Frequency-domain reduced-order model of aligned-spin effective-one-body waveforms with higher-order modes

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We present a frequency-domain reduced-order model (ROM) for the aligned-spin effective-onebody model for binary black holes (BBHs) SEOBNRv4HM that includes the spherical-harmonics modes $(\ell, |m|) = (2, 1), (3, 3), (4, 4), (5, 5)$ beyond the dominant $(\ell, |m|) = (2, 2)$ mode. These higher modes are crucial to accurately represent the waveform emitted from asymmetric BBHs. We discuss a decomposition of the waveform, extending other methods in the literature, that allows us to accurately and efficiently capture the morphology of higher mode waveforms. We show that the ROM is very accurate with median (maximum) values of the unfaithfulness against SEOBNRv4HM lower than 0.001%(0.03%) for total masses in $[2.8, 100]M_{\odot}$. For a total mass of $M = 300 M_{\odot}$, the median (maximum) value of the unfaithfulness increases up to 0.004%(0.17%). This is still at least an order of magnitude lower than the estimated accuracy of SEOBNRv4HM compared to numerical relativity simulations. The ROM is 2 orders of magnitude faster in generating a waveform compared to SEOBNRv4HM. Data analysis applications typically require $\mathcal{O}(10^6-10^8)$ waveform evaluations for which SEOBNRv4HM is in general too slow. The ROM is therefore crucial to allow the SEOBNRv4HM waveform to be used in searches and Bayesian parameter inference. We present a targeted parameter estimation study that shows the improvements in measuring binary parameters when using waveforms that includes higher modes and compare against three other waveform models.

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I. INTRODUCTION

In the past five years, gravitational wave (GW) observations [1–3] have opened up a new window to the Universe. In the first two observing runs of the advanced LIGO [4] and Virgo [5] detectors, ten confident detections of binary black holes (BBHs) and one binary neutron star were made [3], and tens of detection candidates [6] have been identified so far in the third observing run of this network, among them another confident detection of a binary neutron star system [7]. Both the detection and inference of binary parameters for these compact binaries rely heavily on our knowledge of the gravitational waveform emitted in these coalescences as encapsulated in

Published by the American Physical Society under the terms of the Creative Commons Attribution 4.0 International license. Further distribution of this work must maintain attribution to the author(s) and the published article's title, journal citation, and DOI. Open access publication funded by the Max Planck Society. parametrized models of GWs. The construction of stochastic template banks and the Bayesian inference of binary parameters routinely require tens to hundreds of millions of waveform evaluations [8–11]. At the same time, the phasing of the GWs needs to be tracked to an accuracy better than a fraction of a wave cycle to avoid missing events or mismeasuring binary parameters. Therefore, waveform models need to be fast and accurate to extract the binary properties imprinted in the emitted GWs.

Inspiral-merger-ringdown models of GWs from coalescing black hole (BH) binaries have traditionally been constructed in the effective-one-body (EOB) [12–27] or phenomenological [28–37] approaches, and, more recently, models for numerical relativity (NR) or EOB waveforms constructed with surrogate modeling techniques have come to prominence [23,38–46]. EOB models incorporate physical descriptions of the inspiral, merger, and ringdown parts of BBH coalescences. Post-Newtonian (PN) solutions for the inspiral are re-summed and connected with an analytic description of the merger waveform which is tuned to data from NR simulations [47–55]. EOB models are posed as an

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initial value problem for a complicated system of ordinary differential equations (ODEs) describing the approximated dynamics of a compact binary. The emitted GWs are then computed from this orbital dynamics. EOB models have provided accurate and generic descriptions of the GWs for the signals observed so far by LIGO and Virgo detectors. However, observations with third-generation detector networks may require much more accurate waveform models [56]. The integration of the ODEs requires small time steps to obtain an accurate solution. Especially for the long waveforms produced by low mass compact binaries this can take on the order of hours, and thus be too slow for practical data analysis applications.²

Surrogate or reduced-order modeling techniques [38–43, 45,58,59] provide established methods for accelerating slow waveforms while retaining very high accuracy. These techniques have been successfully applied to EOB [23,38–40,45] and NR [41–44,46] waveforms. They work by decomposing waveforms from a training set in orthonormal bases on sparse grids in time or frequency and interpolating or fitting the resulting waveform data pieces over the binary parameter space. The result is a smooth, accurate (as tested against an independent validation set), and fast to evaluate (compared to the original waveform data) GW model. These surrogate models have proven invaluable for GW data analysis.

In this paper, we present a reduced-order model (ROM) for GWs from coalescing binaries with spins aligned with the orbital angular momentum which include the most important higher harmonics of the waveform in addition to the dominant $(\ell.m) = (2, \pm 2)$ spherical harmonic mode, as described by the SEOBNRv4HM model [25]. Higher harmonics in the expansion of the gravitational waveform become important for asymmetric and massive compact binaries [60–65]. The model we construct here, SEOBNRv4HM_ROM, includes the $(\ell, |m|) = (2, 1), (3, 3), (4, 4), (5, 5)$ modes. We show that SEOBNRv4HM_ROM has a mismatch less than $\mathcal{O}(0.1\%)$ with SEOBNRv4HM and that it accelerates waveform evaluation by a factor 100–200.

While SEOBNRv4HM and SEOBNRv4HM_ROM include the contribution of higher harmonics in the waveform, they lack a description of spin precession and eccentricity. In the EOB waveform family, the effect of spin precession has been taken into account in Refs. [20,24] and only recently in Ref. [27] for the case of waveforms with higher harmonics. A surrogate for the latter waveform model is currently under development (see Ref. [66]). For the phenomenological and the numerical relativity surrogate families, similar models including the effect of

precession and higher harmonics have been described in Refs. [36,43,44,46,67]. Waveforms emitted from binary systems in an eccentric orbit have not been studied extensively. So far, only a few inspiral-merger-ringdown eccentric waveform models have been constructed for nonspinning [68,69] and aligned-spin binaries [70,71].

This paper is organized as follows. In Sec. II, we give a brief description of the time-domain SEOBNRv4HM model. In Sec. III, we discuss various techniques we use to build the ROM, notably waveform conditioning in Sec. III A. We continue with a summary of the basis construction and decomposition in Sec. III B and tensor product interpolation in Sec. III C. Domain decomposition in frequency and in parameter space is discussed in Secs. III D and III F, respectively. We summarize how we connect the ROM with PN solutions at low frequency in Sec. III E. We present results in Sec. IV, where we demonstrate the accuracy of the ROM in Sec. IVA and its computational efficiency in Sec. IV B. We showcase a parameter estimation application in Sec. IV C. Finally, we conclude in Sec. V.

II. SEOBNRv4HM MODEL

The gravitational wave signal emitted by a coalescing binary black hole is usually divided into three different regimes: inspiral, merger, and ringdown. During the inspiral, the two black holes move at a relative speed v that is small compared to the speed of light c; therefore, the solution to the two-body problem can be found using a perturbative expansion in the small parameter v/c, the socalled PN expansion [72]. At some point, during the evolution of the binary system, the parameter v/c ceases to be small, and the PN expansion is not valid anymore. This happens roughly at the innermost stable circular orbit and demarcates the end of the inspiral and the beginning of the merger regime. The signal in this regime can only be computed using NR simulations that solve Einstein's equations for a BBH system, fully numerically. Finally, in the ringdown regime, the perturbed black hole formed after the merger of the binary emits gravitational waves at frequencies that can be computed within the black hole perturbation theory formalism [73].

The EOB formalism, first introduced by Buonanno and Damour in Refs. [12,13], provides a natural framework to combine these three regimes and produce a complete waveform with inspiral, merger, and ringdown. Within the EOB formalism, the PN conservative dynamics of a BBH system during the inspiral is resummed in terms of the dynamics of a test particle with an effective mass and spin around a deformed Kerr metric. This improved conservative dynamics is combined with a resummed energy flux [74–76] to produce an inspiral waveform that is close to NR solutions. To improve the agreement with NR waveforms, the EOB conservative dynamics is also *calibrated* using information from NR simulations [18,19]. In the EOB waveform, the merger and ringdown part is

¹The full dynamics of a binary system is obtained by solving Einstein's equations which are partial differential equations.

²A faster method has been proposed, restricted to systems with spins aligned with the orbital angular momentum [57].

built using a phenomenological fit produced using informations from NR waveforms and black hole perturbation theory [23,77]. NR-tuned versions of EOB models are usually referred to as EOBNR.

In this paper, we focus on the SEOBNRv4HM [25] model, an extension of SEOBNRv4 [23] that includes the modes $(\ell, |m|) = (2, 1), (3, 3), (4, 4), (5, 5)$ in the waveform in addition to the $(\ell, |m|) = (2, 2)$ mode already present in SEOBNRv4. This model assumes spins aligned or antialigned with the direction perpendicular to the orbital plane \hat{L}_N , and we define the dimensionless spin parameter for BH i as $\chi_i = \vec{S}_i \cdot \hat{L}_N/m_i^2$, where \vec{S}_i are the BHs' spins and m_i are their masses. SEOBNRv4HM has been validated against several NR waveforms in the mass ratio—aligned spin parameter space in the region $q \equiv m_1/m_2 \in [1, 10], \chi_{1,2} \in [-1, 1]$ yielding typical mismatches of $\mathcal{O}(\leq 1\%)$ for total masses in the range $[20, 200]M_{\odot}$.

III. TECHNIQUES FOR BUILDING THE ROM

In this section, we describe the construction of our ROM, from the preparation of the waveforms to the reduced basis and interpolation techniques. We use techniques developed for previous ROM models [23,39,40], which we generalized to the higher-harmonics case.

We start with a general discussion of how to prepare and decompose waveform data for higher mode waveforms in Sec. III A. In particular, we discuss time-domain conditioning in Sec. III A 1, stationary phase approximation in Sec. III A 2, the orbital carrier phase in Sec. III A 3, the introduction of coorbital modes in Sec. III A 4, and scaling of frequencies in Sec. III A 5. We summarize basis construction in Sec. III B and tensor product interpolation in Sec. III C. We also explain how we decompose the model in both frequency range patches (see Sec. III D) and parameter space patches (see Sec. III F). Hybridization with inspiral waveforms is discussed in Sec. III E.

A. Preparation and decomposition of waveform data

The waveform polarizations h_+ , h_\times are decomposed in spin-weighted spherical harmonics as

$$h_{+} - ih_{\times} = \sum_{\ell \ge 2} \sum_{m=-\ell}^{\ell} {}_{-2}Y_{\ell,m} h_{\ell m}.$$
 (3.1)

The $h_{\ell m}$ are called the harmonics or simply the modes of the gravitational wave, with h_{22} and $h_{2,-2}$ the dominant harmonics corresponding to quadrupolar radiation. These modes $h_{\ell m}$ are affected by convention choices: first, by the choice of polarization vectors defining h_+ , h_\times and, second, by the definition chosen for the source frame in which the waveform is described. For nonprecessing systems, the z axis of the source frame is taken to be the normal to the

orbital plane, with a residual freedom in choosing the origin of phase. One can take two points of view for the definition of phase: either fix the definition of the source frame (for instance, imposing that the initial separation vector is along x) and call "phase" the azimuthal angle of the observer in the source frame or fix the direction to the observer [for instance, in the plane (x, z)] and call "phase" the binary's orbital phase at a given time. We can also consider the definition of the origin of time as part of the source frame definition.

During the inspiral, the individual harmonics obey a simple overall scaling with the orbital phase as

$$h_{\ell m} \propto \exp\left[-im\phi_{\rm orb}\right],$$
 (3.2)

but this scaling does not apply postmerger where the modes are driven by their respective ringdown frequencies.

There are several challenges regarding the conditioning of higher-harmonics waveforms for the purpose of reduced-order modeling. We recall that one relies on two kinds of interpolation here: one is the interpolation of waveform pieces along the tracking parameter, either time or frequency, used to compress data; the other is the interpolation across the parameter space (masses and spins) used in the internals of the ROM, either of waveform quantities directly (as in Refs. [38,41–46] in the empirical interpolation formalism) or, as in our case, of reduced basis projection coefficients [23,39,40]. Both these interpolations require smoothness, and discrete jumps can cause significant (and nonlocal) errors.

As a result, zero crossings in the subdominant harmonics $h_{\ell m}$ (as noticed in Refs. [25,78]) cause difficulties for the usual amplitude/phase representation: if the envelope of a mode crosses zero with a positively defined amplitude, the phase jumps by π , a discontinuity that will harm the interpolation performed when reconstructing the waveform. Among other advantages, this is alleviated by the procedure used in Refs. [42,46] of modeling the waveform in a coorbital frame where the dominant phasing of Eq. (3.2) has been scaled out, so that a more robust real/imaginary representation can be chosen instead; here, we will use the same kind of coorbital quantities, but built in the Fourier domain.

The natural 2π degeneracy in phase also requires care when interpolating across parameter space. Discrete 2π phase jumps leave the waveforms themselves invariant, but can break the interpolation in between waveforms. This issue is particularly relevant when dealing with numerical Fourier transforms of time-domain waveforms: when phase unwrapping the output of the discrete Fourier transform starting from f=0, numerical noise causes the 2π interval to be essentially random. In Refs. [23,39,40], a linear fit of the Fourier-domain phase was removed. Here, we will keep time and phase alignment information throughout the conditioning procedure, so instead, we

will impose a given 2π range for the phase at a reference point, corresponding to the time of alignment.

Other difficulties are caused by the relative alignment of the different harmonics. Dividing the phase of the dominant h_{22} mode by 2, whether in the time or frequency domain, comes with an ambiguity of π then propagated as $m\pi$ to the other modes. Such an ambiguity is not necessarily a problem if the phase alignment is done as a last step when generating the waveform (as is the case in the IMRPhenomHM model [34]); giving up the geometric interpretation of the source-frame definition, it is sufficient that a $[0, 2\pi]$ range in the phase input by the user corresponds to a $[0, 2\pi]$ range in geometric phase. It becomes a problem, however, when we need to interpolate across parameter space to build a ROM. In particular, when working from the Fourierdomain waveform alone, we do not have access to the orbital phase (as read from trajectories) to lift these kind of degeneracies. Here, we will make sure that the alignment is performed in the time domain before taking Fourier transforms, and we will further introduce an artificial carrier signal to have access to a proxy of the orbital phase in the Fourier domain.

We detail below our conditioning procedure, chosen to circumvent these issues.

1. Time-domain conditioning

In building this ROM, we will carry along time and phase alignment information all the way to the final Fourier-domain waveforms. This is in contrast to previous Fourier-domain waveform models (SEOBNRv4_ROM and IMRPhenoMD) where the time and phase are adjusted after generating the waveform, as will be described below.

Individual harmonics are decomposed as an amplitude³ and phase, following

$$h_{\ell m}(t) = a_{\ell m}(t) \exp\left[-i\phi_{\ell m}(t)\right], \tag{3.3}$$

with the scaling

$$\phi_{\ell m} = m\phi_{\rm orb} + \Delta\phi_{\ell m}. \tag{3.4}$$

In the early inspiral regime, for low frequencies, the phases $\Delta \phi_{\ell m}$ are approximately constant. We choose the same polarization convention as in Ref. [79], for which we have

$$\Delta\phi_{22} \to 0,$$

$$\Delta\phi_{21} \to \frac{\pi}{2},$$

$$\Delta\phi_{33} \to -\frac{\pi}{2},$$

$$\Delta\phi_{44} \to \pi,$$

$$\Delta\phi_{55} \to \frac{\pi}{2},$$
(3.5)

in the low-frequency limit. When getting closer to merger, deviations from Eq. (3.5) become more important. In the notations of the EOB factorized waveforms [16,17], these phase deviations come from the phases $e^{i\delta_{\ell m}}$ and tail factors $T_{\ell m}$ (see Eqs. (14) and (21) in Ref. [17]), and from nonquasicircular corrections close to merger (see Eq. (22) in Ref. [17]).

We choose the source frame convention for our model by imposing that its direction x is along the separation vector between the two black holes $n(t_{\rm align})$, with an arbitrary time of alignment in the late inspiral $t_{\rm align} = -1000M$ (with t=0 being defined as the amplitude peak of h_{22}). In practice, rather than using $n(t_{\rm align})$, we simply impose

$$\phi_{22}(t_{\text{align}}) = 0,$$
 (3.6)

and we use the orbital phase $\phi_{\rm orb}$ as read from the EOB dynamics to resolve the π ambiguity and impose $\phi_{\rm orb} \simeq 0$ rather than $\phi_{\rm orb} \simeq \pi$. These alignment properties will be reproduced, up to small numerical errors, by the reconstructed ROM waveforms.

2. Stationary phase approximation

As we will use it to guide our conditioning procedure, we recall here the stationary phase approximation (SPA) for waveforms with higher harmonics. First, we introduce the Fourier transform for a time-domain signal h as

$$\tilde{h}(f) = \int dt e^{2i\pi ft} h(t). \tag{3.7}$$

Note the sign difference in the exponential with respect to the more usual definition (used in particular in Ligo Algorithm Library (LAL) [80]). This choice is made for convenience, as it will ensure that Fourier-domain modes $\tilde{h}_{\ell m}$ with m>0 and m<0 have support at positive and negative frequencies, respectively. One can come back to the LAL Fourier convention by the simple map $f \leftrightarrow -f$, which for real signals $h(t) \in \mathbb{R}$ translates as $\tilde{h}_{\text{LAL}}(f) = \tilde{h}^*(f)$.

Let us first consider a generic signal with an amplitude and phase as $h(t) = a(t)e^{-i\phi(t)}$ and define $\omega \equiv \dot{\phi}$. The SPA is applicable under the assumptions $|\dot{a}/(a\omega)| \ll 1$, $|\dot{\omega}/\omega^2| \ll 1$ and $|(\dot{a}/a)^2/\dot{\omega}| \ll 1$, that are well verified in the inspiral. Defining a time-to-frequency correspondence t(f) implicitly by

³In general, it would be preferable to consider $a_{\ell m}$ as a slowly varying envelope rather than a positive amplitude, in particular allowing it to change sign, as we expect zero crossings of certain subdominant modes like h_{21} .

$$\omega(t(f)) = 2\pi f,\tag{3.8}$$

the Fourier transform of the signal is then $\tilde{h}_{\rm SPA}(f)=A_{\rm SPA}(f)e^{-i\Psi_{\rm SPA}(f)}$ with

$$A_{\text{SPA}}(f) = a(t(f))\sqrt{\frac{2\pi}{\dot{\omega}(t(f))}},$$
 (3.9a)

$$\Psi_{\text{SPA}}(f) = \phi(t(f)) - 2\pi f t(f) + \frac{\pi}{4}.$$
 (3.9b)

Applying this to the individual $h_{\ell m}$ modes (3.3), treating the $\Delta \phi_{\ell m}$ as constants, defining $\omega_{\ell m} = \dot{\phi}_{\ell m} \simeq m\omega_{\rm orb}$, each mode will have a separate time-to-frequency correspondence,

$$\omega_{\ell m}(t^{\ell m}(f)) = 2\pi f, \tag{3.10}$$

and the phase

$$\Psi_{\ell m}^{\mathrm{SPA}}(f) = m\phi_{\mathrm{orb}}(t^{\ell m}(f)) - 2\pi f t^{\ell m}(f) + \Delta\phi_{\ell m} + \frac{\pi}{4}. \tag{3.11}$$

This is the Fourier-domain equivalent to the time-domain relation (3.4). We note useful relations between different mode numbers. The various $t^{\ell m}(f)$ functions are related by

$$t^{\ell m}\left(\frac{mf}{2}\right) = t^{22}(f),\tag{3.12}$$

while the phases satisfy

$$0 = 2\Psi_{\ell m} \left(\frac{mf}{2}\right) - m\Psi_{22}(f) - 2\Delta\phi_{\ell m} + m\Delta\phi_{22} - (2-m)\frac{\pi}{4}. \tag{3.13}$$

This last relation holds regardless of the time and phase alignment of the waveform: as either a phase change $\delta\phi_{\ell m}=m\delta\phi$ or a time change $\delta\Psi_{\ell m}(f)=-2\pi f\delta t$ would leave $2\Psi_{\ell m}(mf/2)-m\Psi_{22}(f)$ invariant. It is sensitive, however, to the quantities $\Delta\phi_{\ell m}$ (that we treat here as constants in the early inspiral), which depend on the choice of polarization convention.

Finally, we recall that we can build a time-to-frequency correspondence directly from the Fourier-domain waveform as

$$t(f) \equiv -\frac{1}{2\pi} \frac{d\Psi}{df}.$$
 (3.14)

Note that this definition of time is, strictly speaking, only accurate in the inspiral phase, where the SPA applies and

the two definitions (3.14) and (3.8) coincide. However, it can be used as a proxy for time everywhere, as we can evaluate (3.14) for any frequency f.

3. Orbital carrier

To carry over information about the alignment of the respective mode from the time domain to the Fourier domain, we find it convenient to introduce a fictitious carrier signal k(t), which evolves with the orbital phase instead of twice the orbital phase as the h_{22} mode does,

$$k(t) \equiv a_{22}(t) \exp\left[-i\frac{\phi_{22}(t)}{2}\right].$$
 (3.15)

The choice made here of keeping the same amplitude as the h_{22} mode is quite arbitrary but will ensure that it decays in the ringdown, giving us a smooth Fourier transform for this carrier. Note that this construction is artificial, as the carrier does not correspond to any physical signal.

As mentioned before, the carrier half-phase $\phi_{22}/2$ comes with a π degeneracy. We can alleviate this by forcing the carrier phase to be within π of the orbital phase, as read from the EOB dynamics, at the time of alignment. This is, in fact, our main motivation for building this carrier in the time domain; it allows us to avoid the issues listed above, with all the conditioning being ultimately tied to the orbital phase, a quantity that is smooth across parameter space.

The Fourier transform of the carrier signal introduced in (3.15) is decomposed as an amplitude and phase as

$$\tilde{k}(f) = A_k(f) \exp\left[-i\Psi_k(f)\right], \tag{3.16}$$

where $A_k = |\tilde{k}|$ will be discarded in the rest of the conditioning. When the SPA applies, we have approximately

$$\Psi_k(f) \simeq \Psi_k^{\text{SPA}}(f) = \phi_{\text{orb}}(t^k(f)) - 2\pi f t^k(f) + \frac{\pi}{4}, \quad (3.17)$$

with $t^k(f)$ defined like in (3.8) as

$$\omega_{\text{orb}}(t^k(f)) = 2\pi f. \tag{3.18}$$

Since $\tilde{k}(f)$ is computed via a FFT, nothing forbids arbitrary jumps of 2π of the phases between different points in parameter space. We use the relation above to remove this 2π ambiguity in Ψ_k . At the frequency $f_{\rm align}$ such that $t(f_{\rm align}) = t_{\rm align}$, we impose

$$\left| \Psi_k(f_{\text{align}}) - \left(\phi_{\text{orb}}(t_{\text{align}}) - 2\pi f_{\text{align}} t_{\text{align}} + \frac{\pi}{4} \right) \right| < \pi.$$
(3.19)

⁴The converse is not true: since t(f) is not monotonic at high frequencies, building an unambiguous mapping f(t) is only possible in the inspiral.

In this way, Ψ_k is directly tied to $\phi_{\rm orb}$ that is smooth in parameter space in our time-domain conditioning procedure.

We will factor out the Fourier-domain phase of the carrier, build a ROM for the carrier separately, and then factor in the modeled carrier phase when reconstructing the waveform.

4. Fourier-domain coorbital modes and waveform building blocks

Next, we build coorbital modes by scaling out the Fourier-domain phase of the carrier following

$$\tilde{h}_{\ell m}^{c}(f) = \tilde{h}_{\ell m}(f) \exp\left[im\Psi_{k}(f/m)\right]. \tag{3.20}$$

These modes are built so as to factor out the main contribution to the phase of the Fourier-domain modes, to leave the coorbital modes $\tilde{h}_{\ell m}^c$ with an approximately constant phase in the inspiral regime. Namely, for the inspiral regime, where the SPA is valid, $t^k(f) = t^{\ell m}(mf)$, and applying (3.11) and (3.17) gives

$$\Psi_{\ell m}(f) \simeq m \Psi_k \left(\frac{f}{m}\right) + \Delta \phi_{\ell m} + (1 - m) \frac{\pi}{4}. \tag{3.21}$$

Note that our Fourier-domain construction is approximate, and these "coorbital" quantities $\tilde{h}_{\ell m}^c$ do not correspond exactly to a coorbital frame defined in the time domain as in Refs. [42,46].

We stress that these modes are not strictly coorbital, in the sense that they are not built from a time-domain coorbital frame built from the orbital phase. Indeed, the definition (3.20) is rooted in the Fourier domain, and its physical meaning is unclear in the high-frequency range where the SPA does not apply anymore.

Thus, the basic building blocks for the ROM will be:

- (i) $\Psi_k = -\text{Arg}[\tilde{k}]$, the Fourier-domain carrier phase;
- (ii) $\operatorname{Re}(\tilde{h}_{\ell m}^c)$, the real part of the coorbital modes;
- (iii) $\operatorname{Im}(\tilde{h}_{\ell m}^c)$, the imaginary part of the coorbital modes. Conversely, to rebuild the modes $\tilde{h}_{\ell m}$ from these waveform pieces, it is enough to factor in the carrier phase as in (3.20).

5. Scaling of frequencies using ringdown frequency

One of the prerequisites of our ROM procedure is to represent the waveform on a common frequency grid. However, the frequency range covered varies with physical parameters, notably spin. In SEOBNRv4_ROM, this was alleviated by extending waveforms to higher frequencies. Here, we choose to apply a scaling to the frequencies of the waveform building blocks, depending on the ringdown frequency. For every mode (ℓ, m) and the carrier, we define

$$y_{\ell m} = \frac{2\pi}{\omega_{\ell m}^{\text{QNM}}} Mf, \qquad (3.22a)$$

$$y_k = \frac{4\pi}{\omega_{22}^{\text{QNM}}} Mf, \qquad (3.22b)$$

where $\omega_{\ell m}^{\rm QNM}$ is the quasinormal mode frequency and varies for different waveforms as it depends on the spin of the remnant black hole. We will then use for all waveforms a common grid of this rescaled parameter y. Given this scaling, we have to carefully adjust the starting frequency of the waveforms of our training set so that the frequency range of the carrier phase Ψ_k covers all modes after undoing the scaling. The maxima values of $y_{\ell m}$, y_k where we cut the data are $(y_{22}^{\rm max}, y_{21}^{\rm max}, y_{33}^{\rm max}, y_{44}^{\rm max}, y_{55}^{\rm max}) = (1.7, 1.7, 1.55, 1.35, 1.25)$ and $y_k^{\rm max} = 2.5$. This technique is only used for building the high-frequency ROM (see Sec. III F); for the low-frequency ROM, the ringdown frequency is irrelevant, and the scaling would induce an additional cost in generating the waveforms of the training set.

B. SVD decomposition

We decompose all waveform data pieces defined in Sec. III A 4 into respective singular value decomposition (SVD) bases and subsequently interpolate the projection coefficients in each SVD basis over the parameter space, as discussed in Sec. III C. This method follows earlier work in Refs. [23,39,40].

We start with a waveform data piece $X(f_i; \vec{\theta})$, given on a discrete grid in frequency $\{f_i\}_{i=1}^m$ and on a regular grid of points $\vec{\theta}$ in the three-dimensional binary parameter space in mass-ratio q and aligned BH spins χ_i , (q,χ_1,χ_2) . We flatten the parameter grid and arrange the data in matrix form $X_{ij} = X(f_i; \theta_j) \in \mathbb{R}^{m \times n}$, where n is the total number of input waveforms.

We then resample the data in a log-spaced frequency grid of 300 points. The number of points used for resampling is based on previous studies (see Refs. [23,40]). We compute the SVD [81,82] $X = V\Sigma U^T$ and obtain an orthonormal basis for the column space of the matrix X in the first $r = \operatorname{rank}(X)$ columns of V. The SVD provides us with a decomposition of the range space of X, $\operatorname{range}(X) = \operatorname{span}\{v_1, \ldots, v_r\}$, where the v_j are the *left singular vectors* of X.

Given the basis $\mathcal{B}_X = V$, we expand the waveform data pieces $x_j(f_i)$ that make up the columns of X in this basis and can write the expansion $x_j \approx \sum c_X(\theta_j) \cdot \mathcal{B}_X$ with projection coefficient matrix $c_X = \mathcal{B}_X^T \cdot X$. To construct a waveform model, we need to predict the coefficients c_X at a desired parameter space point θ^* . To do that, we need to fit or interpolate c_X over the parameter space. This is discussed in Sec. III C.

C. Tensor-product spline interpolation

In the following, we describe how we obtain projection coefficients at arbitrary parameter space points. In

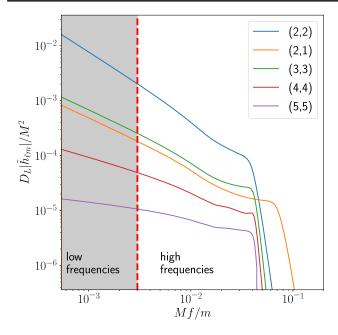


FIG. 1. The complete ROM for each waveform mode is build by hybridizing a low- and high-frequency ROM. The x axis shows the geometric frequency rescaled for each mode (ℓ, m) as Mf/m, following the natural inspiral scaling of the frequency of the waveform modes with m. The low-frequency subdomain (black shaded region) starts at a geometric frequency of Mf/m = 0.00025 and transitions to the high-frequency subdomain at Mf/m = 0.003.

low-dimensional spaces, we can afford to use dense grids built from the Cartesian product of one-dimensional sets of points. We choose cubic splines as the univariate interpolants and obtain a tensor-product interpolant (TPI) [39,59,83] for a three-dimensional coefficient tensor c_{ijk} , which can be written as

$$I_{\otimes}[c](q,\chi_1,\chi_2) = \sum_{ijk} c_{ijk}(\Psi_i \otimes \Psi_j \otimes \Psi_k)(q,\chi_1,\chi_2). \eqno(3.23)$$

Here, the Ψ are B-spline basis functions [84] of order 3 for the chosen one-dimensional sets of parameter space points in each dimension. We use "not-a-knot" boundary conditions to avoid having to specify derivatives at the domain boundaries. We built the model using TPI, a CYTHON/C package [83] to provide tensor product spline interpolation in PYTHON, and later implemented the model in LAL [80].

D. Patching in geometric frequency

Here, we discuss dividing the waveform domain in geometric frequency into separate subdomains, where we build a separate ROMs. In Sec. III F, we will instead discuss how to tackle nonuniform resolution requirements over the binary parameter space.

In the early inspiral waveforms, modes tend to be well approximated by the PN expansion, and an accurate ROM can be built from a relatively small training set. In contrast, the high geometric frequency part of the waveform modes encapsulates the late inspiral, merger, and ringdown part of the signal, which is more complex and nonlinear, and this is also the regime where EOB waveforms are tuned against NR where they are available. Building an accurate ROM for the high geometric frequency part of the waveform modes consequently requires a higher density of training set waveforms.

Therefore, it is natural to treat the low and high geometric frequency parts of the waveform separately, following the construction of previous ROMs [40]. This allows us to make the training set for the low geometric frequency part of the waveforms significantly smaller and reduce the computational cost of the training. Waveforms for low mass binaries are the most costly waveforms to generate. The cost is exacerbated due to the presence of higher modes with |m| > 2, since they require a lower starting orbital frequency to cover the same gravitational wave frequency range as the dominant mode.

In Fig. 1, we show the subdivision into low and high geometric frequency subdomains. The low-frequency subdomain is connected with PN waveforms modes in the early inspiral, as discussed in Sec. III E. We generated the SEOBNRv4HM waveforms at a sufficiently low frequency (at 15 Hz and a total mass of 5 M_{\odot} to allow for some tapering) to have the complete set of higher harmonics included in SEOBNRv4HM present at a frequency $Mf = 0.0005 * 5/2 \approx$ 0.0012, where the low geometric frequency subdomain starts. We choose the geometric transition frequency between the low- and high-frequency subdomains to be Mf = 0.003 * m, using the natural inspiral scaling of the frequency of the waveform modes with m.. For the highfrequency subdomain, we generated waveforms, choosing the starting frequency as described in Sec. III A 5, ensuring the generated waveforms after undoing the scaling (3.22a) will cover this transition frequency. The complete waveform modes are then generated by blending together the low- and high-frequency parts at the frequency using a variant of the Planck taper function described in Ref. [85].

E. Hybridization with TaylorF2

Here, we describe how we carry out the hybridization of the ROM waveform with the TaylorF2 inspiral waveform.

After evaluating the ROM waveform for all modes, we generate the TaylorF2 amplitude and phase for the (2,2) mode from the lowest frequency necessary to be able to start all inspiral modes at a user-specified frequency. We blend the TaylorF2 and ROM amplitude and phase for the (2,2) mode using the same Planck taper function used to connect high-and low-frequency ROM. We can obtain the higher mode PN inspiral waveforms by rescaling the (2,2) amplitude and phase. For the phase, we follow Eqs. (3.21) and (3.5) to

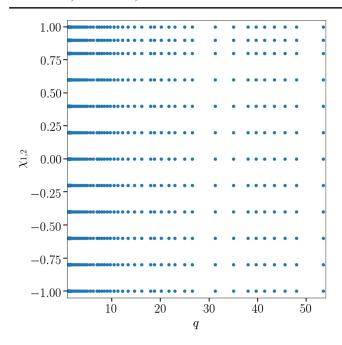


FIG. 2. Location in parameter space (q,χ_1,χ_2) of the waveforms used to build the inspiral ROM. For this ROM, both spin components use the same grid.

compute the carrier phase from the TaylorF2 (2,2) phase and rescale it to obtain the phase for each mode. We then align the inspiral phase with the ROM phase for each mode and blend them together on a common frequency grid. For the amplitude, we rescale the TaylorF2 (2,2) amplitude according to the PN amplitudes given in Ref. [86], Eqs. (12a)–(12t).

F. Patching in parameter space

As already noted for the previous ROMs of EOBNR waveforms (see Refs. [23,39,40]), model features often require more resolution in particular parts of the parameter space. However, regular grids do not allow for local refinement in (q,χ_1,χ_2) . Therefore, we partition the binary parameter space into subdomains on which resolution requirements can be satisfied with a particular regular grid choice.

The low-frequency ROM does not need any special treatment and was built using waveforms placed on a Cartesian grid with 64 points in q and 12 points in $\chi_{1,2}$ as shown in Fig. 2. Here, the one-dimensional grids in χ_1 and χ_2 were chosen to be identical. The grids for q and $\chi_{1,2}$ are the same as the ones used for SEOBNRv4_ROM (see Sec. VII in Ref. [23]), except that we limit the grid to q = 50.

On the other hand, as already noted in Refs. [23,40], modeling the nonlinear merger and ringdown part of the waveform in the high geometric frequency ROM requires a higher resolution when approaching large positive values of the primary spin. Therefore, we build two different high-frequency ROMs based on the value of the primary's spin, with one ROM having a finer grid in the region of high χ_1 . The inclusion of higher modes in the

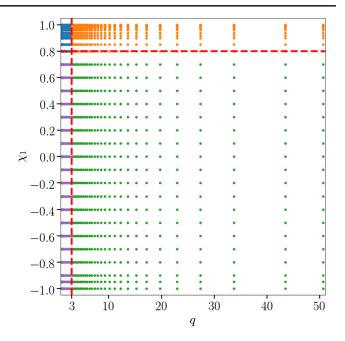


FIG. 3. Location in the parameter space (q, χ_1) of the waveforms used to build the high-frequency ROM. The dashed red lines in the plot are the boundaries of the different patches, at $\chi_1 = 0.8$ and q = 3.

SEOBNRv4HM_ROM model also requires additional resolution near equal mass. The modes with odd m vanish by symmetry on the line q=1 and $\chi_1=\chi_2$, and their behavior in the vicinity is nontrivial to model (see Refs. [25,78]). Therefore, we build two different high geometric frequency ROMs in mass ratio, one of which is covering the region $q \to 1$ with a finer grid. In total, we then have four high-frequency ROMs to correctly model the merger and ringdown part of the signal.

The two-dimensional (2D) projection of the grid in (q,χ_1) for these four ROMs is shown in Fig. 3. Since no special choice is made for the grid in χ_2 , we have omitted plotting the grid in this dimension. We set domain boundaries at q=3 and $\chi_1=0.8$ for these four ROMs. In Table I, we collect information on how the four patches are placed in parameter space and the number of grid points in each dimension.

TABLE I. The grids for the high-frequency ROMs on the four patches in parameter space shown in Fig. 3. The physical domain covered by each patch is defined by a Cartesian product of intervals in binary parameters (q, χ_1, χ_2) . We also indicate the number of grid points in each parameter per patch.

Patch	In	terv	als in (q, χ)	Points per interval		
Patch 2 Patch 3	[1, 3] [3, 50]	U	[-1, 0.8] (0.8, 1] [-1, 0.8] [0.8, 1]	U	[-1, 1] [-1, 1]	$24 \times 24 \times 24$ $24 \times 17 \times 24$ $31 \times 24 \times 24$ $30 \times 19 \times 24$

IV. RESULTS

In this section, we discuss the accuracy and the increase in efficiency of SEOBNRv4HM_ROM compared to SEOBNRv4HM. Finally, we also perform a parameter estimation study to demonstrate the potential of this model in data analysis applications.

A. Accuracy of the model

We start by defining the faithfulness function, used to assess the closeness between two waveforms when higherorder modes are included. We then use this faithfulness measure to determine how accurately the ROM reproduces SEOBNRy4HM waveforms.

1. Definition of faithfulness

A GW signal emitted by a spinning, nonprecessing, and noneccentric BBH is characterized by 11 parameters in the detector frame. These parameters are the BH masses m_1 and m_2 , the (constant) projection of the spins in the direction perpendicular to the orbital plane χ_1 and χ_2 , the angular position of the line of sight measured in the source's frame (ι, φ_0) , the sky location of the source in the detector frame (θ, ϕ) , the luminosity distance D_L , the time of arrival t_c of the signal, and finally the polarization angle ψ . The detector response can be written as

$$h \equiv F_{+}(\theta, \phi, \psi) h_{+}(\iota, \varphi_{0}, D_{L}, \xi, t_{c}; t)$$
$$+ F_{\times}(\theta, \phi, \psi) h_{\times}(\iota, \varphi_{0}, D_{L}, \xi, t_{c}; t), \qquad (4.1)$$

where masses and spins are combined in the vector $\boldsymbol{\xi} \equiv (m_1, m_2, \chi_1, \chi_2)$ and the functions $F_+(\theta, \phi, \psi)$ and $F_\times(\theta, \phi, \psi)$ are the antenna patterns [87,88]. This equation can be rewritten as

$$h = \mathcal{A}(\theta, \phi) [\cos \kappa(\theta, \phi, \psi) h_{+}(\iota, \varphi_{0}, D_{L}, \xi, t_{c}; t) + \sin \kappa(\theta, \phi, \psi) h_{\times}(\iota, \phi, D_{L}, \xi, t_{c}; t)],$$

$$(4.2)$$

with $\kappa(\theta, \phi, \psi)$ being the *effective polarization* [61] defined in the range $[0, 2\pi)$ as

$$e^{i\kappa(\theta,\phi,\psi)} = \frac{F_{+}(\theta,\phi,\psi) + iF_{\times}(\theta,\phi,\psi)}{\sqrt{F_{+}^{2}(\theta,\phi,\psi) + F_{\times}^{2}(\theta,\phi,\psi)}}, \quad (4.3)$$

where the function $\mathcal{A}(\theta,\phi)$ is an overall amplitude and is defined as

$$\mathcal{A}(\theta,\phi) = \sqrt{F_{+}^{2}(\theta,\phi,\psi) + F_{\times}^{2}(\theta,\phi,\psi)}.$$
 (4.4)

Given a GW signal h_s (SEOBNRv4HM in our case) and a template waveform h_t (SEOBNRv4HM_ROM in this context), we define the faithfulness (or match) as [25,61.89]

$$\mathcal{F}(i_{s}, \varphi_{0s}, \kappa_{s}) \equiv \max_{t_{c}, \varphi_{0t}, \kappa_{t}} \left[\frac{(h_{s}, h_{t})}{\sqrt{(h_{s}, h_{s})(h_{t}, h_{t})}} \bigg|_{\substack{t_{s} = t_{t} \\ \xi_{s} = \xi_{t}}} \right], \quad (4.5)$$

where parameters with the subscript "s" ("t") refer to the signal (template) waveform. The expression above does not depend on $\mathcal{A}(\theta,\phi)$; therefore, the only dependence on (θ,ϕ,ψ) is encoded in $\kappa(\theta,\phi,\psi)$. For the faithfulness calculation, we optimize over the phases φ_{0t} and κ_t and the time of arrival t_c because they are not interesting from an astrophysical perspective. We use the standard definition of the inner product between two waveforms (see Refs. [87,88]),

$$(a,b) \equiv 4\text{Re} \int_{f_{\text{low}}}^{f_{\text{high}}} df \frac{\tilde{a}(f)\tilde{b}^*(f)}{S_n(f)}, \qquad (4.6)$$

where denotes the Fourier transform, * indicates the complex conjugate, and $S_n(f)$ is the one-sided power spectral density (PSD) of the detector noise. For this computation, we use the Advanced LIGO "zero-detuned high-power" design sensitivity curve [90]. The integral is computed between the frequencies $f_{\text{low}} = 20 \text{ Hz}$ and $f_{\text{high}} = 3 \text{ kHz}$. The same definition of faithfulness has been used to determine the agreement between SEOBNRv4HM and numerical relativity waveforms (see Ref. [25]). Since the faithfulness given in Eq. (4.5) depends on the signal parameters $(\iota_s, \varphi_{0s}, \kappa_s)$, we will summarize the results using the maximum and the average *unfaithfulness* (or mismatch) $[1 - \mathcal{F}(\iota_s, \varphi_{0s}, \kappa_s)]$ over these parameters, namely [61,89,92],

$$\mathcal{U}_{\text{max}} \equiv \max_{l_s, \varphi_{0_s}, \kappa_s} (1 - \mathcal{F}) \equiv 1 - \min_{l_s, \varphi_{0_s}, \kappa_s} \mathcal{F}(l_s, \varphi_{0_s}, \kappa_s), \tag{4.7}$$

$$\begin{split} \bar{\mathcal{U}} &\equiv \langle 1 - \mathcal{F} \rangle_{\iota_{s}, \varphi_{0s}, \kappa_{s}} \\ &\equiv 1 - \frac{1}{8\pi^{2}} \int_{0}^{2\pi} d\kappa_{s} \int_{-1}^{1} d(\cos \iota_{s}) \int_{0}^{2\pi} d\varphi_{0s} \mathcal{F}(\iota_{s}, \varphi_{0s}, \kappa_{s}). \end{split}$$

$$(4.8)$$

2. Faithfulness against SEOBNRv4HM

To avoid biases in data analysis applications when using the ROM instead of SEOBNRv4HM, it is important to verify that the additional modeling error introduced in the construction of the ROM is negligible compared to the inaccuracy of the SEOBNRv4HM waveforms with respect to the NR simulations. Since the typical unfaithfulness

⁵In Ref. [25], we used an old version of the Advanced LIGO zero-detuned high-power design sensitivity curve (in Ref. [91]). We have checked that the difference in the faithfulness calculation when using the new version of the sensitivity curve was negligible. For this reason, here we report the calculations performed with the new curve described in Ref. [90].

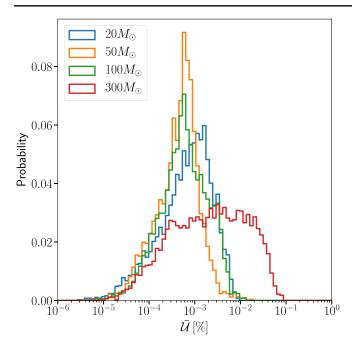


FIG. 4. Histograms of the unfaithfulness $\bar{\mathcal{U}}$ (in percent) between SEOBNRv4HM and ROM waveforms for different values of the total mass.

between SEOBNRv4HM and NR waveforms is $\mathcal{O}(1\%)$ (see Figs. (11) and (12) in Ref. [25]), it is therefore natural to require the unfaithfulness between SEOBNRv4HM and SEOBNRv4HM_ROM to be $\mathcal{O}(0.1\%)$ or less. To that end, we have generated 10,000 SEOBNRv4HM waveforms with random (uniformly distributed) values of (q,χ_1,χ_2) and computed their match against the same waveforms produced with SEOBNRv4HM_ROM.

We summarize these results in Fig. 4, where we show a histogram with the unfaithfulness $\bar{\mathcal{U}}$ computed between the ROM and SEOBNRv4HM waveforms for different values of the total mass. For each total mass, we report in Table II the median and maximum values of these unfaithfulness distributions.

The median of these mismatch distributions is weakly dependent on the total mass, and it is always less than or equal to 0.002%, while their maximum value is always less than or equal to 0.08%. In Fig. 5, we display the distribution of mismatches shown in Fig. 4 as a function of (q,χ_1) and for different values of the total mass. The largest mismatches between the ROM and SEOBNRv4HM are

TABLE II. Median and maximum values of the $\bar{\mathcal{U}}$ distributions in Fig. 4 for different values of the total mass.

Total mass (M_{\odot})	20	50	100	300
$\overline{\operatorname{Med}_{(q,\chi_1,\chi_2)}ar{\mathcal{U}}(\%)}$	0.001	0.001	0.001	0.002
$\max_{\substack{(q,\chi_1,\chi_2) \ (q,\chi_1,\chi_2)}} \bar{\mathcal{U}}(\%)$	0.01	0.02	0.01	0.08

obtained for $M=300.0~M_{\odot}$ and large negative χ_1 , as it is clear from Figs. 4 and 5 (bottom right panel). ROM GW modes are generated up to a maximum frequency (in geometric units) that scales with the inverse of the total mass of the system. For large total masses, the lack of signal above this maximum frequency is the main source of inaccuracy of the ROM. This maximum frequency for each mode is proportional to its least damped quasinormal mode frequency as described in Eq. (3.22a). The mismatch is larger for large negative spins because the least damped quasinormal mode frequency decreases in this region of the parameter space. We highlight that in this region the ROM waveforms still have mismatches ≤0.1% against SEOBNRv4HM waveforms as demanded at the beginning of this section. The results described above do not change substantially when considering the distribution of \mathcal{U}_{max} instead of $\bar{\mathcal{U}}$. In Table III, we report the median and maximum values of these distributions.

For total masses $M \le 20~M_{\odot}$, it is more convenient to summarize the results of the faithfulness calculations as a histogram with a fixed m_2 instead of the total mass. In Fig. 6, we show the $\bar{\mathcal{U}}$ distribution when fixing $m_2 = 1.4~M_{\odot}$ and varying m_1 in the interval $1.4~M_{\odot} \le m_1 \le 18.6~M_{\odot}$ such that the total mass of the system is always $M \le 20~M_{\odot}$. The median of this distribution is 0.0003%, while its maximum is 0.01%. In Fig. 7, we report the $\bar{\mathcal{U}}$ distribution in Fig. 6 as a function of (m_1,χ_1) . The accuracy of the ROM in this case degrades for large values of m_1 and large positive spins, but it is still well within the requirements. Also, in this case, the results are not very different when looking at the $\mathcal{U}_{\rm max}$ distribution for which the median is still 0.0003%, while the maximum increases to 0.02%.

These analyses demonstrate that the modeling error introduced by the ROM is negligible with respect to the difference between SEOBNRv4HM and NR waveforms. For this reason, the mismatch of the ROM against the NR waveforms is essentially the same as SEOBNRv4HM (see Figs. 11 and 12 in Ref. [25] and Fig. 6 in Ref. [44]⁶).

B. Computational performance

In this section, we discuss the computational performance of the ROM in terms of wall time for generating a waveform. We first compare the ROM to SEOBNRv4HM and then to other waveform models that include higher-order modes.

⁶The NR surrogate NRHyBsur3dq8 has a typical unfaithfulness against the NR simulations of $\mathcal{O}(10^{-3}\%)$, which is negligible with respect to the unfaithfulness between the NR simulations and the model SEOBNRv4HM (that is of $\mathcal{O}(1\%)$). Therefore, in this case, we can consider the NRHyBsur3dq8 waveform equivalent to an NR waveform. We make the same assumption in the parameter estimation study in Sec. IV C.

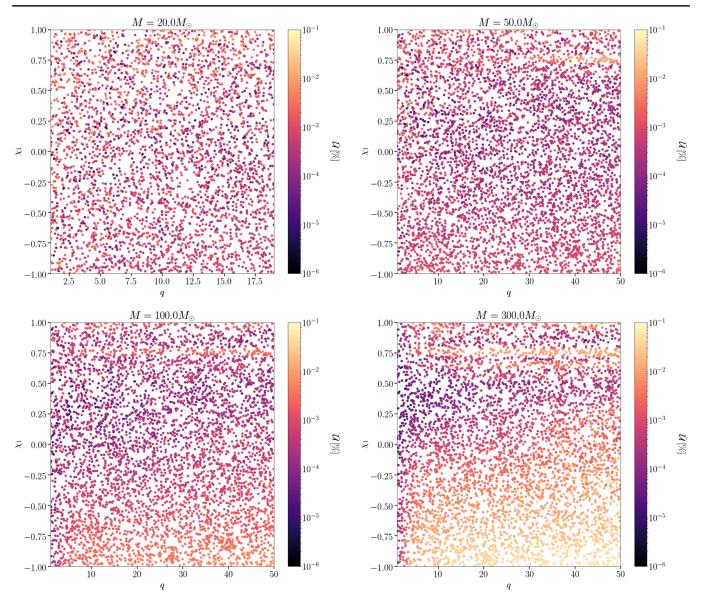


FIG. 5. Unfaithfulness $\bar{\mathcal{U}}$ between the ROM and SEOBNRv4HM as a function of (q,χ_1) and for different values of the total mass. For $M=20~M_{\odot}$ there are no data in the region q>19 because for these system m_2 would have an unphysical subsolar mass.

1. Speedup with respect to SEOBNRv4HM

The speedup of the ROM with respect to SEOBNRv4HM is computed by the ratio of the wall times of the two models for generating a frequency-domain waveform at the same parameters. Since SEOBNRv4HM is a time-domain

TABLE III. Median and maximum values of the \mathcal{U}_{max} distributions for different values of the total mass.

Total mass (M_{\odot})	20	50	100	300
$\operatorname{Med}_{(q,\chi_1,\chi_2)} \mathcal{U}_{\max}(\%)$	0.001	0.001	0.001	0.004
$\max_{(q,\chi_1,\chi_2)} \mathcal{U}_{\max}(\%)$	0.01	0.02	0.03	0.17

model, we first generate the waveform in the time domain at a sample rate of 16384 Hz and then compute its Fourier transform. The ROM waveform is already in the frequency domain, and it is generated using the sampling interval set to 1/T, where T is the duration in seconds of the associated time-domain waveform. The maximum frequency of the SEOBNRv4HM_ROM waveform is set to 8192 Hz.

In Fig. 8, we show this speedup as a function of the total mass and for different values of the mass ratio. The speedup is of order 100. It increases with mass ratio and decreases with total mass. The maximum speedup is found around a total mass of $50~M_{\odot}$. Since the spins have only a limited effect on the waveform duration, the speedup depends only weakly on them.

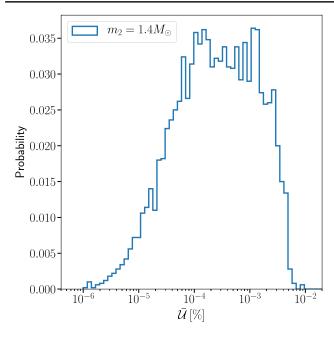


FIG. 6. Histogram of the unfaithfulness $\bar{\mathcal{U}}$ between the ROM and SEOBNRv4HM. The SEOBNRv4HM waveforms used in the match calculations have m_2 fixed to 1.4 M_{\odot} and m_1 uniformly distributed in the range 1.4 $M_{\odot} \leq m_1 \leq 18.6~M_{\odot}$.

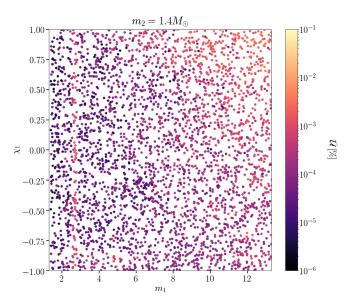


FIG. 7. Unfaithfulness $\bar{\mathcal{U}}$ between the ROM and SEOBNRv4HM as a function of (m_1, χ_1) and with $m_2 = 1.4 \ M_{\odot}$.

2. Wall time comparison

We now perform a comparison of the wall time for generating a waveform between SEOBNRv4HM_ROM and two waveform models that also include higher-order modes, namely, IMRPhenomHM [34] and NRHyBsur3dq8 [44]. As in Sec. IV B 1, we define wall time as the time to produce a frequency-domain waveform at the same parameters. Since NRHyBsur3dq8 is a time-domain model, we first

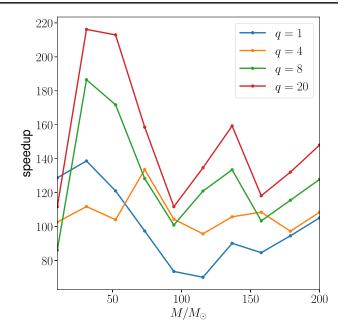


FIG. 8. Speedup of waveform generation of the ROM with respect to SEOBNRv4HM as a function of the total mass and for different values of mass ratio.

generate the waveform in the time domain at a sample rate of 16,384 Hz and then we compute its Fourier transform. For IMRPhenomHM and SEOBNRv4HM_ROM, the waveforms are generated in the frequency domain with a maximum frequency of 8192 Hz and a sampling interval of 1/T, where T is the duration in seconds of the associated time-domain waveform. The waveform models SEOBNRv4HM_ROM, IMRPhenomHM, and NRHyBsur3dq8 include a different numbers of modes in the waveform, respectively, $5[(\ell, |m|) =$ $(2,2),(2,1),(3,3),(4,4),(5,5), 6[(\ell,|m|)=(2,2),(2,1),$ (3,3), (3,2), (4,4), (4,3), and $11[(\mathcal{E}, |m|) = (2,2), (2,1),$ (2,0), (3,3), (3,2), (3,1), (3,0), (4,4), (4,3), (4,2), (5,5)modes. Since the total wall time is an increasing function of the number of modes, we also compute wall times normalized by the number of modes to factor out this effect. In Fig. 9, we show the wall time for generating a waveform with the different models as a function of the total mass for q = 8. SEOBNRv4HM_ROM has wall times of $\mathcal{O}(10)$ ms and is roughly ten times faster than IMRPhenomHM or NRHyBsur3dq8. When normalizing the wall time to the number of modes, SEOBNRv4HM_ROM is still about ten times faster than IMRPhenomHM, but only ~3 times faster than NRHyBsur3dq8.

C. Parameter estimation study

In this section, we use the SEOBNRv4HM_ROM model in a parameter estimation application. For this purpose, we create two mock signals (or injections) with the same binary parameters, using either SEOBNRv4HM or NRHyBsur3dq8 to generate the waveform. We then use SEOBNRv4HM_ROM; SEOBNRv4_ROM; and, as a comparison between waveform

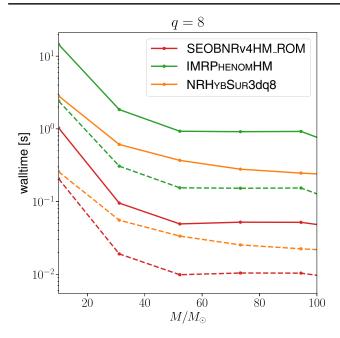


FIG. 9. Wall time comparison between different spin-aligned waveform models with higher-order modes as a function of the total mass and for q=8. The dashed lines indicate the wall time normalized by the number of modes included in the model, respectively 5 modes for SEOBNRv4HM_ROM, 6 modes for IMRPhenomHM, and 11 modes for NRHyBsur3dq8.

models for the second case, IMRPhenomHM [34] and NRHyBsur3dq8 to compute posterior distributions from the mock signals. The analysis of the first mock signal will demonstrate the improvements in measuring binary parameters when using a model with higher harmonics with respect to a model including only the dominant $(\ell, |m|) =$ (2,2) mode. The analysis of the second mock signal will give us a sense of possible biases due to modeling errors in the original SEOBNRv4HM model with respect to NR-surrogate waveforms, which are close to NR simulations. In creating the mock signals, we do not add detector noise. This choice is made to avoid additional uncertainty and bias introduced by a random noise realization. It is the natural choice given that the goal of this parameter estimation analysis is to check for possible biases due to inaccuracies in waveform models.

1. Setup

We choose parameters for the mock signals in order to emphasize the effect of higher modes in the waveform. Since the contribution of higher-order modes in the emitted GWs increases with the mass ratio, we use for the mock signals q=8, the largest mass ratio available for the model

NRHyBsur3dq8. For the total mass, we use $M=67.5~M_{\odot}$ such that the values of the component masses $m_1=60~M_\odot$ and $m_2 = 7.5 M_{\odot}$ are consistent with the masses of BBH systems observed during O1 and O2 (see Refs. [3] and [94]). The waveform models are restricted to nonprecessing spins, and we pick $\chi_{1z} = 0.5$ and $\chi_{2z} = 0.3$. To maximize the effect of the higher modes, we inject the signal at edge-on inclination $(i = \pi/2)$ with respect to the observer. The coalescence phase ϕ_c is chosen to be 1.2 radian, while the polarization phase ψ is set to 0.7 rad. The signal has been injected at Global Positioning System time (GPS) 1,249,852,257 s with a sky position defined by its right ascension of 0.33 rad and its declination of -0.6 rad. Finally, the distance of the mock signal is chosen by demanding a network signal-to-noise ration (SNR) of 21.8 in the three detectors (LIGO Hanford, LIGO Livingston, and Virgo) when using the Advanced LIGO and Advanced Virgo PSD at design sensitivity [90]. The resulting distance is 627 Mpc. We used PyCBC's PYCBC_GENERATE_HWINJ [95] to prepare the mock signal. To carry out Bayesian parameter estimation, we used the Markov chain Monte Carlo code LALInferenceMCMC [10,96]. We choose a uniform prior in component masses in the range $[3, 100]M_{\odot}$. Aligned component spins are assumed to be uniform in [-1, 1]. The priors on the other parameters are the standard ones described in Appendix C.1 of Ref. [3].

2. Results

Let us first focus on the case in which the mock signal is generated with SEOBNRv4HM. In Fig. 10, we summarize the results of the parameter estimation analysis for some relevant binary parameters. The top left panel shows the marginalized 2D posterior for the component source-frame masses, and the top right panel shows the marginalized 2D posterior for the mass ratio q and the spin parameter $\chi_{\text{eff}} = (m_1 \chi_1 + m_2 \chi_2)/(m_1 + m_2)$. In the bottom left panel, we present the marginalized 2D posterior with inclination ι and luminosity distance d_L , and, finally, in the bottom right panel, we report the matched filter SNR. The star in the plots corresponds to the true value used for the mock signal, while the 2D contours of the posterior distributions represent 90% credible regions. The waveform templates used to infer binary parameters are SEOBNRv4_ROM (blue curve) and SEOBNRv4HM_ROM (red curve). It is clear from the plots that all the parameters reported in Fig. 10 are more precisely measured when using SEOBNRv4HM_ROM instead of SEOBNRv4_ROM. The posterior volume represents the degeneracy of the gravitational wave signal, and in the absence of detector noise, this degeneracy is intrinsic to the waveforms. The inclusion of higher harmonics in SEOBNRv4HM_ROM breaks the degeneracy between the parameters $q - \chi_{\text{eff}}$ and $\iota - D_L$ and allows us to measure them more precisely. These results are consistent with what was previously found in the literature [65,97,98]. As expected, SEOBNRv4HM_ROM also measures a larger matched filter SNR.

⁷A new version of the IMRPhenom waveform model with higher-order modes became only very recently available (see Refs. [37,93]), therefore we have not been able to include it in our study. We defer comparisons with this model to future analysis.

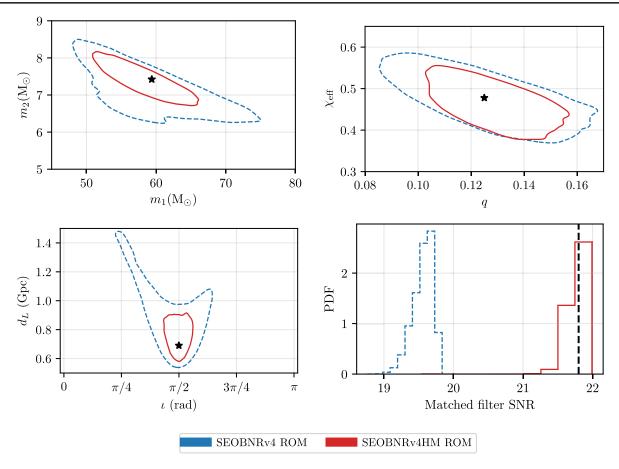


FIG. 10. 90% credible regions and histograms of posterior distributions for a q=8 BBH. The signal waveform is SEOBNRv4HM_ROM, and the stars represent binary parameters used for the signal. The mock signals are recovered with SEOBNRv4_ROM and SEOBNRv4HM_ROM waveform models. *Top left:* component masses in the source frame. *Top right:* mass-ratio and effective aligned spin parameter. *Bottom left:* inclination angle and luminosity distance. *Bottom right:* matched filter SNR.

Let us now consider the case in which the mock signal is represented by NRHyBsur3dq8. In Fig. 11, we show the marginalized 2D posterior for the mass ratio q and the

spin parameter $\chi_{\rm eff}$ (left panel) and the marginalized 2D posterior with inclination ι and luminosity distance d_L (right panel) as measured by the waveform models

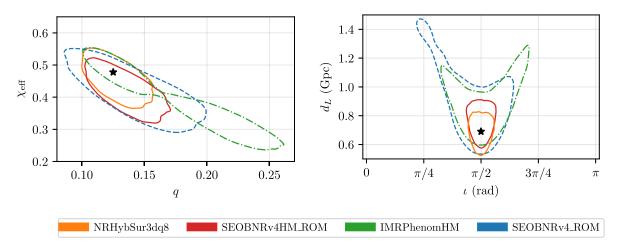


FIG. 11. 90% credible regions and histograms of posterior distributions for a q = 8 BBH. The signal waveform is NRHyBsur3dq, and the stars represent binary parameters used for the signal. The mock signals are recovered with SEOBNRv4_ROM, SEOBNRv4HM_ROM, IMRPhenomHM, and NRHyBsur3dq8 waveform models. *Left:* mass ratio and effective aligned spin parameter. *Right:* inclination angle and luminosity distance.

SEOBNRv4HM_ROM (red curve), SEOBNRv4_ROM (blue curve), IMRPhenomHM (green curve), and NRHyBsur3dq8 (orange curve). As before, the star in the plots corresponds to the true value used for the mock signal, while the 2D contours of the posterior distributions represent 90% credible regions. From the plots in Fig. 11, it is clear that, as before, SEOBNRv4HM_ROM recovers the binary parameters more precisely than SEOBNRv4 ROM. It is important to highlight that with SEOBNRv4HM_ROM the binary parameters are recovered inside the 90% credible regions. This means that for this quite asymmetric system at a moderately high SNR of \sim 20 the bias due to modeling errors in the original SEOBNRv4HM model compared to NR waveforms is negligible with respect to the statistical uncertainty. In contrast, the marginal posterior distributions recovered for IMRPhenomHM are in general broader compared to the ones recovered by SEOBNRv4HM_ROM, are notably bimodal in mass ratio and effective spin, and extend a lot farther along the line of q— $\chi_{\rm eff}$ degeneracy. In distance and inclination, the IMRPhenomHM posterior shows little improvement over SEOBNRv4_ROM which does not include higher harmonics. Finally, the marginal posteriors for NRHyBsur3dq8 are quite similar in size to those for SEOBNRv4HM_ROM, but better centered around the true parameter values. This is as expected since the likelihood should peak at the true parameter values when the signal and template use the same waveform. The mock signal is also sufficiently loud for the posteriors to be likelihood rather than prior—dominated, resulting in an unbiased parameter recovery.

This study shows that using SEOBNRv4HM_ROM for parameter estimation yields unbiased measurements of the binary parameters at a moderately high SNR even in a configuration where the effect of higher harmonics in the waveform is large. We defer a more comprehensive analysis to future studies.

V. CONCLUSION

In this paper, we have presented a fast and accurate ROM or surrogate model for the time-domain SEOBNRv4HM EOB waveform [25]. This model assumes spins aligned with the orbital angular momentum of the binary and includes the $(\ell, |m|) = (2, 1), (3, 3), (4, 4), (5, 5)$ spherical harmonic modes beyond the dominant $(\ell, |m|) = (2, 2)$ mode.

While the construction of this Fourier-domain ROM broadly follows previous work [39,40], we have introduced the following new features to accurately represent the higher harmonics and make the model more flexible (see Sec. III). While previous models used an amplitude/phase decomposition of the Fourier-domain waveform, we here define a carrier signal [see Eq. (3.15)] based on the time-domain orbital phase. Subsequently, we extract the carrier phasing from each Fourier-domain waveform mode [see Eq. (3.20)]. This essentially makes the phase of modes

almost constant in the inspiral, defining what we here call "coorbital modes." This choice allows us to avoid zero crossings in the subdominant harmonics which could spoil the smoothness of the training data and make accurate interpolation of the waveform data over parameter space very difficult. We perform alignment in the time domain to keep track of the time of coalescence of the training set waveforms, and this information is preserved in the ROM. We use here an alternative approach to dealing with the fact that the ringdown frequency varies over the parameter space, but waveform data need to be given on a common frequency grid to build a ROM. We rescale the geometric frequency parameter so that the ringdown is reached before a fixed termination frequency, which demarcates the end of the frequency grid. We use the inverse rescaling during the evaluation of the ROM. We extend the ROM to arbitrarily low frequencies by splicing it together with multipolar PN waveforms. Therefore, it can in principle be used for arbitrarily light compact binary systems. We decompose waveform input data in orthonormal bases using the SVD and build a model by constructing a tensor product spline over the three-dimensional parameter space of mass ratio and the two aligned spins of a binary. To increase model accuracy and efficiency, we use domain decomposition in frequency and in parameter space.

In Sec. IVA, we demonstrate that the ROM has a very high faithfulness (or match) with SEOBNRv4HM. Maximizing over inclination, reference phase and effective polarization of the source waveform [see Eq. (4.7)], the maximum mismatch over the remaining source parameters is below 0.03% for binaries with a total mass below $100~M_{\odot}$ and below 0.2% for binaries at 300 M_{\odot} (see Table III). Even for this very conservative choice, the mismatch is at least an order of magnitude lower than the unfaithfulness of SEOBNRv4HM against NR simulations. Therefore, the additional modeling error introduced in building the ROM is strongly subdominant, and the ROM very accurately represents the SEOBNRv4HM waveform model. In Sec. IVB, we show that our ROM accelerates waveform evaluation by a factor 100-200 compared to SEOBNRv4HM and favorably compares against other higher-mode waveform models for BBH systems, being about an order of magnitude faster. We showcase in Sec. IVC (see Figs. 10 and 11) that our ROM can recover component masses and spins, and especially the distance and inclination angle for a quite asymmetric and spinning BBH with increased precision compared to the SEOBNRv4_ROM waveform, which only models the dominant mode. In addition, we show that the ROM accurately recovers binary parameters, irrespective of whether the source is represented by a SEOBNRv4HM or a NRHyBsur3dq8 waveform. Our ROM gives a significantly more accurate parameter recovery compared to the phenomenological IMRPhenomHM waveform and is close to the NRHybSur3dq NR-surrogate model, while being more versatile and covering a significantly larger parameter space.

This ROM should prove a very useful tool for GW data analysis to describe systems where the contribution of higher harmonics is important in terms of additional signal-to-noise ratio and discriminating power for detection and parameter inference. We stress that the ROM is very fast and reproduces the SEOBNRv4HM model with a great accuracy over the widest range in parameter space of all inspiral-merger-ringdown higher-mode models available to date, from mass ratio 1 to 1:50 where aligned spins can

take values in the full range allowed for Kerr BHs, up to extremal spins.

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- [1] B. P. Abbott *et al.* (LIGO Scientific and Virgo Collaborations), Phys. Rev. Lett. **116**, 061102 (2016).
- [2] B. P. Abbott *et al.* (LIGO Scientific and Virgo Collaborations), Phys. Rev. X 6, 041015 (2016); 8, 039903(E) (2018).
- [3] B. P. Abbott *et al.* (LIGO Scientific and Virgo Collaborations), Phys. Rev. X **9**, 031040 (2019).
- [4] J. Aasi *et al.* (LIGO Scientific Collaborations), Classical Quantum Gravity **32**, 074001 (2015).
- [5] F. Acernese *et al.* (VIRGO Collaboration), Classical Quantum Gravity **32**, 024001 (2015).
- [6] LIGO Scientific and VIRGO Collaborations, Gracedb—gravitational-wave candidate event database, ligo/virgo o3 public alerts, https://gracedb.ligo.org/superevents/public/O3/.
- [7] B. P. Abbott *et al.* (LIGO Scientific and Virgo Collaborations), Astrophys. J. Lett. **892**, L3 (2020).
- [8] I. W. Harry, B. Allen, and B. Sathyaprakash, Phys. Rev. D 80, 104014 (2009).
- [9] G. M. Manca and M. Vallisneri, Phys. Rev. D 81, 024004 (2010).
- [10] J. Veitch et al., Phys. Rev. D 91, 042003 (2015).
- [11] G. Ashton et al., Astrophys. J. Suppl. Ser. 241, 27 (2019).
- [12] A. Buonanno and T. Damour, Phys. Rev. D **59**, 084006 (1999).
- [13] A. Buonanno and T. Damour, Phys. Rev. D 62, 064015 (2000).
- [14] T. Damour, P. Jaranowski, and G. Schaefer, Phys. Rev. D **62**, 084011 (2000).
- [15] T. Damour, Phys. Rev. D 64, 124013 (2001).
- [16] T. Damour, B.R. Iyer, and A. Nagar, Phys. Rev. D 79, 064004 (2009).
- [17] Y. Pan, A. Buonanno, M. Boyle, L. T. Buchman, L. E. Kidder, H. P. Pfeiffer, and M. A. Scheel, Phys. Rev. D **84**, 124052 (2011).
- [18] A. Taracchini, Y. Pan, A. Buonanno, E. Barausse, M. Boyle, T. Chu, G. Lovelace, H. P. Pfeiffer, and M. A. Scheel, Phys. Rev. D 86, 024011 (2012).
- [19] A. Taracchini, A. Buonanno, Y. Pan, T. Hinderer, M. Boyle et al., Phys. Rev. D 89, 061502 (2014).
- [20] Y. Pan, A. Buonanno, A. Taracchini, L. E. Kidder, A. H. Mroué, H. P. Pfeiffer, M. A. Scheel, and B. Szilágyi, Phys. Rev. D 89, 084006 (2014).

- [21] T. Damour and A. Nagar, Phys. Rev. D 90, 044018 (2014).
- [22] A. Nagar, T. Damour, C. Reisswig, and D. Pollney, Phys. Rev. D 93, 044046 (2016).
- [23] A. Bohé et al., Phys. Rev. D 95, 044028 (2017).
- [24] S. Babak, A. Taracchini, and A. Buonanno, Phys. Rev. D 95, 024010 (2017).
- [25] R. Cotesta, A. Buonanno, A. Bohé, A. Taracchini, I. Hinder, and S. Ossokine, Phys. Rev. D 98, 084028 (2018).
- [26] A. Nagar et al., Phys. Rev. D 98, 104052 (2018).
- [27] S. Ossokine et al., arXiv:2004.09442.
- [28] P. Ajith et al., Phys. Rev. Lett. 106, 241101 (2011).
- [29] L. Santamaría et al., Phys. Rev. D 82, 064016 (2010).
- [30] M. Hannam, P. Schmidt, A. Bohé, L. Haegel, S. Husa, F. Ohme, G. Pratten, and M. Pürrer, Phys. Rev. Lett. 113, 151101 (2014).
- [31] S. Husa, S. Khan, M. Hannam, M. Pürrer, F. Ohme, X. Jiménez Forteza, and A. Bohé, Phys. Rev. D 93, 044006 (2016)
- [32] S. Khan, S. Husa, M. Hannam, F. Ohme, M. Pürrer, X. Jiménez Forteza, and A. Bohé, Phys. Rev. D 93, 044007 (2016).
- [33] A. K. Mehta, C. K. Mishra, V. Varma, and P. Ajith, Phys. Rev. D 96, 124010 (2017).
- [34] L. London, S. Khan, E. Fauchon-Jones, C. García, M. Hannam, S. Husa, X. Jiménez-Forteza, C. Kalaghatgi, F. Ohme, and F. Pannarale, Phys. Rev. Lett. 120, 161102 (2018).
- [35] S. Khan, K. Chatziioannou, M. Hannam, and F. Ohme, Phys. Rev. D 100, 024059 (2019).
- [36] S. Khan, F. Ohme, K. Chatziioannou, and M. Hannam, Phys. Rev. D 101, 024056 (2020).
- [37] G. Pratten, S. Husa, C. Garcia-Quiros, M. Colleoni, A. Ramos-Buades, H. Estelles, and R. Jaume, arXiv:2001. 11412.
- [38] S. E. Field, C. R. Galley, J. S. Hesthaven, J. Kaye, and M. Tiglio, Phys. Rev. X 4, 031006 (2014).
- [39] M. Pürrer, Classical Quantum Gravity 31, 195010 (2014).
- [40] M. Pürrer, Phys. Rev. D 93, 064041 (2016).
- [41] J. Blackman, S. E. Field, C. R. Galley, B. Szilágyi, M. A. Scheel, M. Tiglio, and D. A. Hemberger, Phys. Rev. Lett. 115, 121102 (2015).

- [42] J. Blackman, S. E. Field, M. A. Scheel, C. R. Galley, C. D. Ott, M. Boyle, L. E. Kidder, H. P. Pfeiffer, and B. Szilágyi, Phys. Rev. D 96, 024058 (2017).
- [43] J. Blackman, S. E. Field, M. A. Scheel, C. R. Galley, D. A. Hemberger, P. Schmidt, and R. Smith, Phys. Rev. D 95, 104023 (2017).
- [44] V. Varma, S. E. Field, M. A. Scheel, J. Blackman, L. E. Kidder, and H. P. Pfeiffer, Phys. Rev. D 99, 064045 (2019).
- [45] B. D. Lackey, M. Pürrer, A. Taracchini, and S. Marsat, Phys. Rev. D 100, 024002 (2019).
- [46] V. Varma, S. E. Field, M. A. Scheel, J. Blackman, D. Gerosa, L. C. Stein, L. E. Kidder, and H. P. Pfeiffer, Phys. Rev. Research 1, 033015 (2019).
- [47] F. Pretorius, Phys. Rev. Lett. 95, 121101 (2005).
- [48] M. Campanelli, C. O. Lousto, P. Marronetti, and Y. Zlochower, Phys. Rev. Lett. 96, 111101 (2006).
- [49] J. G. Baker, J. Centrella, D.-I. Choi, M. Koppitz, and J. van Meter, Phys. Rev. Lett. 96, 111102 (2006).
- [50] B. Bruegmann, J. A. Gonzalez, M. Hannam, S. Husa, U. Sperhake, and W. Tichy, Phys. Rev. D 77, 024027 (2008).
- [51] J. Centrella, J. G. Baker, B. J. Kelly, and J. R. van Meter, Rev. Mod. Phys. 82, 3069 (2010).
- [52] A. H. Mroue et al., Phys. Rev. Lett. 111, 241104 (2013).
- [53] K. Jani, J. Healy, J. A. Clark, L. London, P. Laguna, and D. Shoemaker, Classical Quantum Gravity 33, 204001 (2016).
- [54] J. Healy, C. O. Lousto, Y. Zlochower, and M. Campanelli, Classical Quantum Gravity **34**, 224001 (2017).
- [55] M. Boyle et al., Classical Quantum Gravity 36, 195006 (2019).
- [56] M. Pürrer and C.-J. Haster, Phys. Rev. Research 2, 023151 (2020).
- [57] A. Nagar and P. Rettegno, Phys. Rev. D 99, 021501 (2019).
- [58] Z. Doctor, B. Farr, D. E. Holz, and M. Pürrer, Phys. Rev. D 96, 123011 (2017).
- [59] Y. E. Setyawati, M. Pürrer, and F. Ohme, Classical Quantum Gravity 37, 075012 (2020).
- [60] D. A. Brown, P. Kumar, and A. H. Nitz, Phys. Rev. D 87, 082004 (2013).
- [61] C. Capano, Y. Pan, and A. Buonanno, Phys. Rev. D 89, 102003 (2014).
- [62] I. Harry, J. Calderón Bustillo, and A. Nitz, Phys. Rev. D 97, 023004 (2018).
- [63] V. Varma, P. Ajith, S. Husa, J. C. Bustillo, M. Hannam, and M. Pürrer, Phys. Rev. D 90, 124004 (2014).
- [64] V. Varma and P. Ajith, Phys. Rev. D 96, 124024 (2017).
- [65] C. Kalaghatgi, M. Hannam, and V. Raymond, Phys. Rev. D 101, 103004 (2020).
- [66] B. Gadre, M. Pürrer, S. Field, and S. Ossokine (to be published).
- [67] G. Pratten et al., arXiv:2004.06503.
- [68] E. A. Huerta et al., Phys. Rev. D 97, 024031 (2018).
- [69] I. Hinder, L. E. Kidder, and H. P. Pfeiffer, Phys. Rev. D 98, 044015 (2018).
- [70] Z. Cao and W.-B. Han, Phys. Rev. D 96, 044028 (2017).
- [71] D. Chiaramello and A. Nagar, arXiv:2001.11736.
- [72] L. Blanchet, Living Rev. Relativity 17, 2 (2014).
- [73] E. Berti, V. Cardoso, and A. O. Starinets, Classical Quantum Gravity **26**, 163001 (2009).

- [74] Y. Pan, A. Buonanno, R. Fujita, E. Racine, and H. Tagoshi, Phys. Rev. D 83, 064003 (2011); 87, 109901(E) (2013).
- [75] T. Damour, B.R. Iyer, and A. Nagar, Phys. Rev. D 79, 064004 (2009).
- [76] T. Damour and A. Nagar, Phys. Rev. D 76, 064028 (2007).
- [77] W. Del Pozzo and A. Nagar, Phys. Rev. D **95**, 124034 (2017).
- [78] A. Nagar, G. Riemenschneider, G. Pratten, P. Rettegno, and F. Messina, arXiv:2001.09082.
- [79] L. Blanchet, Living Rev. Relativity 9, 4 (2006).
- [80] LIGO Scientific and Virgo Collaborations, LALSuite, https://git.ligo.org/lscsoft/lalsuite, 2018.
- [81] G. H. Golub and C. F. V. Loan, *Matrix Computations*, 3rd ed. (Johns Hopkins, 1996).
- [82] J. W. Demmel, Applied Numerical Linear Algebra (SIAM, 1997).
- [83] M. Pürrer and J. Blackman, TPI—Tensor Product Interpolation Package for Python, 2018, https://github.com/ mpuerrer/TPI.
- [84] C. de Boor, A Practical Guide to Splines (Springer, New York, 2001).
- [85] I. Hinder, S. Ossokine, H. P. Pfeiffer, and A. Buonanno, Phys. Rev. D 99, 061501 (2019).
- [86] C. K. Mishra, A. Kela, K. G. Arun, and G. Faye, Phys. Rev. D 93, 084054 (2016).
- [87] B. S. Sathyaprakash and S. V. Dhurandhar, Phys. Rev. D 44, 3819 (1991).
- [88] L. S. Finn and D. F. Chernoff, Phys. Rev. D 47, 2198 (1993).
- [89] I. Harry, S. Privitera, A. Bohé, and A. Buonanno, Phys. Rev. D 94, 024012 (2016).
- [90] L. Barsotti, P. Fritschel, M. Evans, and S. Gras (LIGO Collaboration), Updated advanced ligo sensitivity design curve, LIGO Document T1800044-v5, 2018, https://dcc.ligo.org/LIGO-T1800044/public.
- [91] D. Shoemaker (LIGO Collaboration), Advanced LIGO anticipated sensitivity curves, LIGO Document T0900288-v3, 2010, https://dcc.ligo.org/cgi-bin/DocDB/ShowDocument? docid=2974.
- [92] A. Buonanno, Y.-b. Chen, and M. Vallisneri, Phys. Rev. D 67, 104025 (2003); 74, 029904 (2006).
- [93] C. García-Quirós, M. Colleoni, S. Husa, H. Estellés, G. Pratten, A. Ramos-Buades, M. Mateu-Lucena, and R. Jaume, arXiv:2001.10914.
- [94] B. P. Abbott *et al.* (LIGO Scientific, Virgo Collaborations), Astrophys. J. **882**, L24 (2019).
- [95] A. Nitz, I. Harry, D. Brown, C. M. Biwer, J. Willis, T. D. Canton, C. Capano, L. Pekowsky, T. Dent, A. R. Williamson et al., gwastro/pycbc: Pycbc release v1.15.4, 2010, https://doi.org/10.5281/zenodo.3630601.
- [96] L. S. Collaboration, The lalinference package of the lal software suite is available from https://wiki.ligo.org/ DASWG/LALSuite.
- [97] P. B. Graff, A. Buonanno, and B. S. Sathyaprakash, Phys. Rev. D 92, 022002 (2015).
- [98] F. H. Shaik, J. Lange, S. E. Field, R. O'Shaughnessy, V. Varma, L. E. Kidder, H. P. Pfeiffer, and D. Wysocki, arXiv: 1911.02693.