

Title: Quasinormal-mode filters and their impacts on future black-hole spectroscopy

Speakers: Sizheng Ma

Series: Strong Gravity

Date: September 22, 2022 - 1:00 PM

URL: <https://pirsa.org/22090090>

Abstract: In this talk, I present a novel framework to do black-hole spectroscopy. This approach is based on a new technique so-called "quasinormal-mode filter", which can be classified into two subsets: rational filter and full filter. On the theoretical level, I explain how to use the rational filter to understand the ringdown of numerical-relativity waveforms. On the observational level, I introduce a way to incorporate the filter into Bayesian analysis. The new Bayesian framework not only allows us to analyze the ringdown of a real gravitational-wave event without Markov chain Monte Carlo, but also yields a natural estimate of the ringdown start time. By applying our method to GW150914, we find strong and self-consistent evidence for the first overtone from multiple perspectives. On the other hand, the relationship between the full filter and the metric reconstruction is discussed. Its connection to a numerical-relativity technique "Cauchy-characteristic Matching" is provided. In the final part of the talk, I also briefly present our recent progress on fully relativistic 3D Cauchy-characteristic Matching.

Zoom link: <https://pitp.zoom.us/j/95593408817?pwd=U2RXZDV2WUtNDlaRDVBVTJjTEdBQT09>



The QNM filter: a novel framework in black-hole spectroscopy

Sizheng Ma (Caltech)

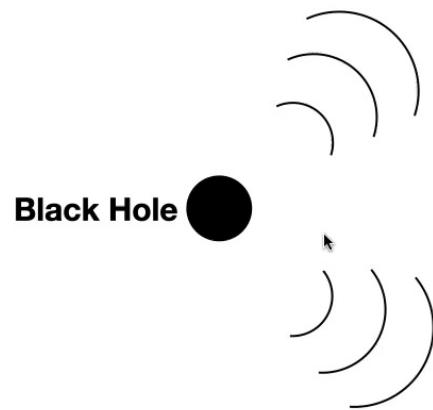
In collaboration w/: Ling Sun (OzGrav, ANU), Yanbei Chen (Caltech)

09/22/2022 @Perimeter Institute

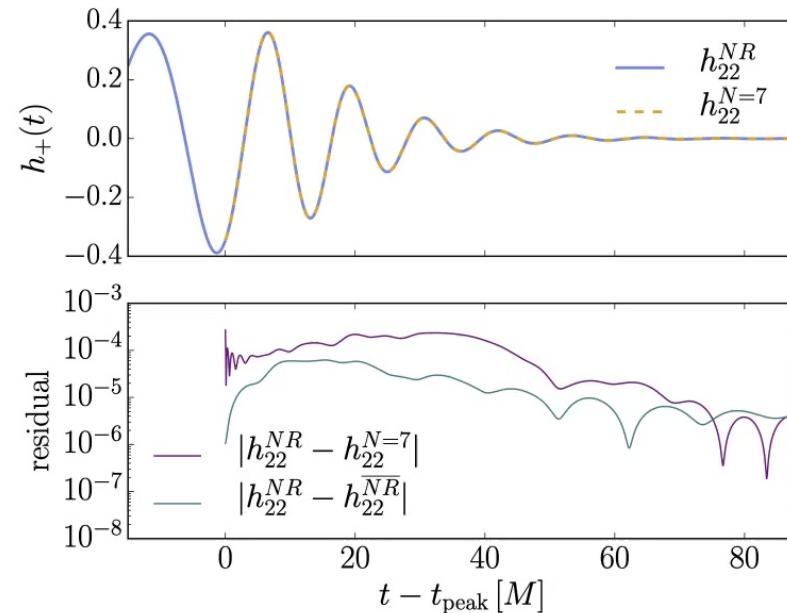
sma@caltech.edu



BH spectroscopy and ringdown modeling



$$\omega_{lmn}$$



- Overfitting?

$$h_{lm} = \sum_n A_n e^{-i\omega_{lmn} t}$$



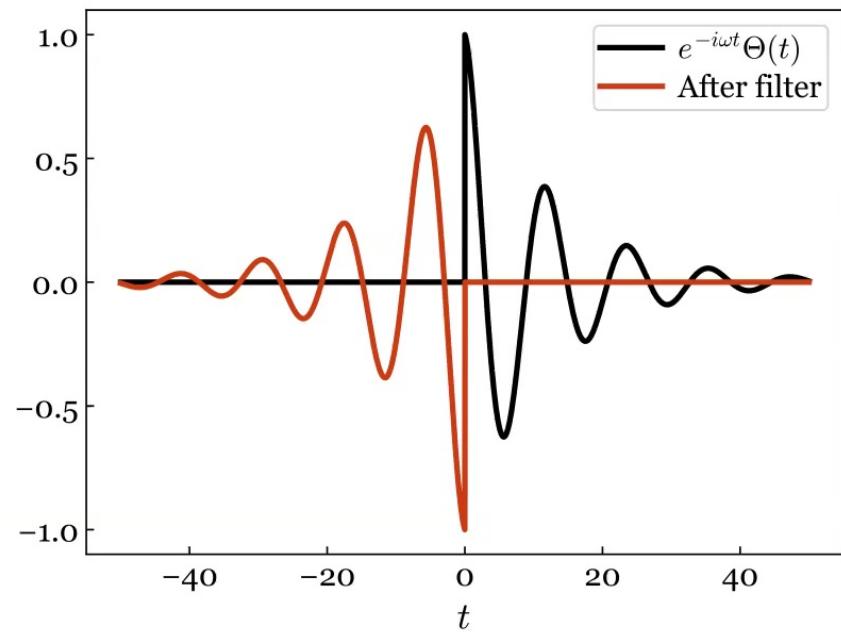
A new approach – a frequency-domain

$$h(t) = e^{-i\omega_{lmn}(t-t_0)} \Theta(t - t_0)$$

Time domain operator: $\left(\frac{d}{dt} + i\omega_{lmn} \right) h(t) = 0$

$$\begin{array}{c} \downarrow \\ \omega - \omega_{lmn} \\ \downarrow \end{array}$$

$$\frac{\omega - \omega_{lmn}}{\omega - \omega_{lmn}^*}$$



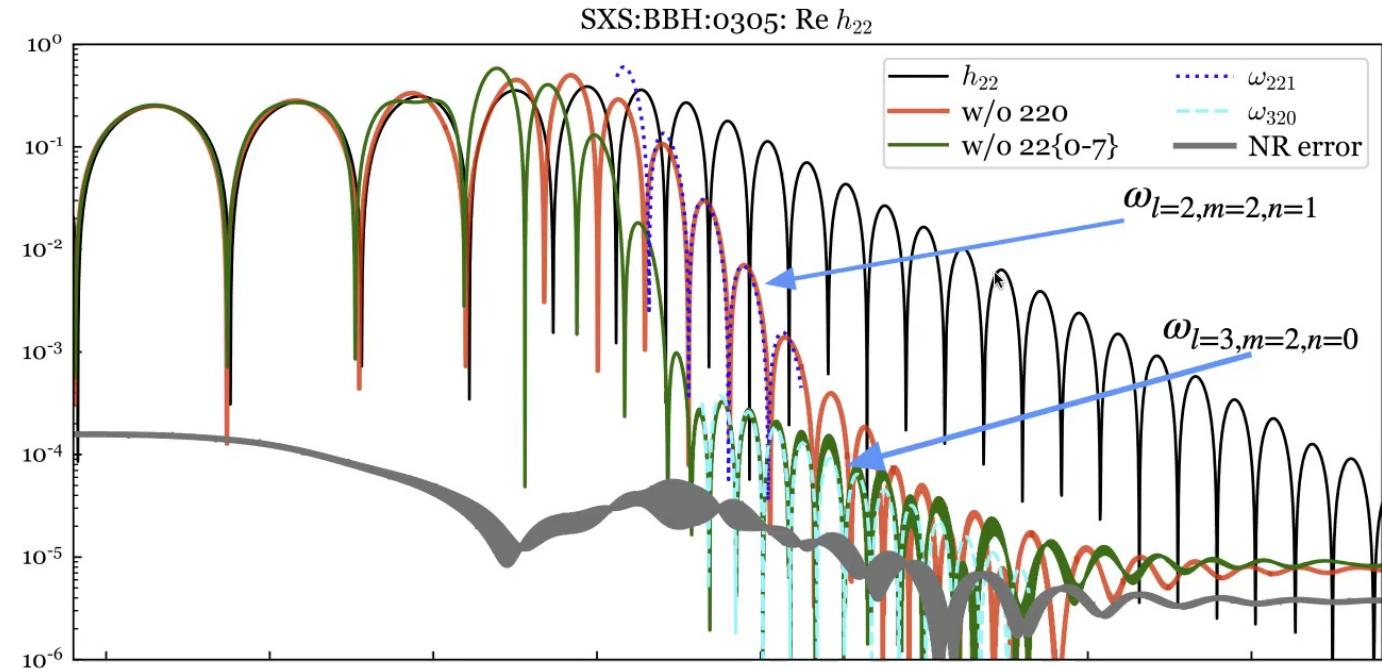
arXiv: 2207.10870



The spherical-spheroidal mixing mode

$$\text{Filter} \sim \prod_{lmn} (\omega - \omega_{lmn})$$

GW150914-like
numerical-relativity simulation

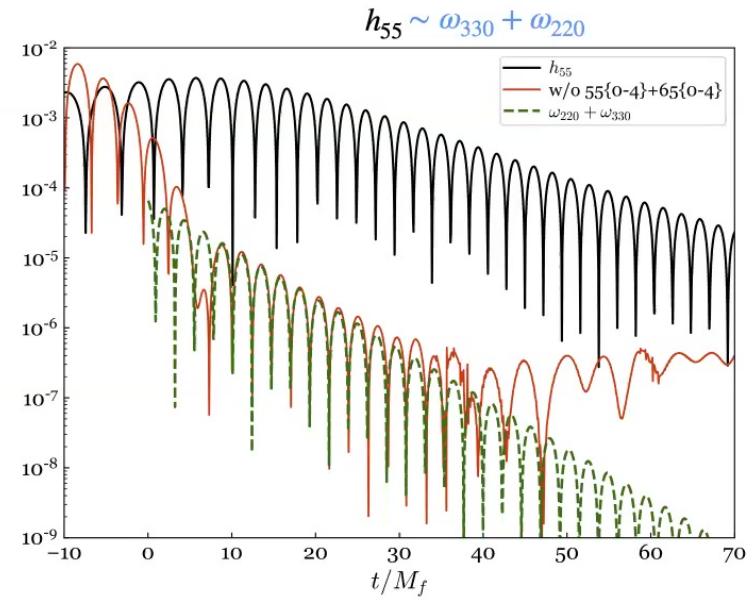
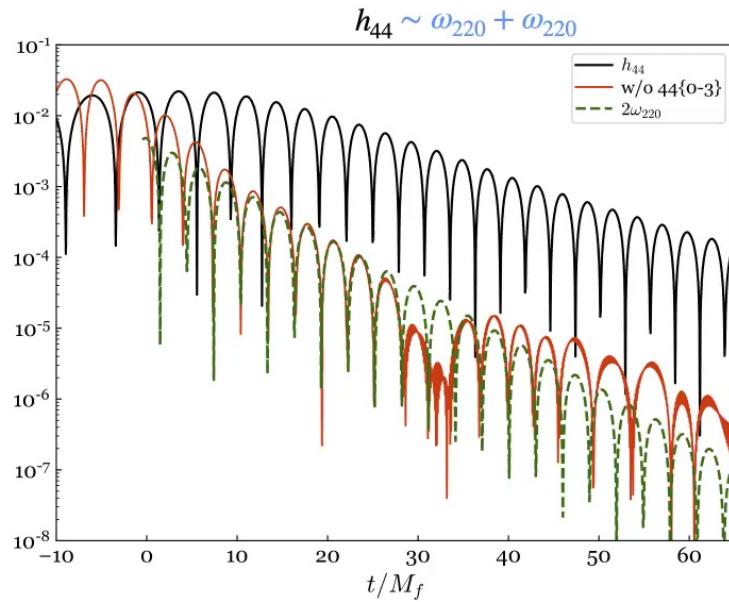


arXiv: 2207.10870

Second-order quasinormal modes

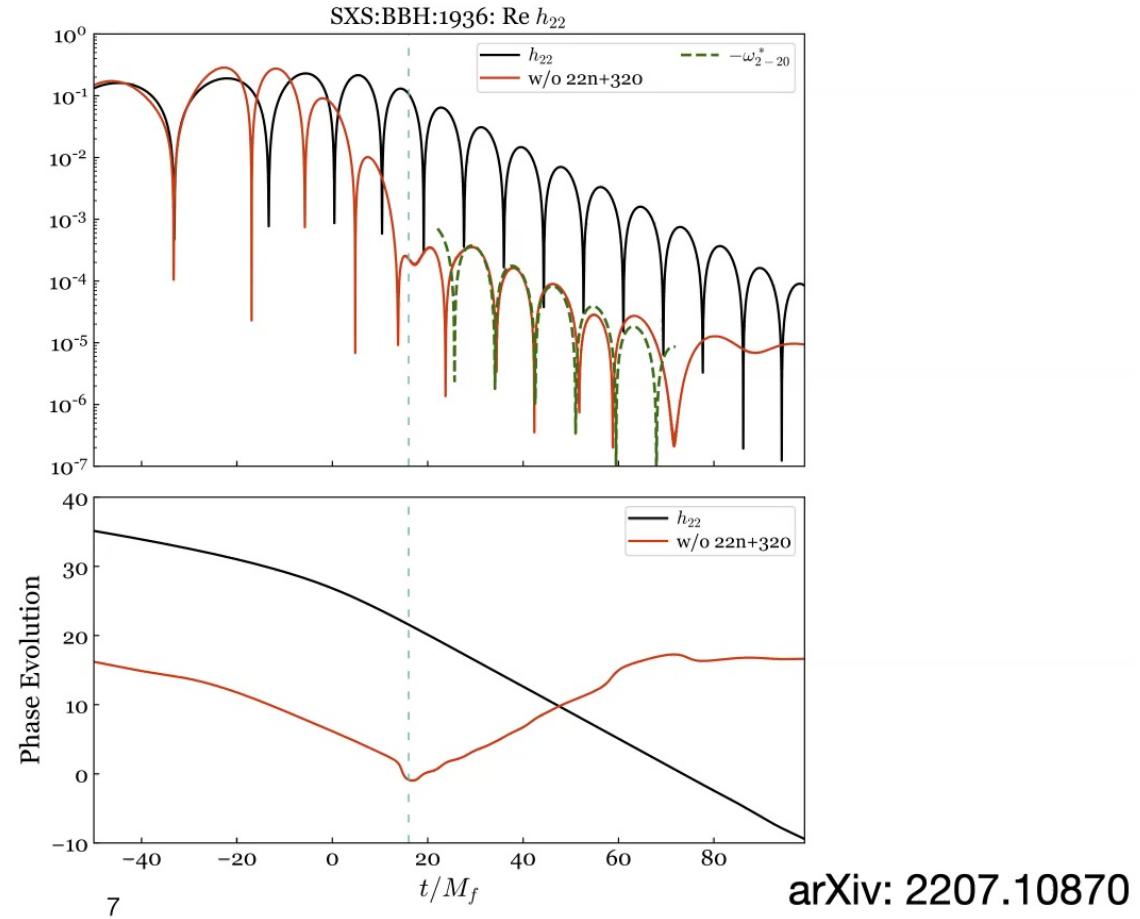
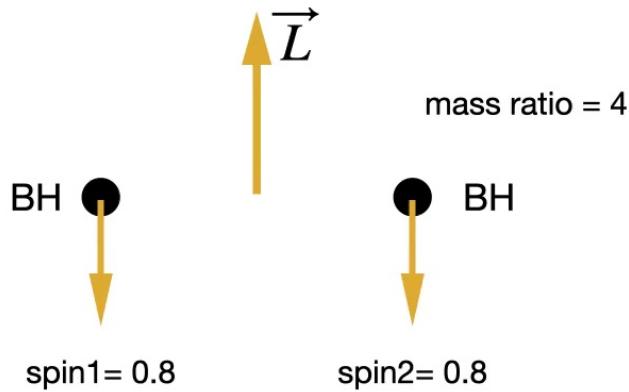


$$h_{lm} = \sum_n A_n e^{-i\omega_{lmn}t} + B e^{-i(\omega_{l_1 m_1 n_1} + \omega_{l_2 m_2 n_2})t}$$





Retrograde modes



Takeaway messages



- The goal is to reveal subdominant QNMs by removing the leading one
- The filter is imposed on the entire waveform (avoid spectral leakage)
 1. Its impact on inspiral portion: a trivial negative time and phase shift (see 2207.10870 for details)
 2. Its impact on the corresponding QNM: flipping it (see 2207.10870 for details)
 3. Its impact on other QNMs: reducing their amplitudes (Ma et al. in prep)
- The corresponding QNM can be removed regardless of its amplitude and phase, as well as inclination and polarization angles!

arXiv: 2207.10870



Outline

- Properties of the rational filter in terms of numerical-relativity waveforms
(arXiv: [2207.10870](#))
- Incorporating the rational filter into Bayesian inference (Ma, et al., in prep.)
- The full filter (briefly, arXiv: [2207.10870](#), 2203.03174)
- Cauchy-Characteristic Matching (Ma, et al., in prep.)



Incorporating the filter into Bayesian inference

Two additional properties:

- $\mathcal{F}_{lmn}(-f) = [\mathcal{F}_{lmn}(f)]^*$



The filtered signal is still real

- $|\mathcal{F}_{lmn}(f)| = 1$, namely $\mathcal{F}_{lmn}(f) = e^{i\delta_{lmn}(f)}$



The filter has no impact on PSDs!



Incorporating the filter into Bayesian inference

Two additional properties:

- $\mathcal{F}_{lmn}(\bar{f}) = [\mathcal{F}_{lmn}(f)]^*$



The filtered signal is still real

- $|\mathcal{F}_{lmn}(f)| = 1$, namely $\mathcal{F}_{lmn}(f) = e^{i\delta_{lmn}(f)}$



The filter has no impact on PSDs!

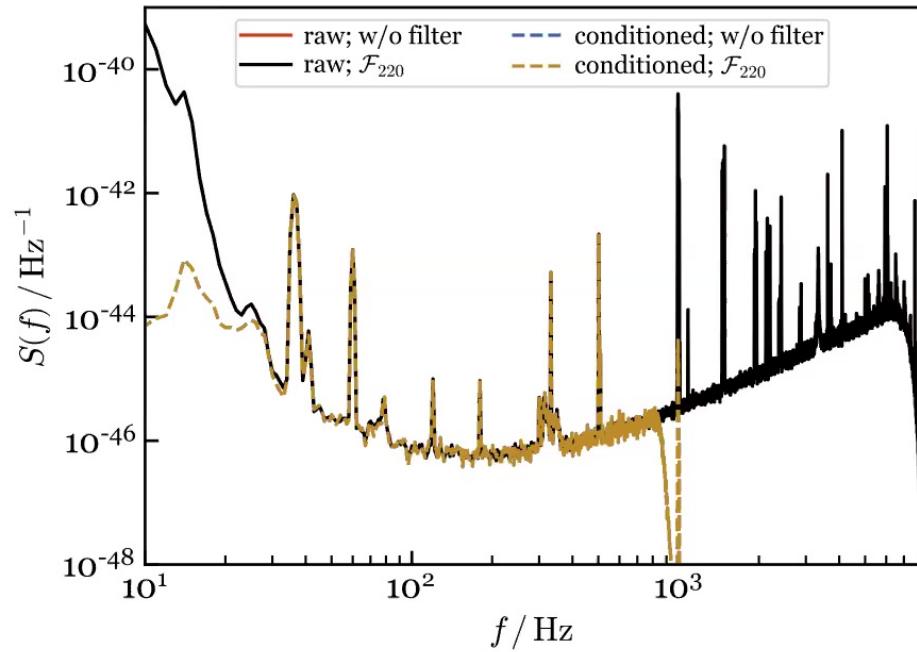


The filter has no impact on PS

$$\tilde{n}_f^F = e^{i\delta_{lmn}(f)} \tilde{n}_f$$



$$E[\tilde{n}_f^F \tilde{n}_{f'}^{F*}] = E[\tilde{n}_f \tilde{n}_{f'}^*] = \frac{1}{2} \delta(f - f') S(f)$$



32 s data of GW150914 Hanford (Welch method)

Ma, et al., in preparation

11

Isi & Farr (2021), arXiv: 2107.05609 & <https://github.com/maxisi/ringdown>

A new likelihood function

Ma, et al.



Time Domain

Isi et al., arXiv: 2107.05609, 1905.00869

Traditional likelihood function

$$\ln P^{\text{old}}(d_i | A_{lmn}, \phi_{lmn}, M_f, \chi_f, t_0, \text{etc.}) = -\frac{1}{2} \sum_{ij} (d_i - h_i) C_{ij}^{-1} (d_j - h_j)$$

d_i : GW data $\in [t_0, t_0 + T]$
 h_i : ringdown template
 C_{ij} : autocovariance function

$$h_t = \sum_{lmn} A_{lmn} e^{-(t-t_0)/\tau_{lmn}} \cos [2\pi f_{lmn}(t-t_0) + \phi_{lmn}]$$

New likelihood function

$$\ln P^{\text{new}}(d_i | M_f, \chi_f, t_0) = -\frac{1}{2} \sum_{ij} d_i^F C_{ij}^{-1} d_j^F$$

d_i^F : filtered GW data $\in [t_0, t_0 + T]$
 C_{ij} : autocovariance function



Need only remnant mass and spin



The new likelihood is merely a 2D function

The corresponding QNM can be removed regardless of its amplitude and phase, as well as inclination and polarization angles!

$$\begin{aligned} \ln P^{\text{old}}(d_i | A_{lmn}, \phi_{lmn}, M_f, \chi_f, t_0, \text{etc.}) \\ = -\frac{1}{2} \sum_{ij} (d_i - h_i) C_{ij}^{-1} (d_j - h_j) \end{aligned}$$

versus

$$\begin{aligned} \ln P^{\text{new}}(d_i | M_f, \chi_f, t_0) \\ = -\frac{1}{2} \sum_{ij} d_i^F C_{ij}^{-1} d_j^F \end{aligned}$$



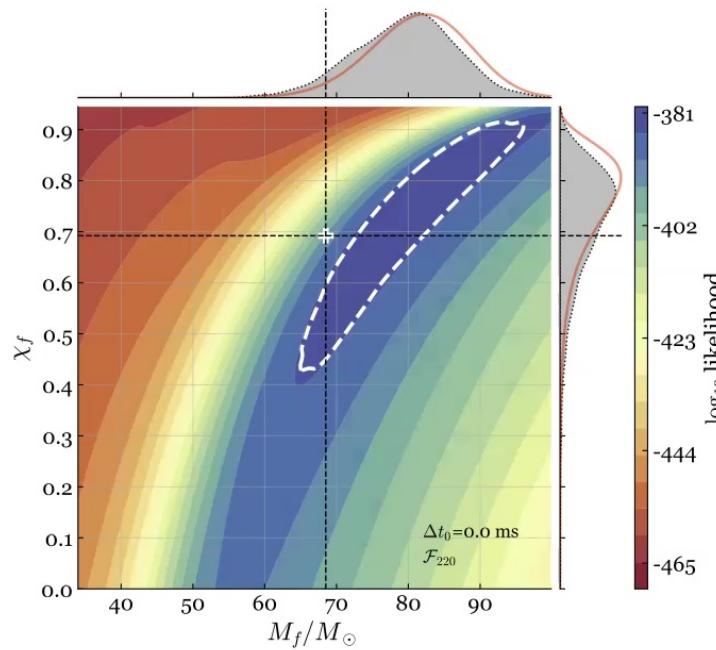
- There is no need to use fancy samplers (e.g. MCMC) to draw samplings
- We can directly plot the likelihood as a function of mass and spin (with t_0 being a hyperparameter)

Taking GW150914 as an example



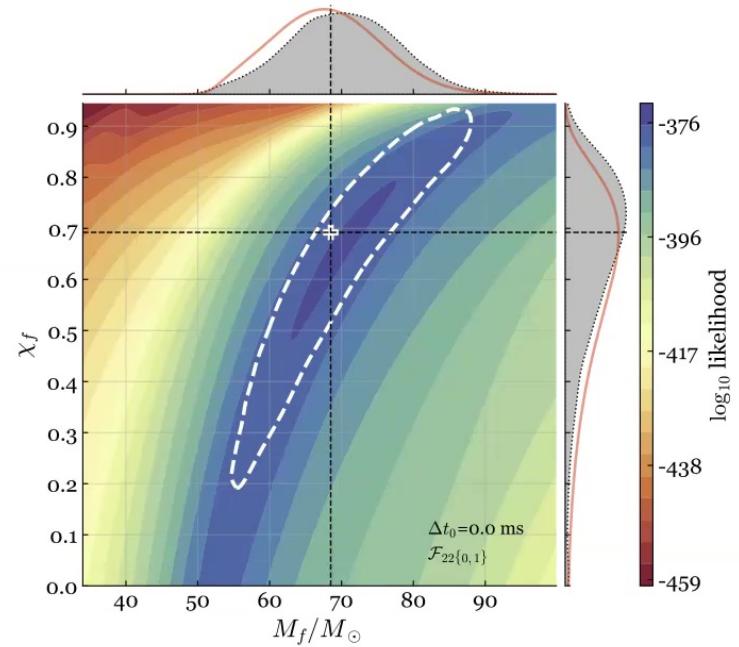
At $t_0 = t_{\text{ref}}$

Fundamental mode only



$t_{\text{ref}} = 1126259462.4083 \text{ GPS at the geocenter}$
 $T = 0.2 \text{ s}$

Fundamental mode + the first overtone



- The white dashed contour: the traditional MCMC method, with 90% credible region enclosed
- The plus sign: the IMR result



The performance of the new likelihood function

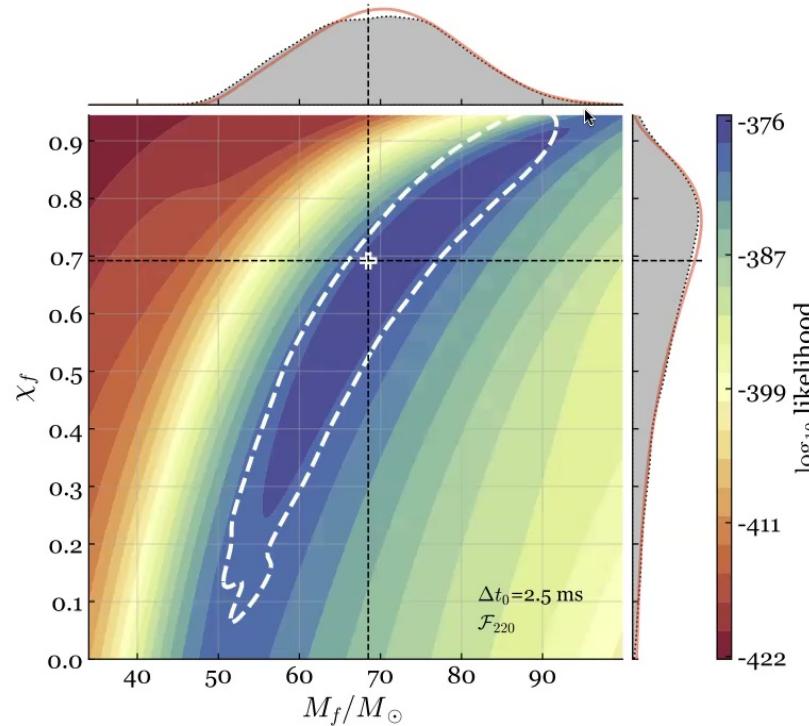
- Fast and efficient: 5.5×10^{-3} s for a single evaluation; 7.8 s for a low-res 2D plot (personal laptop, not fully parallelized); 3 mins for a high-res 2D plot (24-core cluster, not fully parallelized).
- Fully parallelizable: The evaluation of each pair of mass and spin is fully independent.
- The code doesn't slow down after increasing the number of modes because multiple filters can be imposed simultaneously.
- The posteriors are consistent with the traditional MCMC analysis.

Ma, et al., in preparation

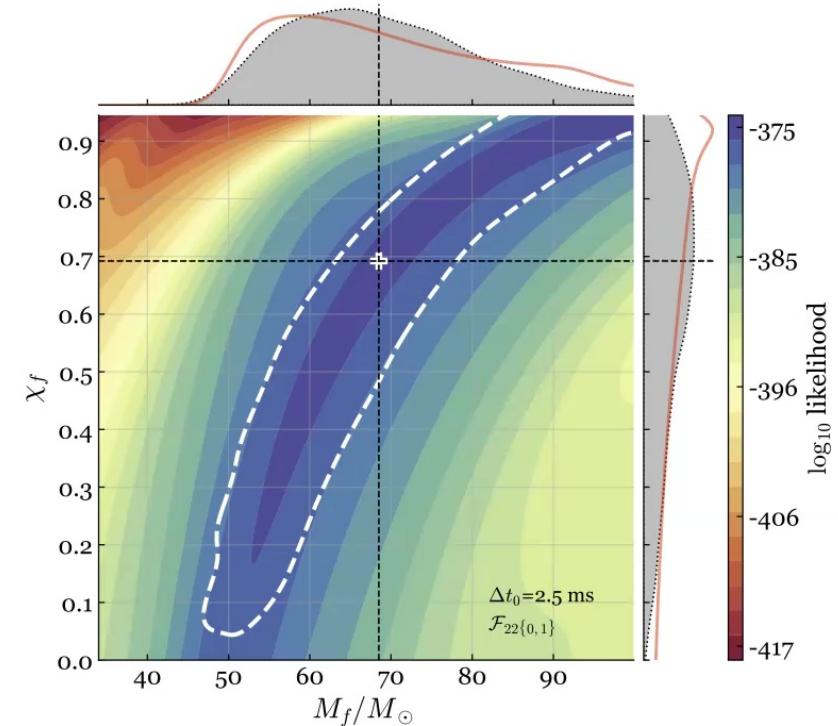


Taking GW150914 as an example

At $t_0 = t_{\text{ref}} + 2.5 \text{ ms}$ **Fundamental mode only**



Fundamental mode + the first overtone



Ma, et al., in preparation



The performance of the new likelihood function

- Fast and efficient: 5.5×10^{-3} s for a single evaluation; 7.8 s for a low-res 2D plot (personal laptop, not fully parallelized); 3 mins for a high-res 2D plot (24-core cluster, not fully parallelized).
- Fully parallelizable: The evaluation of each pair of mass and spin is fully independent.
- The code doesn't slow down after increasing the number of modes because multiple filters can be imposed simultaneously.
- The posteriors are consistent with the traditional MCMC analysis.

Ma, et al., in preparation

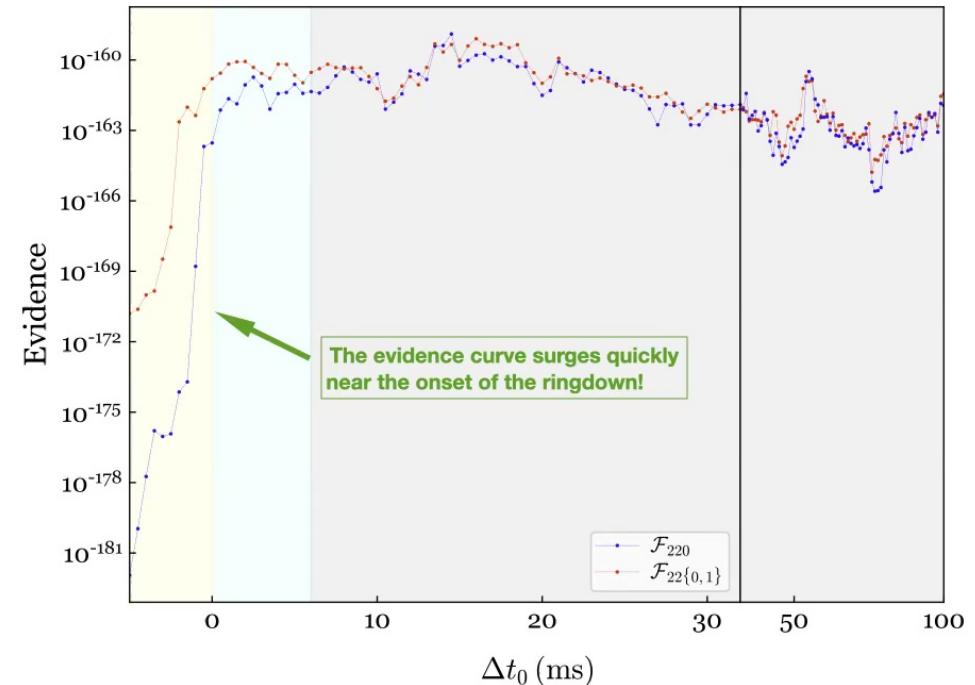


Determining the start time of ringdown (hyperparameter)

- The model evidence $P_{lmn}(t_0) = \int \int dM_f d\chi_f p(d | M_f, \chi_f, t_0) \times \text{prior}$

Red: Fundamental mode + the first overtone
 Blue: Fundamental mode

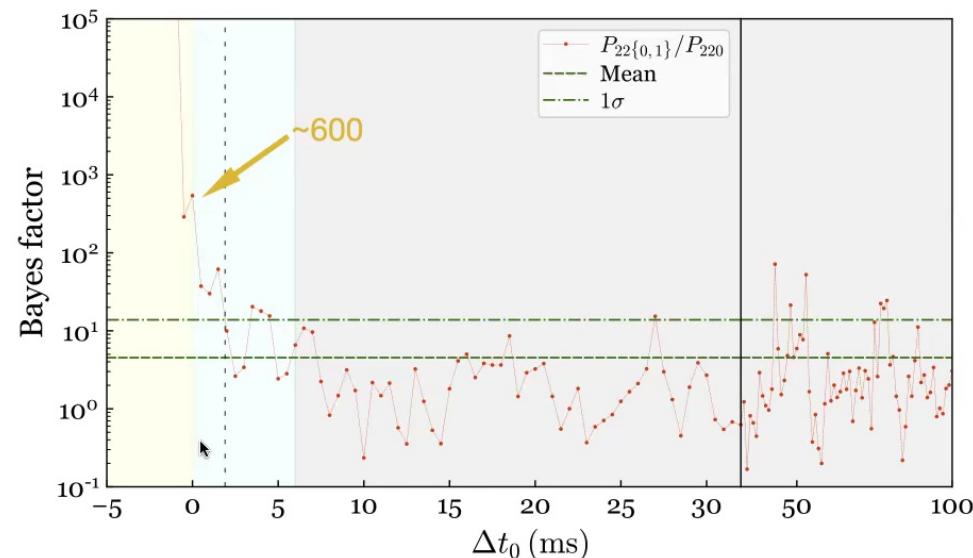
GW event trigger?



The Bayes factor of the first overtone in the ringdown of GW150914



- The model evidence $P_{lmn}(t_0) = \int \int dM_f d\chi_f p(d | M_f, \chi_f, t_0) \times \text{prior}$
- The Bayes factor: $K_{221}(t_0) = \frac{P_{22\{0+1\}}}{P_{220}}$



Ma, et al., in preparation



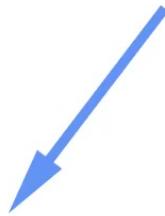
Three approaches to ringdown analysis

1. The traditional MCMC method (building a template)
2. The filter method (removing all ringdown signals)
3. A mixed approach!



The mixed approach

1. Remove a subset of QNM(s) via the filter(s)
2. Fit the filtered data to the remaining QNM(s) model via MCMC



Obtain information of the remaining modes
w/o the impact of other modes (cross-check)

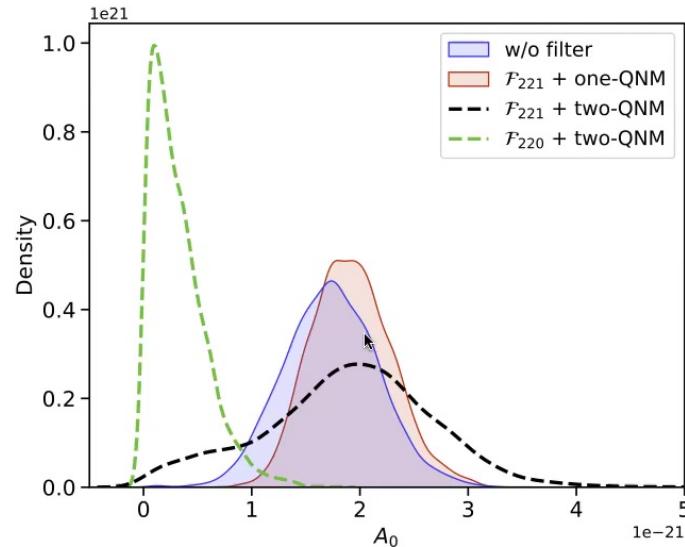


Infer remnant mass and spin from each single QNM
(test the no-hair theorem)

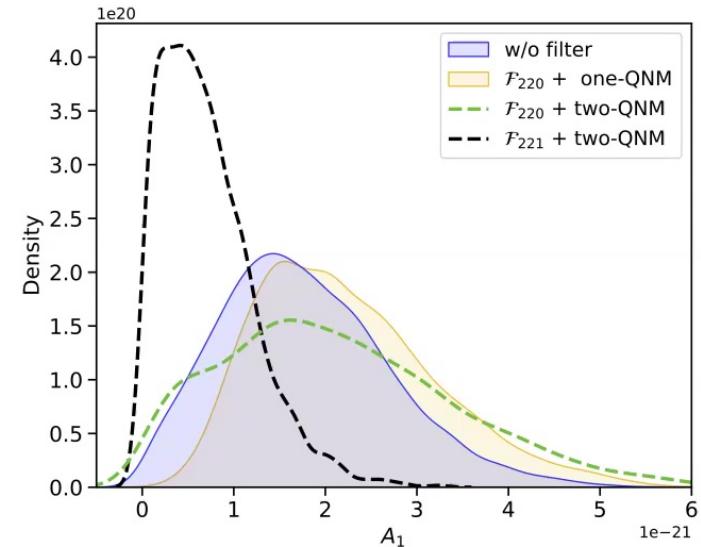


Taking GW150914 as an example

Obtain information of the remaining modes
w/o the impact of other modes (cross-check)



The amplitude of the fundamental mode



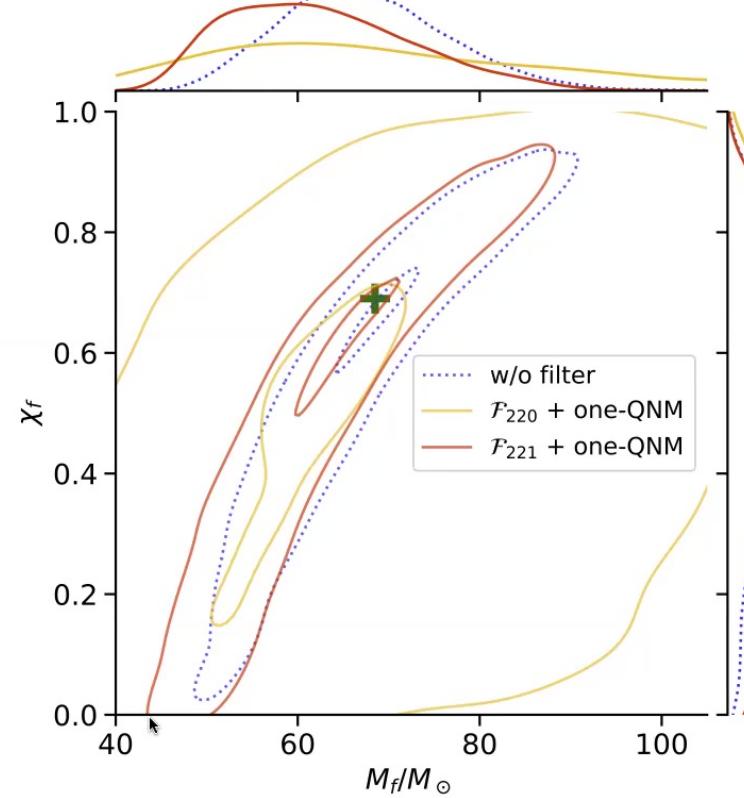
The amplitude of the first overtone

Ma, et al., in preparation



Taking GW150914 as an example

Infer remnant mass and spin from each single QNM
(test the no-hair theorem)



Ma, et al., in preparation



Outline

- Properties of the rational filter in terms of numerical-relativity waveforms
(arXiv: [2207.10870](#))
- Incorporating the rational filter into Bayesian inference (Ma, et al., in prep.)
- The full filter (briefly, arXiv: [2207.10870](#), [2203.03174](#))
- Cauchy-Characteristic Matching (Ma, et al., in prep.)

QNM filter = rational filter + full filter



The full filter

$$R_{lm}^{\text{up}} \sim \begin{cases} r^3 e^{i\omega r_*}, & r_* \rightarrow +\infty, \\ D_{lm}^{\text{out}} e^{i\omega r_*} + \Delta^2 D_{lm}^{\text{in}} e^{-i\omega r_*}, & r_* \rightarrow -\infty, \end{cases}$$



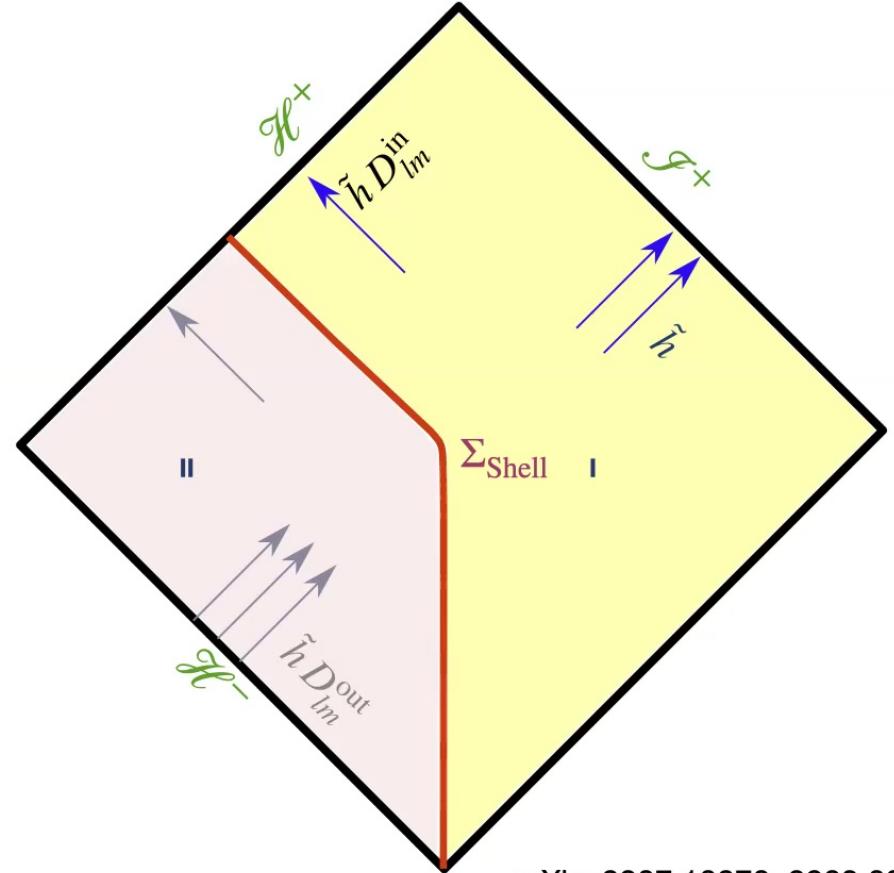
$$D_{lm}^{\text{out}}(\omega_{lmn}) \triangleq 0$$



$$D_{lm}^{\text{out}} \sim \prod_n (\omega - \omega_{lmn})$$



$$\mathcal{F}_{lm}^D = \frac{D_{lm}^{\text{out}}}{D_{lm}^{\text{out}*}}$$



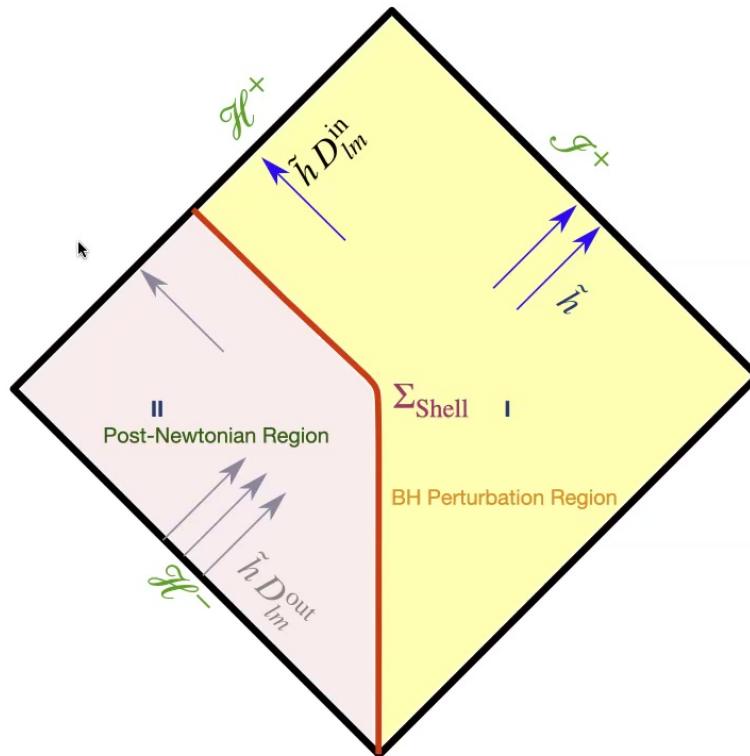
arXiv: 2207.10870, 2203.03174



The hybrid waveform model

Nichols and Chen (2010) & (2012)

arXiv: 1007.2024, 1109.0081



25

Ma et al., arXiv: 2203.03174



The image wave

$$R_{lm}^{\text{up}} \sim \begin{cases} r^3 e^{i\omega r_*}, & r_* \rightarrow +\infty, \\ D_{lm}^{\text{out}} e^{i\omega r_*} + \Delta^2 D_{lm}^{\text{in}} e^{-i\omega r_*}, & r_* \rightarrow -\infty, \end{cases}$$

$$\mathcal{F}_{lm}^D = \frac{D_{lm}^{\text{out}}}{D_{lm}^{\text{out}*}}$$

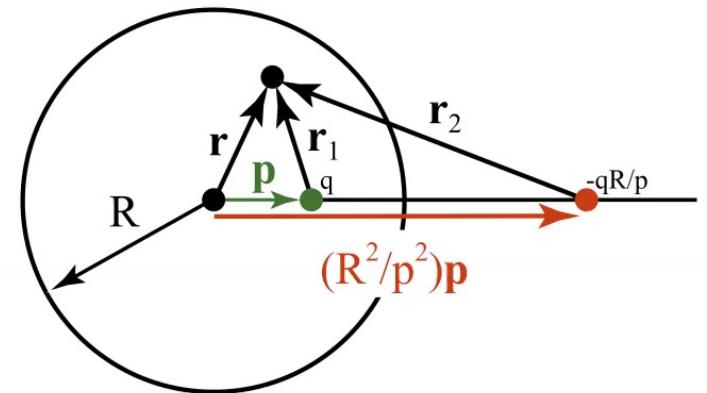
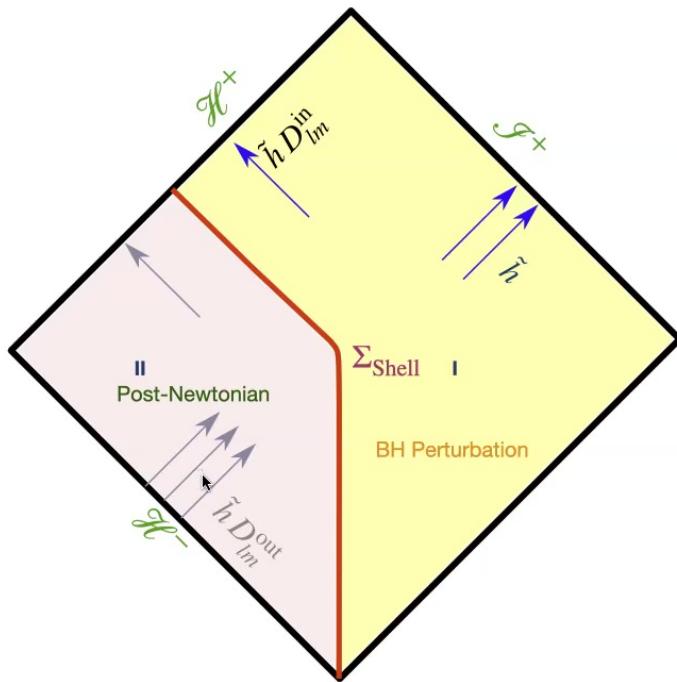
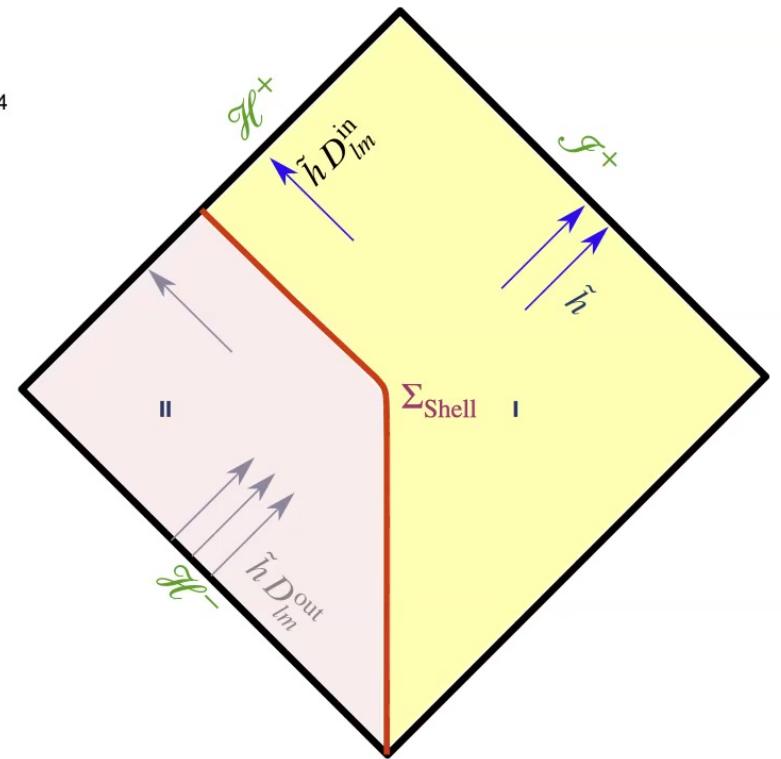


Image Credit: Wikipedia

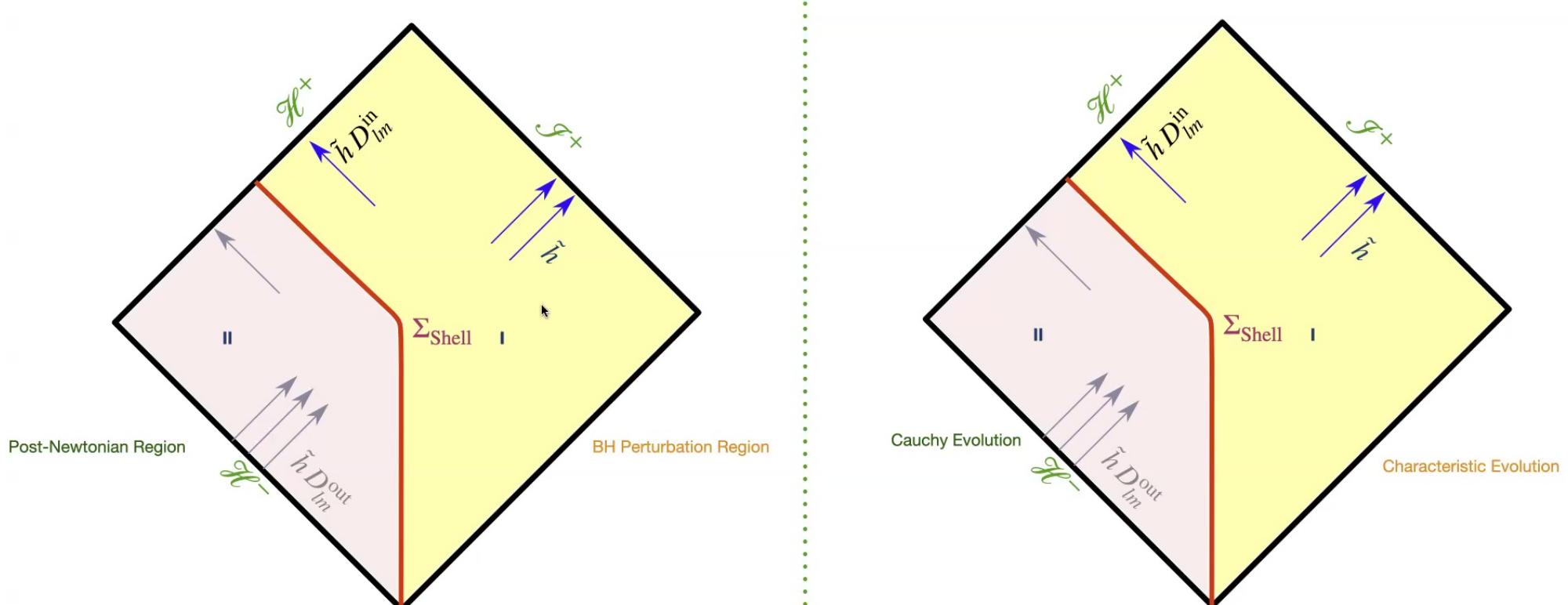
Metric reconstruction near BBHs and its applications



- Computing gravitational-wave echoes. Ma et al., arXiv: 2203.03174
- Second-order BH perturbations?



Numerical-relativity version of the hybrid method: Cauchy-Characteristic Matching



Fully relativistic, 3D CCM in SpECTRE, Ma et al., in preparation

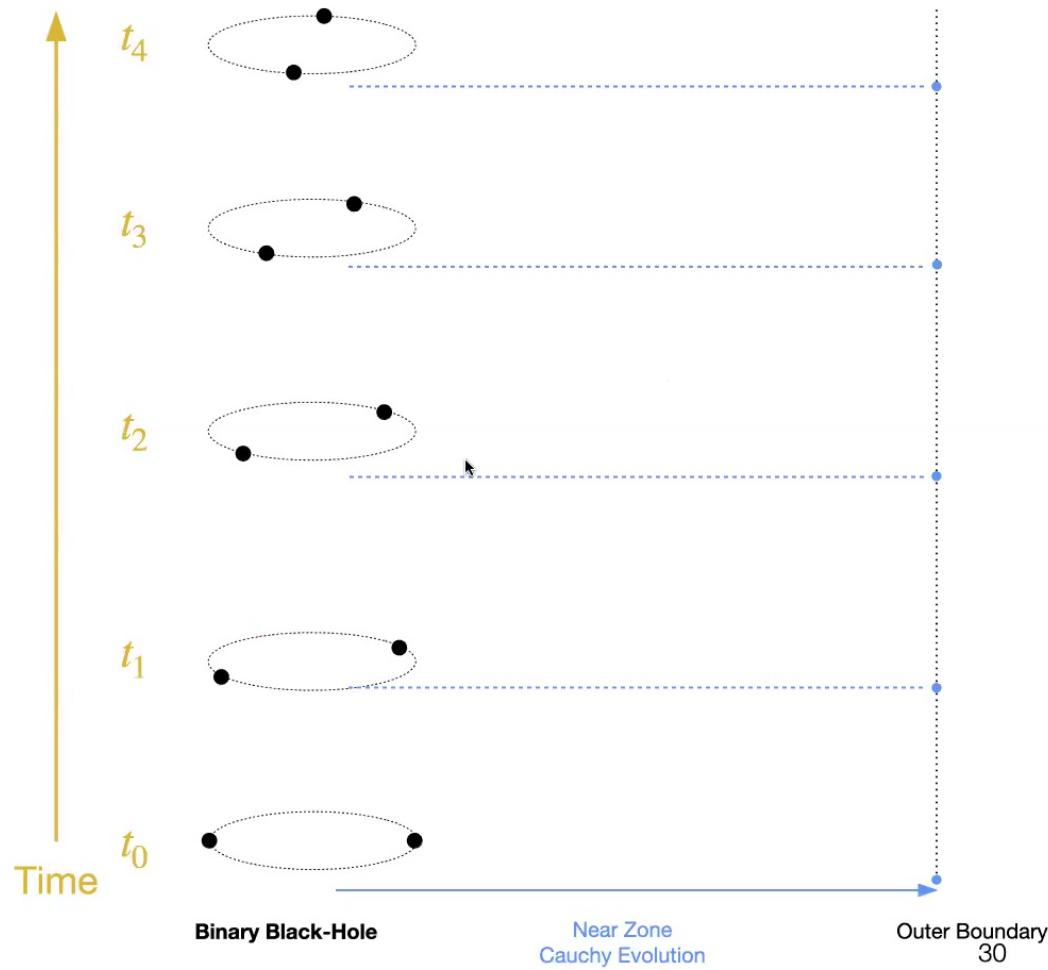


Outline

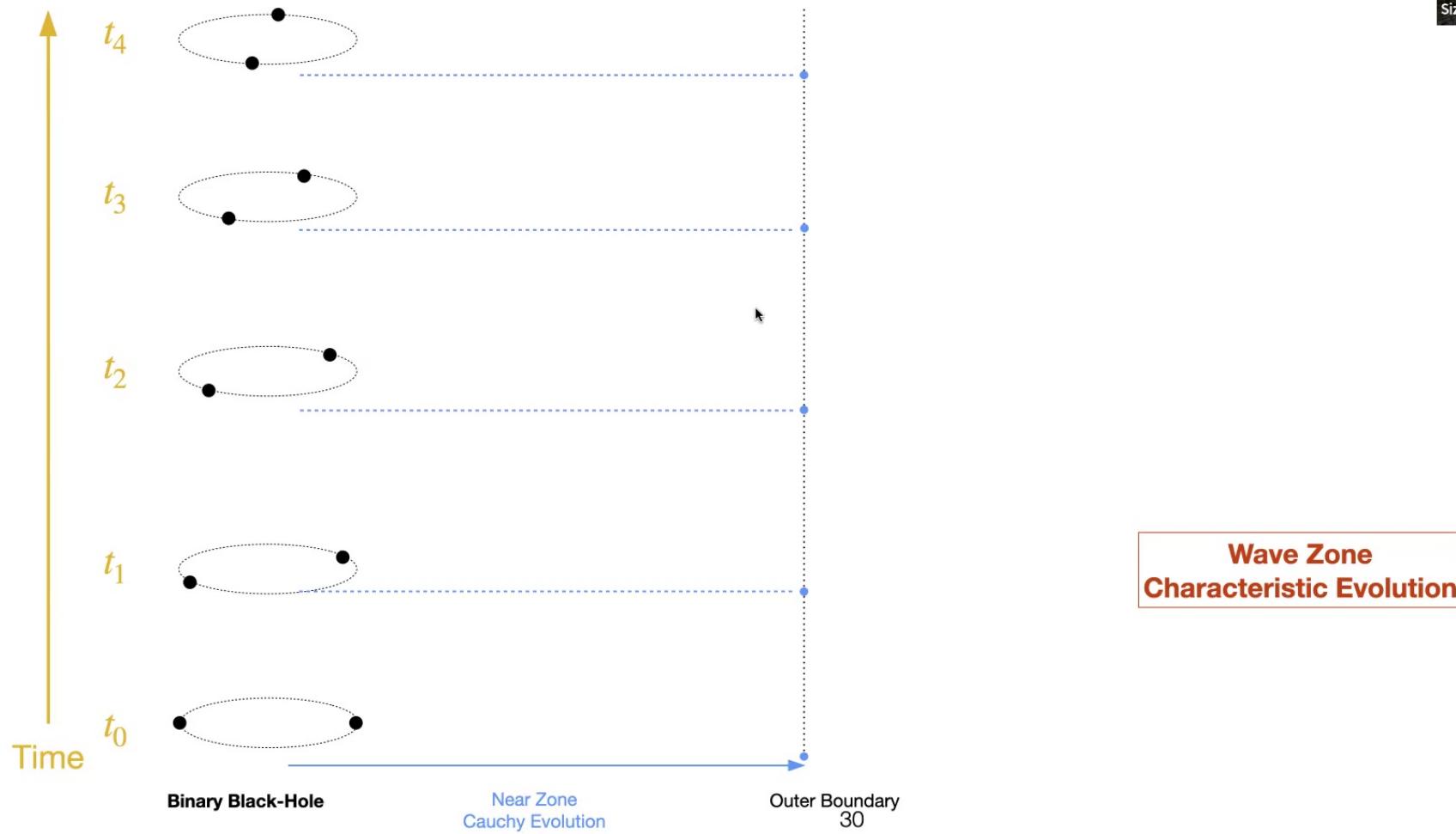
- Properties of the rational filter in terms of numerical-relativity waveforms
(arXiv: [2207.10870](https://arxiv.org/abs/2207.10870))
- Incorporating the rational filter into Bayesian inference (Ma, et al., in prep.)
- The full filter (briefly, arXiv: [2207.10870](https://arxiv.org/abs/2207.10870), 2203.03174)
- Cauchy-Characteristic Matching (briefly, Ma, et al., in prep)



Cauchy-Characteristic Extraction



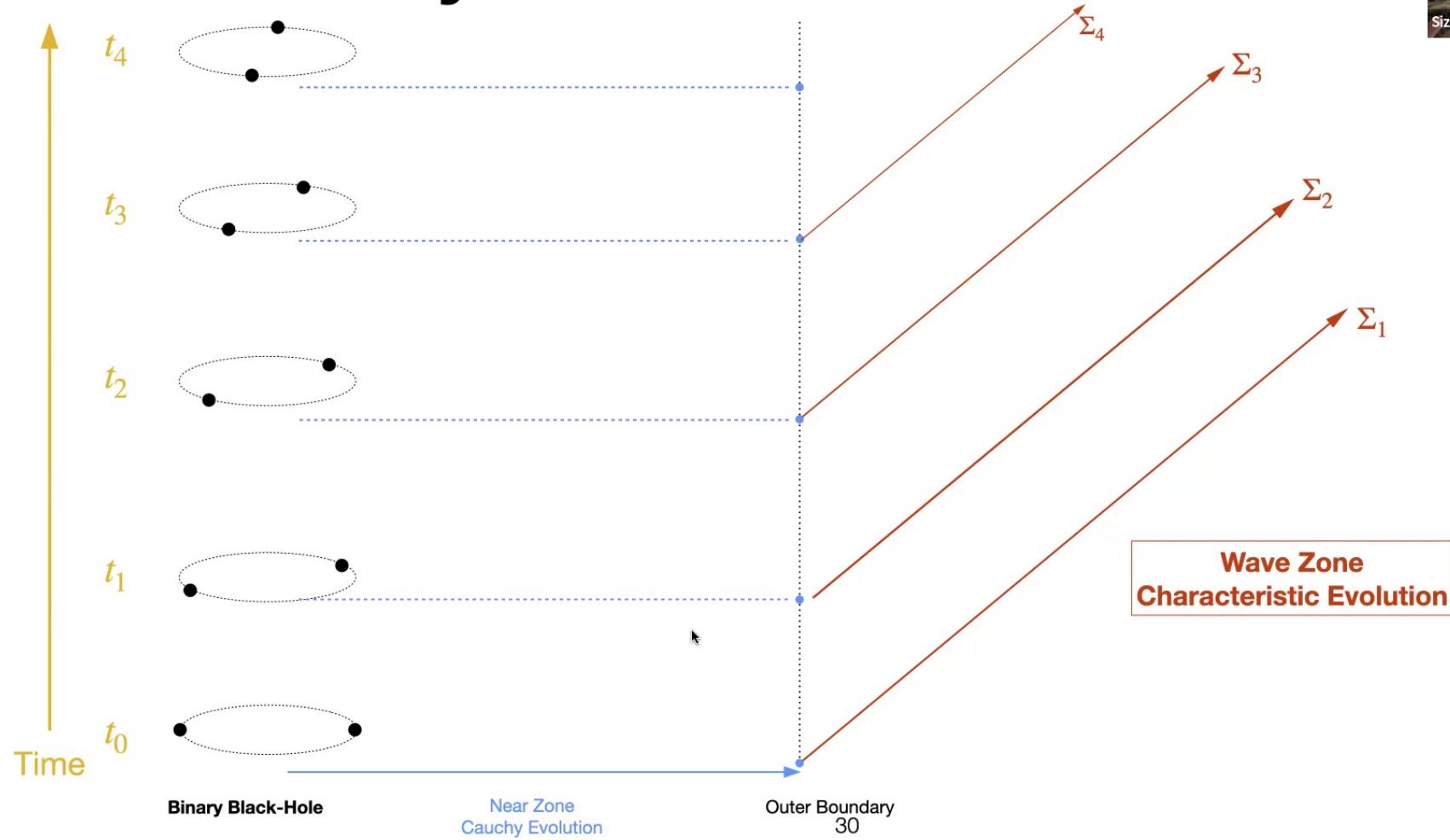
Cauchy-Characteristic Extraction



Cauchy-Characteristic Extraction

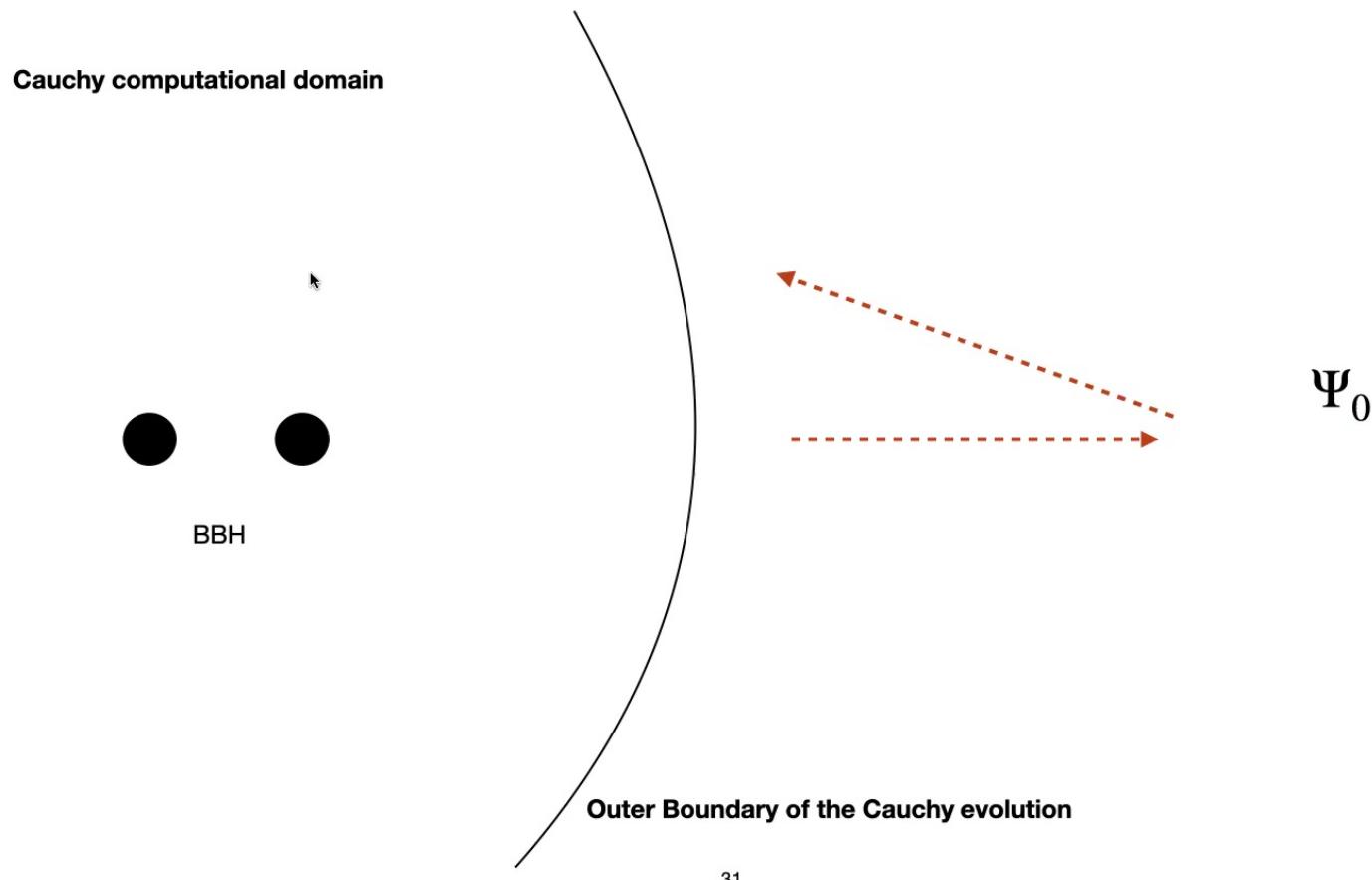


LIGO



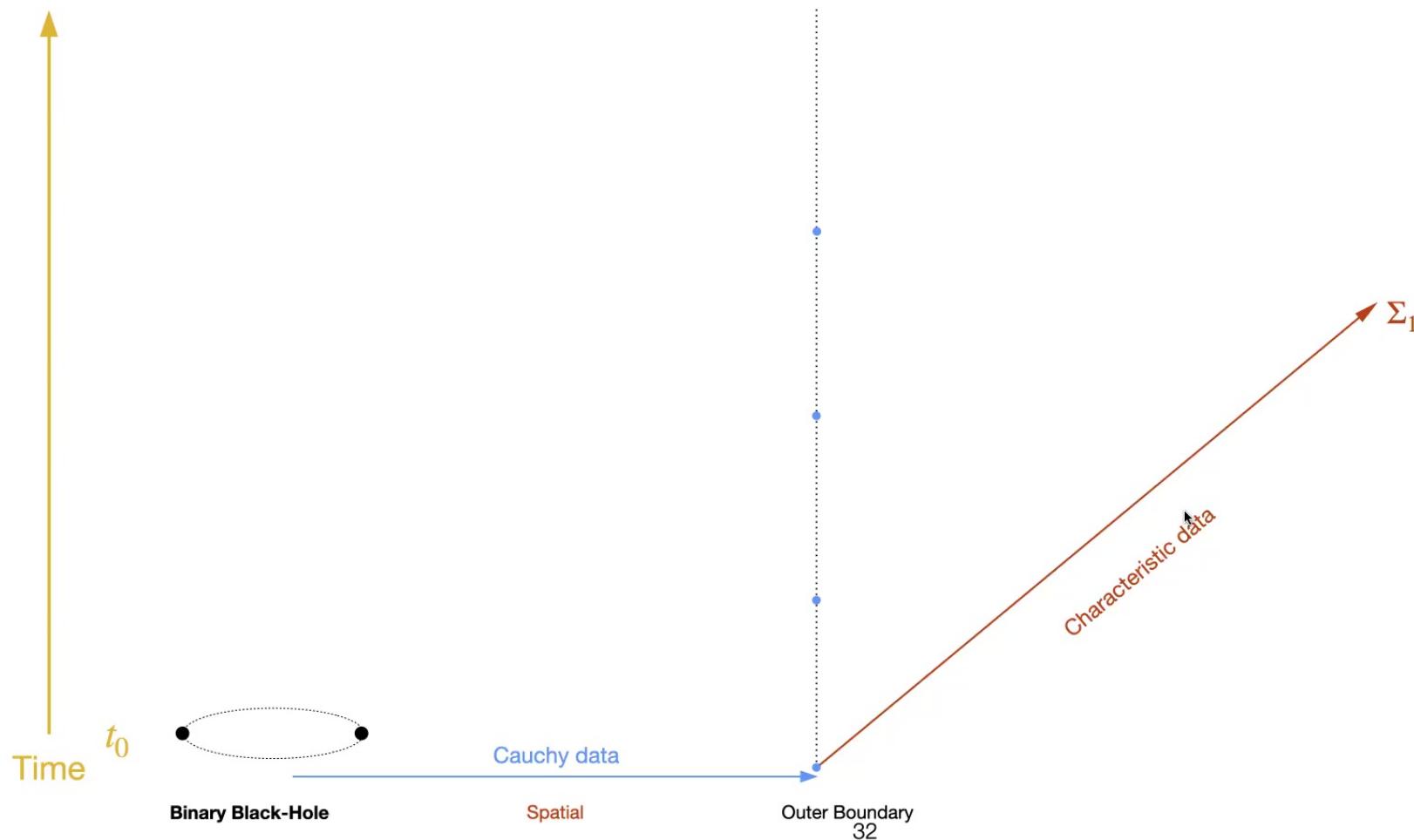


The backscattered gravitational wave

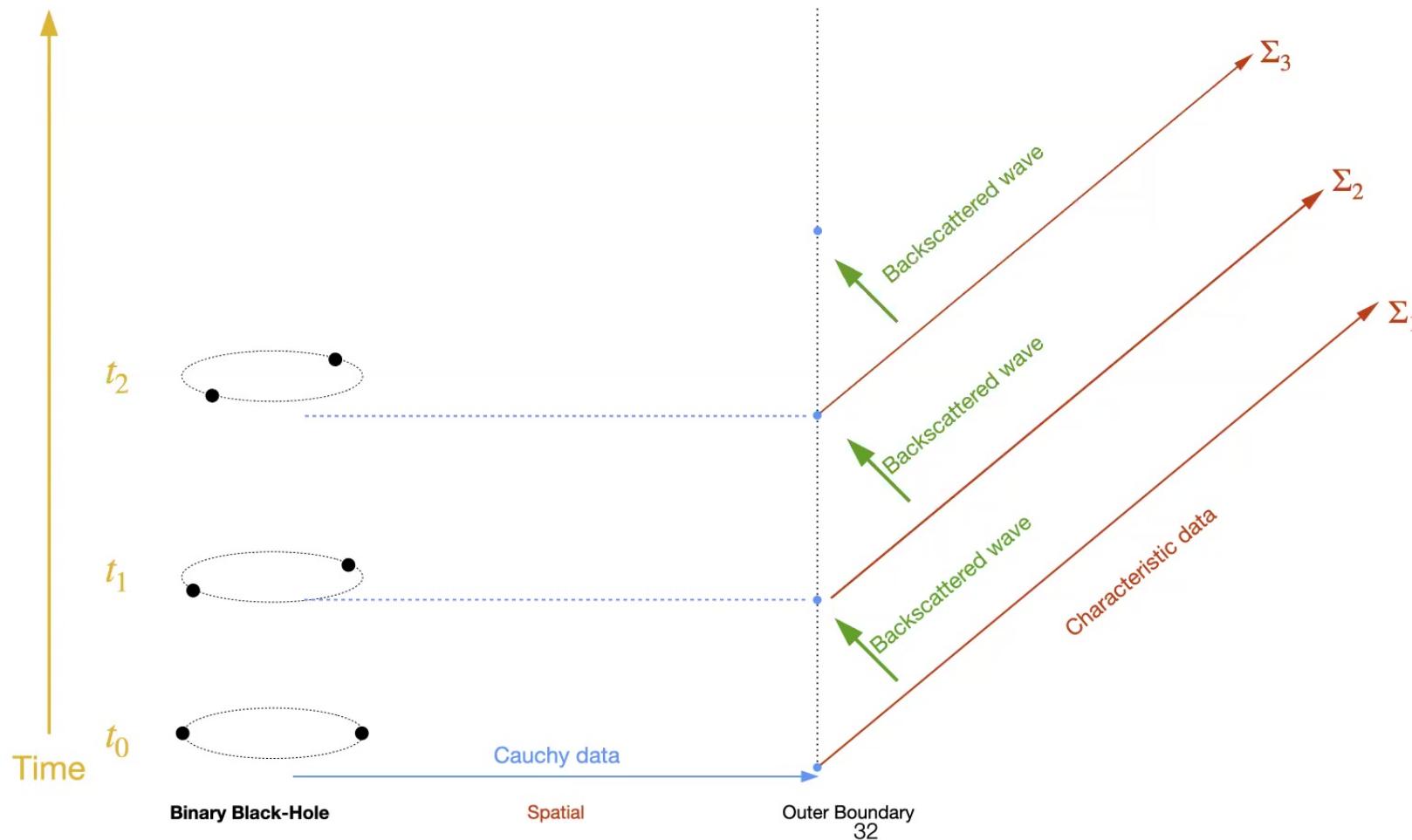


31

Cauchy-Characteristic Matching



Cauchy-Characteristic Matching

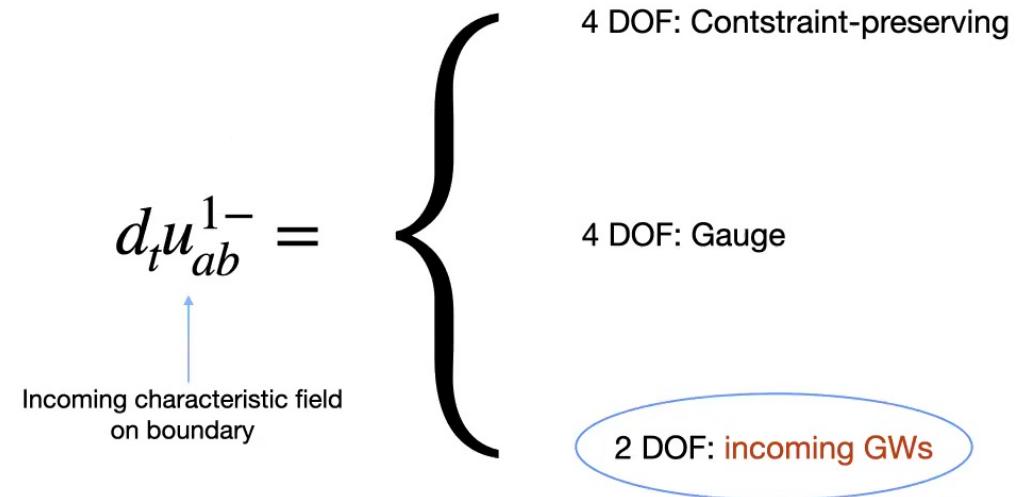




The Bjorhus boundary conditions

$$\partial_t u^\alpha + A_\beta^{k\alpha} \partial_k u^\beta = F^\alpha$$

Generalized Harmonic
(SpECTRE)





The performance of the fully relativistic 3D

- Spurious numerical reflections at the outer boundary are reduced
- The interface between the Cauchy and characteristic system is transparent to physical GWs

Ma et al., in preparation

Summary



1. The rational filter and the corresponding Bayesian framework
 - This novel framework is a powerful tool to do BH spectroscopy (theoretically and observationally)
 - A LIGO O4 pipeline?
2. The full filter; metric reconstruction; echo
3. Cauchy-characteristic matching

sma@caltech.edu