$ is allowed for.

Results of the parametrized tests for the LIGO-Virgo detections of binary black hole coalescences have already appeared elsewhere [4,5,12]. The aim of this paper is twofold: (a) to construct ROQs for IMRPhenomPv2 waveforms with parametrized deformations, for use on future detections, and (b) to explicitly demonstrate the robustness and sensitivity of the method as a whole, which had not yet been done in previous publications.

The structure of this paper is as follows. In Sec. II, we briefly recall the waveform model used and explore analytically how the phase varies with the chosen deformations and as a function of mass. We describe the setup of the parametrized tests and explain how results from multiple detections can be combined to arrive at stronger bounds on GR violations. Section III describes the construction of the reduced-order quadrature for waveforms with testing parameters. Next, in Sec. IV, we present some checks of the correctness and robustness of the data analysis pipeline. In Sec. V, we show how well the parametrized tests can bound GR violations by combining information from all available sources. Furthermore, we investigate how testing parameters in the template waveforms respond when deviations in one or more parameters are present in the signal. Section VI provides a summary and conclusions.

Throughout this paper, we set $c = G = 1$ unless specified otherwise.

II. WAVEFORM MODEL AND PARAMETRIZED TESTS

A. Waveform model

The starting point for the parametrized tests is the phenomenological frequency domain waveform model which in the LIGO Algorithm Library is designated as IMRPhenomPv2 [33]. This waveform model describes an approximate signal of a precessing binary by applying a rotation transformation [33,67] to an underlying aligned spin waveform model, here taken to be IMRPhenomD [31,32]. The orbital precession dynamics are given in terms of an effective spin parametrization [33,68]. For an in-depth description, we refer to these papers; here we only give a quick overview.

The phasing of IMRPhenomPv2 consists of three regimes, whose physical meaning and parametrization are as follows:

1. The inspiral regime is parametrized by post-Newtonian coefficients $\{\varphi_0, \ldots, \varphi_7\}$ and $\{\varphi_{5i}, \varphi_{6i}\}$, as well as phenomenological parameters $\{\sigma_0, \ldots, \sigma_4\}$. The latter are contributions at high effective PN order (up to 5.5 PN) to correct for nonadiabaticity in late inspiral and for unknown high-order PN coefficients in the adiabatic regime.

2. The intermediate regime transitions between inspiral and merger-ringdown; it is parametrized by the phenomenological coefficients $\{\beta_0, \ldots, \beta_3\}$.

3. The merger-ringdown regime is parametrized by a combination of phenomenological and analytical black hole perturbation theory parameters $\{\alpha_0, \ldots, \alpha_5\}$.

Note that the PN coefficients $\{\varphi_0, \ldots, \varphi_7\}$ and $\{\varphi_{5i}, \varphi_{6i}\}$ have their usual dependences on the binary’s component masses and spins. The other phenomenological parameters are fixed by calibration against numerical relativity waveforms. For the functions of frequency in which the above parameters appear, we refer to [32]; see also Table I in [12]. The transition from the inspiral to the intermediate regime happens at a frequency $f = f_1 = 0.018/M$ (where $M$ is the total mass) and from the intermediate to the merger-ringdown regime at $f = f_2 = 0.5f_{RD}$, with $f_{RD}$ a “ringdown frequency”, in such a way that the waveform is $C^1$ continuous.

Figure 1 shows the modulus of the waveform $|\tilde{h}(f)|$, highlighting the three regimes; also shown is the Fourier transform to the time domain $\tilde{h}(t)$ and the corresponding instantaneous frequency as a function of time.
Not all of the coefficients mentioned above will be used in the parametrized tests. In the inspiral regime, \( \varphi_5 \) is completely degenerate with the phase at coalescence \( \varphi_c \); similarly, the pair \((\alpha_0, \sigma_1)\) is degenerate with \((\varphi_c, t_c)\), with \( t_c \) as the time at coalescence. The pairs \((\beta_0, \beta_1)\) and \((\alpha_0, \alpha_1)\) are set by the requirement of C\(^1\) continuity between the different regimes. We also omit \( \alpha_4 \), which occurs in the same term as \( \alpha_3 \), meaning that there will be some amount of degeneracy between the two. Finally, we do not use \( \{\sigma_2, \sigma_3, \sigma_4\} \), whose fractional calibration uncertainties were larger \([(\text{a few}) \times 10^{-1}] \) than those of the other phenomenological parameters \([(\text{a few}) \times 10^{-2}] \), though all calibration uncertainties were observed to be below measurement uncertainties for the binary black hole coalescence detections that were made [69].

The way our parametrized tests are implemented is by allowing for fractional deviations from the GR values for all of the remaining coefficients \( p_i \) in turn

\[
p_i^{GR}(m_1, m_2, S_1, S_2) \to (1 + \delta p_i) p_i^{GR}(m_1, m_2, S_1, S_2),
\]

where \( m_1, m_2 \) are the component masses and \( S_1, S_2 \) are the component spins; one has

\[
\{\delta p_1\}_i = \{\delta \varphi_0, \ldots, \delta \varphi_7, \delta \varphi_{5i}, \delta \hat{\phi}_2, \delta \hat{\phi}_3, \delta \hat{\alpha}_2, \delta \hat{\alpha}_3, \delta \hat{\alpha}_4\}.
\]

We note that in GR, \( \varphi_1 \equiv 0 \), so that as an exception, we let \( \delta \varphi_1 \) be an absolute rather than a relative deviation.

Including extrinsic parameters coming from the detector response, in practice, the full parameter sets of the resulting waveform models will be

\[
\mathbf{\tilde{\lambda}} = \{t_c, \varphi_c, D_L, \theta, \phi, \psi, m_1, m_2, \chi_1, \chi_2, \chi_p, \theta_f, \alpha_0, \delta \hat{p}_i\}.
\]

Here, \( t_c \) and \( \varphi_c \) are, respectively, the time and phase at coalescence; \( D_L \) is the luminosity distance; \((\theta, \phi)\) give the sky position; \( \psi \) is a polarization angle; \( m_1 \) and \( m_2 \) are the component masses; \( \chi_1 \) and \( \chi_2 \) are spin magnitudes; and \( \chi_p \) is an “effective” spin precession parameter given by [68]

\[
\chi_p = \frac{\max(A_1 m_1^2 \chi_{1\perp}, A_2 m_2^2 \chi_{2\perp})}{A_1^2 m_1^2},
\]

where \( A_1 = 2 + 3 m_2/2 m_1, \ A_2 = 2 + 3 m_1/2 m_2, \) and \( \chi_{1\perp}, \chi_{2\perp} \) are the projections of the spin vectors onto the orbital plane, i.e., orthogonal to the direction of the orbital angular momentum \( \hat{\lambda} \) at a specific reference frequency \( f_{\text{ref}} \); \( \theta_f \) is the angle between the line of sight \( \hat{\eta} \) and the total angular momentum \( \hat{\lambda} \) at \( f_{\text{ref}} \), and \( \alpha_0 \) indicates the azimuthal orientation of \( \hat{L} \) at \( f_{\text{ref}} \) [33].

**B. Effect of testing parameters on the phase**

Before going on to evaluate the sensitivity of parametrized tests given stellar mass BBHs as seen in the advanced detectors, we first illustrate the effect on the phase of varying the \( \delta \hat{p}_i \). As it turns out, one of the best-determined PN testing parameters tends to be \( \delta \hat{\phi}_3 \); in the intermediate regime this is \( \delta \hat{\phi}_2 \) and in the merger-ringdown regime it is \( \delta \hat{\alpha}_2 \); these are the parameters we focus on.

Figure 2 shows how the phase as a function of frequency varies with testing parameters \( \Psi(\delta \hat{p}_i; f) \), as well as the difference with the phase in GR, \( \Delta \Psi(\delta \hat{p}_i; f) = \Psi(\delta \hat{p}_i; f) -\Psi_{GR}(f) \), for \( t_c = \varphi_c = 0 \). Two kinds of sources are considered, with masses and spins chosen to be the means of the posterior density functions for the signals that were designated GW150914 [2] and GW151226 [3]. The phases and their differences are plotted from \( f_{\text{low}} = 20 \) Hz and up to a frequency where the dominant \((l = 2, m = 2)\) mode of the ringdown signal can be safely assumed to have ended \((600 \text{ Hz for GW150914 and 800 Hz for GW151226}) \). The qualitative behavior is as expected given the differences between the two. GW150914, being more massive, had a short inspiral regime and the merger occurred at \( f \sim 130 \text{ Hz} \), close to the frequency where the detectors are the most sensitive. By contrast, GW151226 had a much
longer inspiral (with $\sim 55$ cycles in band) and its merger occurred at $f \sim 450$ Hz. For the PN testing parameter $\delta \phi_3$, a much larger phase difference is accumulated in the case of GW151226, which will cause this parameter to be much better measured in the latter case despite the overall smaller signal-to-noise ratio. The intermediate regime parameter $\delta \phi_2$ exhibits a relatively slowly increasing phase difference and levels out between 100 and 200 Hz for both events, which is where the detectors are the most sensitive; hence, we can expect it to be roughly equally well measurable for both events. Finally, for $\delta \alpha_2$, in the case of GW150914, the phase difference reaches a maximum at some point before decreasing again, while for GW151226, the difference keeps increasing up to high frequencies, but not with a larger maximum phase difference; hence, this parameter will be better measurable with GW150914, for which the merger-ringdown regime occurs at frequencies closer to the range of best detector sensitivity. These expectations are borne out by the published results for the two events [4,5,12]. Note that varying the $\delta \beta_i$ has an effect at all frequencies; this is a consequence of the $C^1$ junction conditions between the inspiral, intermediate, and merger-ringdown regimes.
In Fig. 3, we illustrate the phase differences as chirp mass \( \mathcal{M}_c = M\eta^{3/5} \) is varied (where \( M = m_1 + m_2 \) and \( \eta = m_1 m_2/M^2 \)), for particular values of the \( \delta \hat{\rho}_i \); the symmetric mass ratio is fixed at \( \eta = 0.2 \) and again \( t_c = \varphi_c = 0 \); shown are the \( \Delta \Psi(\delta \hat{\rho}_i; f) \) for \( f = 150 \text{ Hz} \), i.e., close to the frequency of optimal sensitivity for the Advanced LIGO detectors. Again, the behavior is as expected. Low \( \mathcal{M}_c \) corresponds to waveforms with significant inspiral in band; deviations in the \( \varphi_i \) then have a large effect on the observable phase. Deviations in the intermediate regime parameters \( \beta_i \) have the largest effect when this regime occurs at frequencies where the detectors are the most sensitive, which corresponds to \( \mathcal{M}_c = 10-20 \, M_\odot \). Finally, shifts in the merger-ringdown parameters \( \alpha_i \) have their largest effect for \( \mathcal{M}_c \gtrsim 20 \, M_\odot \), which brings this regime in the detectors’ most sensitive band. For completeness, we show results for zero spins, as well as aligned spins with \( |S_{1,2}| = 0.9 \); the well-known effect of “orbital hang-up” [70], which prolongs the duration of waveforms in the time domain, then causes similar features to occur at higher \( \mathcal{M}_c \).

### C. Parameter estimation

Parameter estimation is done using the LALInference framework [71,72], in which the posterior density distribution for the parameters \( \lambda \) is obtained as

\[
p(\lambda|H_i,d,I) = \frac{p(\lambda|H_i,I) p(d|H_i,\lambda,I)}{p(d|I)}. \tag{5}
\]

Here, \( H_i \) is the hypothesis corresponding to the waveform model in which \( \delta \hat{\rho}_i \) is an extra free parameter, \( d \) are the data, and \( I \) denotes whatever background information we may have; \( p(d|H_i,\lambda,I) \) is the likelihood function, which up to an overall prefactor is given by

\[
p(d|H_i,\lambda,I) \propto \exp[-(d-h(\lambda)|d-h(\lambda))/2], \tag{6}
\]

with \( h(\lambda) \) the signal model described in Sec. II A and \( \langle \cdot | \cdot \rangle \) as the noise-weighted inner product

\[
\langle a|b \rangle = 4\Re \int_{f_{\text{low}}}^{f_{\text{high}}} \frac{a^*(f) b(f)}{S_n(f)} \, df, \tag{7}
\]

where \( \Re \) denotes the real part, and \( S_n(f) \) is the one-sided noise spectral density. For the second-generation detectors, the lower cutoff frequency is taken to be \( f_{\text{low}} = 20 \, \text{Hz} \), while \( f_{\text{high}} = 2048 \, \text{Hz} \) suffices as an upper cutoff frequency for stellar mass BBH coalescences. The likelihood function is evaluated using the efficient nested sampling algorithm [71]. \( p(\lambda|H_i,I) \) is the prior probability density for the free parameters; for those parameters that also appear in the GR waveform, these are chosen in the same way as in [72], while for \( \delta \hat{\rho}_i \), we choose priors uniform in an interval that is wide enough to contain the supports of the posterior densities; suitable ranges are given in Sec. III B below. Finally, \( p(d|I) \) is the probability of the data, which can be absorbed into an overall normalization factor for the posterior density \( p(\lambda|H_i,d,I) \).

To obtain one-dimensional posterior densities for the parameters \( \delta \hat{\rho}_i \), one marginalizes over all other parameters

\[
p(\delta \hat{\rho}_i|H_i,d,I) = \int d\hat{\rho} p(\hat{\rho},\delta \hat{\rho}_i|H_i,d,I) \tag{8}
\]
where the integration is performed over all parameters in (3) except for $\delta \hat{p}_i$.

Finally, posterior densities from individual events can be conveniently combined to arrive at stronger bounds on the $\delta \hat{p}_i$ under the assumption that the fractional deviations are the same in each event. Assuming independent detections $d_1, d_2, \ldots, d_N$, it is easy to see that

$$
p(\delta \hat{p}_i | H_i, d_1, d_2, \ldots, d_N) = p(\delta \hat{p}_i | H_i)^{1-N} \prod_{n=1}^{N} p(\delta \hat{p}_i | H_i, d_n, I).
$$

(9)

For events with similar signal-to-noise ratios and in the absence of measurement offsets, one can expect the widths of these posteriors to decrease roughly with $\sqrt{N}$.

### III. REDUCED-ORDER QUADRATURES FOR FAST LIKELIHOOD CALCULATIONS

#### A. Basic method

We now proceed to constructing reduced-order quadratures. The technical underpinnings have already been discussed in detail elsewhere [57,58]; here we will only give an overview.

The first step is to approximate the waveform $h(\tilde{\lambda}; f)$ as

$$h(\tilde{\lambda}; f) \approx P_{E_n} [h(\tilde{\lambda}; f)] = \sum_{i=1}^{n} (e_i | h(\tilde{\lambda})) e_i(f),
$$

(10)

where the vectors in the reduced basis $E_n = \{e_i(f)\}_{i=1}^{n}$ are orthonormal with respect to the inner product

$$(a|b) \equiv \int_{f_{min}}^{f_{max}} a^*(f) b(f) df,
$$

(11)

and the approximation is good to within a greedy projection error $\epsilon$:

$$||h(\tilde{\lambda}; f) - P_{E_n} [h(\tilde{\lambda}; f)]||^2 < \epsilon,
$$

(12)

for $\tilde{\lambda} \in \mathcal{T}_N$, where $\mathcal{T}_N$ is a suitably large training set and $||a|| \equiv \sqrt{a^* a}$. From this, one constructs an empirical interpolant to approximate the waveform

$$I_n[h](\tilde{\lambda}; F_k) = \sum_{i=1}^{n} x_i(\tilde{\lambda}) e_i(f),
$$

(13)

where the coefficients $x_i$ are solutions to

$$I_n[h](\tilde{\lambda}; F_k) = h(\tilde{\lambda}; F_k)
$$

(14)

at interpolation points $\{F_k\}_{k=1}^{n}$. Defining the matrix $A_{ij} = e_j(F_i)$, one has

$$I_n[h](\tilde{\lambda}; F_k) = \sum_{i=1}^{n} \sum_{k=1}^{n} (A^{-1})_{ik} h(\tilde{\lambda}; F_k)e_i(f)
$$

(15)

where

$$B_k^i(f) = \sum_{i=1}^{n} (A^{-1})_{ik} e_i(f).
$$

(16)

The $\{F_k\}_{k=1}^{n}$ are chosen from a set $\{f_i\}_{i=1}^{L}$, where $L$ is related to the duration $T$ of the longest waveform considered through

$$L = (f_{max} - f_{min}) T + 1,
$$

(17)

and the $f_i$ are spaced by $\Delta f = 1/T$. The first interpolation point $F_1$ is chosen such that it maximizes the amplitude of the first reduced basis vector, i.e., $|e_1(F_1)| \geq |e_1(f_i)|$ for all $f_i$. Next, one builds an interpolant of $e_2(f)$ using only $e_1$ and $F_1$, and one finds an $F_2$ that maximizes the pointwise interpolation error, i.e., $|I_1[e_2](F_2) - e_2(F_2)| \geq |I_1[e_2](f_i) - e_2(f_i)|$ for all $f_i$. One then continues in this fashion until $n$ interpolation points have been obtained.

Though the interpolant $I_n[h](\tilde{\lambda}; f)$ can be evaluated at any parameter values $\tilde{\lambda}$, the underlying reduced basis $E_n$ satisfies the tolerance criterion (12) only for $\tilde{\lambda} \in \mathcal{T}_N$. Next comes the validation step, where the accuracy of the interpolant is evaluated also for values $\tilde{\lambda}$ that lie outside the training set (though inside the same ranges as for the training set, where the waveform approximant is deemed valid). Arbitrary values are picked, for which it is checked that

$$||h(\tilde{\lambda}; f) - I_n[h](\tilde{\lambda}; f)||^2 < \beta
$$

(18)

for some choice of maximum interpolation error $\beta$. All "bad points" $\tilde{\lambda}$ for which this is not the case get collected and added to the training set $\mathcal{T}_N$, thus creating a new training set on which the algorithm is repeated, leading to a new interpolant $I_n[h](\tilde{\lambda}; f)$. The validation step is repeated until no more bad points are found, leading to the final interpolant $I_{N_f}[h](\tilde{\lambda}; f)$.

Recall that the aim is to speed up the calculation of the likelihood $L = p(d | H_i, \tilde{\lambda}, I)$, the logarithm of which takes the form

$$\log L = \frac{1}{2} [2 \langle d | h(\tilde{\lambda}) \rangle - \langle h(\tilde{\lambda}) | h(\tilde{\lambda}) \rangle - \langle d | d \rangle].
$$

(19)

First, consider the term $\langle d | h \rangle$. Substituting for $h(\tilde{\lambda}; f)$ the empirical interpolant $I_{N_f}[h](\tilde{\lambda}; f)$ and discretizing the integral in the definition of the inner product, one gets
\[ \langle d| h(\tilde{\lambda}) \rangle = 4\Delta f \Re \sum_{i=1}^{L} \frac{d^* (f_i) h(\tilde{\lambda}; f_i)}{S_n(f_i)} \]

\[ \approx 4\Delta f \Re \sum_{i=1}^{L} \sum_{k=1}^{N_L} B_k^i (f_i) h(\tilde{\lambda}; F_k) \frac{d^* (f_i)}{S_n(f_i)} \]

\[ = 4\Delta f \Re \sum_{k=1}^{N_L} \left[ \sum_{i=1}^{L} B_k^i (f_i) \frac{d^* (f_i)}{S_n(f_i)} \right] h(\tilde{\lambda}; F_k) \]

\[ = \sum_{k=1}^{N_L} w_k h(\tilde{\lambda}; F_k), \quad (20) \]

where

\[ w_k \equiv 4\Delta f \Re \sum_{i=1}^{L} B_k^i (f_i) \frac{d^* (f_i)}{S_n(f_i)}. \quad (21) \]

An important point is now that, typically, \( N_L \ll L \), and the likelihood calculations (20), which during the nested sampling process need to be performed “on the fly” for many different parameter values, now only involve the evaluation of \( N_L \) expressions \( h(\tilde{\lambda}; F_k) \), rather than the \( L \) evaluations of \( h(\tilde{\lambda}; f) \) that were required originally. While it is true that the calculation of the ROQ weights \( w_k \) still involves a sum over \( L \) terms, they only need to be evaluated once for every detection. This means that, with the ROQ, this part of the likelihood calculation will be sped up by a factor \( L/N_L \).

Finally, in Eq. (19), there is also the term \( \langle h(\tilde{\lambda})| h(\tilde{\lambda}) \rangle \), which can be approximated by an expression of the form

\[ \langle h(\tilde{\lambda})| h(\tilde{\lambda}) \rangle = \sum_{j=1}^{Q} w_j^Q |h(\tilde{\lambda}; F_j^Q)|^2, \quad (22) \]

where

\[ w_j^Q \approx 4\Delta f \Re \sum_{i=1}^{L} B_j^Q (f_i) \frac{d^* (f_i)}{S_n(f_i)}, \quad (23) \]

for some \( B_j^Q \), and typically \( N_Q \ll L \). The \( B_j^Q \) are obtained through a similar procedure as in the linear case. Here too, the weights \( w_j \) will only have to be calculated once per detection. Note that \( L \) and \( Q \) in the superscripts refer to the linear and quadratic parts of the likelihood respectively.

The above only gives an overview of the rationale behind ROQs. In practice, one needs to write the waveform \( h(\tilde{\lambda}; f) \) in terms of the + and \( \times \) polarizations as \( F_+ h_+(\tilde{\lambda}; f) + F_\times h_\times(\tilde{\lambda}; f) \), with \( F_+ \) and \( F_\times \) as the beam pattern functions. However, it turns out that a single set of functions \( \{ B_j^Q \} \) is sufficient to represent \( h_+ \) and \( h_\times \), and a single set \( \{ B_j^Q \} \) is sufficient to represent the products \( |h_+|^2 \), \( |h_\times|^2 \), and \( \Re h_+^* h_\times \) [57,58].

### B. A ROQ for IMRPhenomPv2 with parametrized deformations

A ROQ for IMRPhenomPv2 in the GR case was already constructed in [58]. Here, we want to build a series of ROQs for IMRPhenomPv2, each including a single testing parameter \( \delta \hat{p}_t \). As a starting point, we use the final reduced basis for the GR waveform \( T_N \) (where \( N \) can be either the \( N_L \) or the \( N_Q \) of the linear and quadratic bases, respectively), and for each basis element we introduce \( N_{\delta \hat{p}_t} = 500 \) samples placed uniformly in the \( \delta \hat{p}_t \) direction (see Fig. 4). The ranges for the various \( \delta \hat{p}_t \) are chosen such that they accommodate the widths of posterior density functions of the LIGO-Virgo events that were recorded so far (with the exception of \( \delta \hat{a}_2 \), \( \delta \hat{a}_3 \), \( \delta \hat{a}_4 \), which are essentially unmeasurable for low-mass events)

\[
\delta \hat{p}_0 \in [-2, 2], \quad \delta \hat{p}_1 \in [-5, 5], \quad \delta \hat{p}_2 \in [-10, 10],
\delta \hat{p}_3 \in [-10, 10], \quad \delta \hat{p}_4 \in [-10, 10], \quad \delta \hat{p}_5 \in [-10, 10],
\delta \hat{p}_6 \in [-10, 10], \quad \delta \hat{p}_7 \in [-20, 20], \quad \delta \hat{p}_8 \in [-20, 20],
\delta \hat{a}_2 \in [-5, 5], \quad \delta \hat{a}_3 \in [-5, 5], \quad \delta \hat{a}_4 \in [-5, 5].
\]

### TABLE I. The different chirp mass bins for which ROQs were built, with the ranges of waveform durations in the GR case as well as sampling in frequency.

| Bin | \( M_c (M_\odot) \) | GR waveform duration (sec) | \( \Delta f \) (Hz) |
|-----|------------------|--------------------------|-----------------|
| A   | [12.3, 45]       | [0.4, 4]                 | 1/4             |
| B   | [7.9, 14.8]      | [3, 8]                   | 1/8             |
| C   | [5.2, 9.5]       | [6, 16]                  | 1/16            |
| D   | [3.4, 6.2]       | [12, 32]                 | 1/32            |
| E   | [2.2, 4.2]       | [23.8, 64]               | 1/64            |
The resulting set $T_{N \times N_{\hat{p}_i}}$ then becomes a training set for the construction of a new ROQ, as outlined in the previous subsection. As for the ROQ of the GR waveform, this is done independently for waveforms in different overlapping chirp mass bins, so as to obtain better likelihood calculation speedups than when all chirp masses would be lumped together. The chirp mass ranges roughly corresponding to different ranges for the length of the waveform in the time domain. We note that away from the GR case there is no clear mapping from chirp mass to waveform length, as the latter is also partially determined by the values of the $\hat{p}_i$. Even so, in each bin we a priori set $\Delta f = 1/T_{\text{max}}$, where $T_{\text{max}}$ is the longest GR waveform in the bin; though waveforms can become longer when $\hat{p}_i \neq 0$, in the end what counts is that all interpolation errors are below the given threshold. To reduce the burden

FIG. 5. Distributions of interpolation errors at different validation steps for some of the testing parameters and chirp mass bins. The left column is for the linear bases, the right column for the quadratic ones. (First row) $\hat{\delta} \phi_3$ for chirp mass bin A in Table I; (second row) $\hat{\delta} \beta_2$ for bin C; (third row) $\hat{\delta} \alpha_2$ for bin C. In some cases no bad points are found, so that the basis does not need to be enlarged and no further validation steps are needed.
on computer memory required, we perform multibanding as explained in [58]; an adaptive frequency resolution $\Delta f(f)$ is applied such that waveforms are sampled less densely at higher frequencies, where there is less power per frequency bin due to the faster frequency sweep (see Fig. 1). However, once a basis has been obtained, we up sample by direct evaluation of the waveform model.

The various bins are shown in Table I. Note that no ROQs were made for systems with $M_c > 45 M_\odot$, since for such binaries the signal will be short enough that parameter estimation will vary, but tend to be the same as the theoretical ones within a factor of 2 or less.

Table II shows the speedups in likelihood calculations due to the ROQs, for different testing parameters and the chirp mass bins of Table I. Note how these are larger for longer signals, where they are the most needed. Speedups in practical parameter estimation will vary, but tend to be the same as the theoretical ones within a factor of 2 or less.

| $\delta \hat{p}_i$ | A | B | C | D | E |
|------------------|---|---|---|---|---|
| $\delta \hat{\rho}_0$ | 4.3 | 7.6 | 28.3 | 38.7 | 47.1 |
| $\delta \hat{\rho}_1$ | 3.0 | 4.6 | 8.6 | 11.7 | 27.1 |
| $\delta \hat{\rho}_2$ | 4.2 | 6.8 | 24.8 | 42.4 | 56.5 |
| $\delta \hat{\rho}_3$ | 4.0 | 6.1 | 20.8 | 36.9 | 55.8 |
| $\delta \hat{\rho}_4$ | 3.9 | 10.1 | 40.3 | 76.5 | 111.2 |
| $\delta \hat{\rho}_{5i}$ | 4.1 | 7.6 | 29.9 | 62.9 | 97.9 |
| $\delta \hat{\tau}_0$ | 3.7 | 9.8 | 39.0 | 76.0 | 114.0 |
| $\delta \hat{\tau}_1$ | 3.8 | 10.1 | 42.1 | 78.1 | 117.1 |
| $\delta \hat{x}_2$ | 3.7 | 9.1 | 39.5 | 74.7 | 112.6 |
| $\delta \hat{x}_3$ | 3.0 | 8.7 | 34.9 | 78.0 | 117.5 |
| $\delta \hat{x}_4$ | 3.5 | 6.8 | 28.5 | 69.8 | 111.2 |
| $\delta \hat{x}_5$ | 2.8 | 9.2 | 39.4 | 88.2 | 124.6 |
| $\delta \hat{x}_6$ | 2.9 | 10.8 | 44.8 | 87.5 | 128.3 |
| $\delta \hat{x}_7$ | 2.8 | 10.4 | 43.3 | 88.1 | 131.6 |

We now perform some checks of the correctness of the data analysis pipeline and its robustness against waveform systematics and instrumental noise. We do this in two ways. One is to construct so-called $p$-$p$ plots, which quantify the statistical inconsistencies of the posterior density distributions. Another consists of analyzing a numerical waveform injected in many different stretches of real detector noise, as a check that the pipeline behaves as it should under the combined effects of the injected waveform being different from the template waveform model and the presence of instrumental glitches in the detector output.

### IV. ROBUSTNESS OF THE TESTS

A requirement for a parameter estimation algorithm is that it is capable of measuring parameters in a statistically reliable way. Detector noise can cause offsets in posterior density functions, but given a large number of signals, it should be the case that the correct parameter value is recovered with a confidence $p$ in a fraction $p$ of the cases. Specifically, assuming GR is correct, for any of the parametrized tests, it should be the case that the value $\delta \hat{p}_i = 0$ lies in a confidence interval of width $p$ for a fraction $p$ of the cases. We check this by adding 100 simulated GR signals (injections) to synthetic, stationary, Gaussian noise, with the predicted power spectral density at design sensitivity for the two Advanced LIGO detectors [73]. The signals have randomly chosen sky positions and orientations and are placed uniformly in comoving volume with $D_L \in [250, 750]$ Mpc, with component masses $m_1$, $m_2 \in [6, 40] M_\odot$, and arbitrarily oriented spins with magnitudes $|S_1|, |S_2| \in [0, 0.9]$. Injections are analyzed with the ROQs whose chirp mass bins they fall into; in reality, one would look at the chirp mass measured with GR templates.

### A. Reliable measurement of testing parameters

- **Testing parameter $\delta \hat{\rho}_i$, typically increases the sizes of the final bases in the different chirp mass bins by only a factor of a few, though with some exceptions; the largest increase happens** to be for $\delta \hat{\rho}_1$ and $M \in [3, 6, 2] M_\odot$, where the linear basis size went from 524 to 5264.

- **Table II** shows the speedups in likelihood calculations—defined as $[(f_{\text{max}} - f_{\text{min}})T + 1]/(N_L + N_O)$—that are achievable with the ROQs. The speedup is greatest for long signals where analyses are the most involved. These are the theoretical speedups; the actual speedups in practical parameter estimation will vary, but tend to be the same as the theoretical ones within a factor of 2 or less.

- Finally, the ROQs were interfaced with the aforementioned LALInference framework. Figure 6 compares some parameter estimation results obtained with and without the ROQ on the same simulated signal. We see that the results are consistent, with posterior density functions not differing by more than what is expected given uncertainties in the sampling process [72]. The robustness of the infrastructure will be tested in more detail in the next section.
p-p plots for a few of the testing parameters are shown in Fig. 7. As an indicator of consistency of the results with absence of bias in the measurements, one can calculate the Kolmogorov-Smirnov (K-S) statistic, which is defined as the maximum (in absolute value) of the difference between distributions; in our case, the latter are simply the p-p distribution on the one hand and the diagonal on the other. We find K-S values of 0.04, 0.09, and 0.04 for \( \hat{\delta} \phi_3 \), \( \hat{\delta} \beta_2 \), and \( \hat{\delta} \alpha_2 \), respectively. We conclude that the analyses work as expected.

B. Numerical relativity injections in real detector noise

Finally, we investigate the response of the parametrized tests to a numerical relativity waveform injected in detector noise that contains instrumental glitches. In particular, we use real data from the S6 data set, but “recolored” to the early Advanced LIGO noise curve from [74]; this procedure changes the average power spectral density but retains (and in fact enhances) any instrumental nonstationarities that were present in the original data. Since instrumental glitches will have a larger effect for short-duration signals, we focus on GW150914. We consider a numerical relativity waveform from the SXS catalog, whose mass ratio and spins are close to the measured means for GR150914; specifically, we pick SXS:BBH:0307 [75]. The intrinsic parameters were \((m_1, m_2) = (40.83, 33.26)M_\odot\), and \(S_1 = (0.092, 0.038, 0.326)\), \(S_2 = (0.215, 0.301, -0.558)\) at \(f_{\text{red}} = 20\) Hz. This same waveform is then injected in 21
different stretches of noise [76, 77] and the parametrized tests are performed. In choosing these stretches, care was taken to pick ones that did not exhibit egregiously large glitches (which can be done by visual inspection of time-frequency spectrograms), since the presence of a sufficiently sizeable departure from Gaussianity of the noise may preclude an event being detected in the first place. The strategy is similar to what was followed in [78] (see their Sec. III E), where the effect of possible nonstationarities on parameter estimation—in the GR case—was also assessed by injecting a particular numerical relativity waveform in different stretches of real detector noise.

As a diagnostic, we define the “GR quantile” as the cumulative probability of a given $\delta \hat{p}_i$ being nonpositive

$$Q_i = \int_{-\infty}^{0} p(\delta \hat{p}_i|H_i, d, I) d\delta \hat{p}_i.$$  \hfill (25)

If the GR quantile is close to zero, then the posterior $p(\delta \hat{p}_i|H_i, d, I)$ exhibits a significant offset towards positive $\delta \hat{p}_i$; if it is close to one, then there is a large offset towards negative values. Given many measurements on the same signal in different noise realizations, we expect the $Q_i$ to be distributed uniformly on the interval $[0, 1]$.

In Fig. 8, we first of all show the 90% credible intervals for the PN testing parameters $\{\delta \hat{p}_0, \ldots, \delta \hat{p}_7\}$ and $\{\delta \hat{p}_5, \delta \hat{p}_6\}$. Different stretches of noise [76, 77] and the parametrized tests are performed. In choosing these stretches, care was taken to pick ones that did not exhibit egregiously large glitches (which can be done by visual inspection of time-frequency spectrograms), since the presence of a sufficiently sizeable departure from Gaussianity of the noise may preclude an event being detected in the first place. The strategy is similar to what was followed in [78] (see their Sec. III E), where the effect of possible nonstationarities on parameter estimation—in the GR case—was also assessed by injecting a particular numerical relativity waveform in different stretches of real detector noise.

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for the 21 stretches of data. We note how the deviations in the PN parameters tend to alternate in sign, due to the fact that there is some correlation between them, and that the $p_i$ themselves have alternating signs. Next, in Fig. 9, we show the distribution of the $Q_i$, which despite the small sample size is indeed suggestive of uniformity on $[0,1]$.

Needless to say, a full investigation for systems like GW150914 would require performing the parametrized tests for a much larger sample of data stretches than the 21 used here and it would be of interest to repeat the study for other choices of masses and spins; due to computational restrictions, this was not practicable. Nevertheless, the outcome is indicative of the expected behavior.

V. MEASUREMENT SENSITIVITIES

Next, we want to assess the power of our parametrized tests in constraining GR violations and their sensitivity to selected GR violations, by adding simulated signals to stationary Gaussian detector noise with the power spectral density of the Advanced LIGO detectors at design sensitivity [73] and performing parameter estimation as in the previous section.

As far as GR violations are concerned, ideally one would like to do this using specific alternative theories of gravity. However, in most cases, the effects of particular theories have only been calculated for the inspiral and then only to leading PN order [13,36,51,79]; to our knowledge, full inspiral-merger-ringdown waveform models with reasonable inclusion of all relevant physical effects so far only exist for GR itself. Hence, we confine ourselves to injections that have a deviation $\delta p_i$ in a particular coefficient $p_i$, or in several of the $p_i$ at the same time, starting from some PN order. However, in the template waveforms used for the measurements, we still only vary a single one of the $\delta p_i$ at a time. As we shall see, if the injections have deviations in multiple coefficients, then single-parameter tests will still pick this up. In fact, even parameters that are not associated with the deviations in the signal must show deviations. Such effects had already been observed in [52,53,55], and should not come as a surprise: template waveform models will use whatever additional freedom they have to accommodate anomalies in the signals. At the same time, only varying one testing parameter leads to a higher measurement accuracy than for multiple parameters being varied at the same time. A drawback is that posterior densities for testing parameters can not be straightforwardly mapped to statements about whatever additional charges, coupling constants, or energy scales may be present in some particular alternative theories. For this to be possible, accurate and complete inspiral-merger-ringdown waveforms for alternative theories would be required, but these are not currently available. However, the purpose of the parametrized tests is not to place bounds on parameters characterizing other theories, but rather to test the theory of general relativity itself, with as high an accuracy as possible.

A. Bounding GR violations

First, we illustrate the ability of the parametrized tests in putting bounds on GR violations, which will get increasingly sharper as information from multiple events is combined. The posterior density functions for each of the $\delta p_i$, obtained from the simulated signals in Sec. IVA lead to combined posterior densities according to the prescription of Eq. (9). As shown in Fig. 10, after a few tens of detections, these will be sharply peaked near the value of zero. In these examples, after 50 (100) detections, the $1 - \sigma$ accuracies on $\delta \hat{p}_3$, $\delta \hat{p}_1$, and $\delta \hat{a}_2$ are, respectively, 0.013 (0.008), 0.020 (0.013), and 0.054 (0.032).

B. Simulated signals with deviations in particular coefficients

We now consider injections that have a deviation in a particular coefficient $p_i$. The GR parameters are picked to be the means of the posterior density distributions for GW150914 [2]. We focus on this type of source so as to have some amount of sensitivity to each of the inspiral, intermediate, and merger-ringdown regimes. The injections are done in stationary Gaussian noise with the predicted

![FIG. 10. Sharper constraints on deviation from GR can be obtained by combining posterior density functions for the $\delta \hat{p}_i$ from all available detections. This is illustrated for $\delta \hat{p}_3$ (left), $\delta \hat{p}_2$ (middle), and $\delta \hat{a}_2$ (right). The black curve shows the median of the joint distribution, and the darker and lighter shadings show the 68% and 95% confidence intervals, respectively.](044033-13)
A few things can be noted:

1. In each case, the posterior density for the testing parameter where the deviation in the signal resides has no support at the GR value of zero, but the support does contain the injected value.
2. The posterior densities of all of the other PN testing parameters, with the exception of $\delta p_1$, show strong offsets away from zero.
3. On the other hand, the intermediate-regime and merger-ringdown testing parameters show much less of a response to a deviation in a PN parameter.
4. The deviations in the PN parameters tend to alternate in sign. This reflects the fact that there is some amount of correlation between these parameters and that the $\varphi_i$ themselves have alternating signs.

The posteriors in Figs. 12 and 13, where either an intermediate-regime parameter or a merger-ringdown parameter in the signal has a deviation, show analogous behavior: for the parameter where the deviation resides, posteriors have no support at zero, but this is also the case for at least one other parameter, usually one in the same regime.

C. Simulated signals with deviations in multiple coefficients

Next, we consider injections in which all the $\delta p_i$ are nonzero starting from some PN order. Two scenarios are considered:

1. All testing parameters starting from 1.5 PN have the same fractional shifts $\delta p_i = 0.5$. This includes the sets $\delta p_{3,4,5,6,7,8}$, $\delta p_{2,3}$, and $\delta p_{2,3,4}$.
2. All testing parameters starting from 1.5 PN have shifts whose sign alternates from one parameter to the next, according to the way they are correlated: $\delta p_{3,5,6,7} = -0.4$ and $\delta p_{4,6} = +0.4$. For the intermediate-regime and merger-ringdown parameters, we choose $\delta p_2 = \delta p_3 = -0.4$ and $\delta \alpha_4 = \delta \alpha_3 = \delta \alpha_2 = 0.4$.

The results are shown in Fig. 14, and can be summarized as follows:

1. Again strong deviations are picked up even by testing parameters that are not associated with the violations in the signal; both for the same- and alternating-sign violations, all of the testing parameters return a posterior density function whose support does not contain the GR value of zero.
FIG. 12. (Top) Posterior densities for testing parameters for an injection with $\delta \hat{\beta}_2 = +0.7$ (orange) and $\delta \hat{\beta}_2 = -0.7$ (blue). (Bottom) Posteriors for an injection with $\delta \hat{\beta}_3 = +0.8$ (orange) and $\delta \hat{\beta}_3 = -0.8$ (blue). In each case, the GR violation is also picked up by the other $\delta \hat{\beta}_i$.

FIG. 13. (Top) Posterior densities for testing parameters for an injection with $\delta \hat{\alpha}_2 = +1.3$ (orange) and $\delta \hat{\alpha}_2 = -1.3$ (blue). (Bottom) Posteriors for an injection with $\delta \hat{\alpha}_4 = +1.6$ (orange) and $\delta \hat{\alpha}_4 = -1.6$ (blue). Here too, in each case, the other $\delta \hat{\alpha}_i$ also pick up the GR violation.
(ii) Even in the case where the signs of all the deviations are the same, we see alternation in the offsets of the posterior densities for PN parameters, following the way they are correlated.

(iii) For PN parameters from 1.5 PN onwards, the measured GR violation is larger than the injected deviation; individual parameters respond to the collective change in the waveform induced by the shifts in all of the testing parameters together. Hence, measuring the $\delta \hat{p}_i$ one by one can enable the discovery of GR violations also when the signal has multiple $p_i$ that deviate from their GR values.

VI. SUMMARY AND CONCLUSIONS

In [4,5,12], the detected binary black hole signals were analyzed using template waveforms that allow for parametrized deviations from GR, so as to test the strong-field dynamics of the theory. In this work, we have introduced reduced-order quadratures that speed up likelihood calculations by factors of a few to more than a hundred, which will significantly ease the computational burden in applying the method to future events. Our chosen waveform model is IMRPhenomPv2, though we note that the method used in this paper can in principle also be applied to reduced-order models for other frequency domain waveforms with parametrized deviations added, such as the ones in [61,62]. We also established the method’s robustness through $p$-$p$ plots for simulated signals in synthetic Gaussian noise and by examining the results for a numerical relativity injection in different stretches of real data from the S6 data set, recolored to the Advanced LIGO final design sensitivity. Finally, the sensitivity of the method was evaluated using both GR injections and injections with GR violations in various parameters.

A range of alternative theories of gravity have been considered, which are often characterized by additional charges or coupling constants. The tests presented here do not easily map to statements about such parameters; putting constraints on particular alternative theories would require full inspiral-merger-ringdown waveforms of similar quality as the ones we have for GR. The regular observation of binary black hole coalescences will be an incentive for theorists to develop such models. However, the main aim of the parametrized tests is to perform stringent tests of GR itself, and as we have demonstrated, our method provides a reliable and accurate way of doing this.

Recently a binary neutron star merger was also discovered [44]. Here too the parametrized tests can be applied, although care should be taken so that the effects of the neutron stars’ tidal deformation are not confused with a violation of GR. This can be done by analyzing the signal up to frequencies of only a few hundred hertz so that tidal effects can be neglected [52–54] or by including tidal deformabilities in the signal. The latter approach has the advantage that the entire signal can be used, but there will also be some loss of sensitivity due to the increased dimensionality of parameter space; which approach will be the most efficient is yet to be determined. Especially for these kinds of events, which involve longer signals than for binary black holes, it would be beneficial to construct reduced-order quadratures; this too is left for future work.

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