BAYESIAN METRIC RECONSTRUCTION WITH GRAVITATIONAL WAVE OBSERVATIONS

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OVERVIEW



Can one hear the shape of a black hole?



Can one hear the shape of a black hole?



BLACK HOLE RINGDOWN



 $FIGURE \ 1:$ Time evolution of a perturbation that scattered with a black hole as seen by an observer far away (left normal scale, right log scale).

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FIGURE 2: Left: Reconstructed final mass M_f and final spin a_f using different parts of the signal. Right: Reconstructed l = m = 2, n = 0 QNM frequency and damping time using different starting times as well as IMR result. Taken from B.P. Abbott et al. (LIGO Scientific and Virgo Collaborations), Phys. Rev. Lett. 116, 221101, 2016, https://doi.org/10.1103/PhysRevLett.116.221101, [1]

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FIGURE 3: Reconstructed final mass M_f and final spin magnitude χ_f . Left: using different overtone numbers to fit the ringdown, as well as IMR result. Right: as function of different starting times and overtone numbers, as well as IMR. Taken from Maximiliano Isi, Matthew Giesler, Will M. Farr, Mark A. Scheel, and Saul A. Teukolsky, Phys. Rev. Lett. 123, 111102, 2019, https://doi.org/10.1103/PhysRevLett.123.111102, [2]

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FIGURE 4: Left: Combined relative errors of the l = 2, n = 0 QNM for different combinations of $\{M, a_0, b_0\}$ and Schwarzschild M = 1 as reference. **Right:** Some of the corresponding perturbation potentials. Taken from SHV and Kostas D. Kokkotas, Phys. Rev. D 100, 044026 2019, https://doi.org/10.1103/PhysRevD.100.044026, [3].

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Methods

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- parametrized black hole space-time (Rezzolla-Zhidenko metric)
- perturbation equations ($\delta R_{\mu\nu} = 0$)
- computation of QNMs (higher order WKB method)
- RZ models, QNM subsets, QNM precision
- Bayesian analysis via Markov chain Monte Carlo (PYMC3)

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Blocks can be modified, general idea is the same!

BACKGROUND METRIC

We use the Rezzolla-Zhidenko (RZ) metric¹

- parametrization for spherically symmetric and static black holes
- continued fraction expansion for $\tilde{A}(x)$ and $\tilde{B}(x)$
- relation to PPN parameters β and γ possible

$$ds^{2} = -N^{2}(r)dt^{2} + \frac{B^{2}(r)}{N^{2}(r)}dr^{2} + r^{2}d\Omega^{2}, \qquad x \equiv 1 - \frac{r_{0}}{r}, \qquad N^{2} = xA(x), \quad (1)$$

¹Rezzolla and Zhidenko [4], Phys. Rev. D 90, 084009, 2014

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(1)

$$A(x) = 1 - \varepsilon (1 - x) + (a_0 - \varepsilon)(1 - x)^2 + \tilde{A}(x)(1 - x)^3,$$
(2)

$$B(x) = 1 + b_0(1-x) + \tilde{B}(x)(1-x)^2.$$
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$$\varepsilon = -\left(1 - \frac{2M}{r_0}\right), \qquad a_0 = \frac{(\beta - \gamma)(1 + \varepsilon)^2}{2}, \qquad b_0 = \frac{(\gamma - 1)(1 + \varepsilon)}{2}.$$
 (4)

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Methods

PERTURBATION EQUATIONS

We study "theory agnostic" gravitational axial perturbations

- we consider $\delta R_{\mu\nu} = 0$ for the RZ metric
- · corresponds to GR, but also holds for some scalar tensor theories

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$$\frac{\mathrm{d}^2}{\mathrm{d}r^{*2}}Z + \left[\omega^2 - V_l(r)\right]Z = 0, \qquad (5)$$

• also include parametrized modification of the potential (K)

$$V_l(r) = \frac{l(l+1)}{r^2} N^2(r) - \frac{K}{r} \frac{\mathsf{d}}{\mathsf{d}r^*} \frac{N^2(r)}{B(r)},\tag{6}$$

QUASI-NORMAL MODES / WKB

Quasi-Normal Modes (QNMs) describe ringdown of black holes

- defined by purely outgoing $(r \rightarrow \infty)$ and ingoing $(r \rightarrow r_0)$ waves
- computation of QNMs via higher order WKB method

$$\frac{iQ_0}{\sqrt{2Q_0''}} - \Lambda_2 - \Lambda_3 - \Lambda_4 - \Lambda_5 - \Lambda_6 = n + \frac{1}{2},$$
(7)

with $Q(r^*) \equiv \omega_n^2 - V_l(r^*)$ evaluated at the maximum of potential².

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$$\Lambda_2(n) = \left((-11(V^{(3)})^2 + 9V^{(2)}V^{(4)} - (30(V^{(3)})^2)n + \right)$$
(8)

$$18V^{(2)}V^{(4)}n - (30(V^{(3)})^2)n^2 + 18V^{(2)}V^{(4)}n^2) \Big) / 144(V^{(2)})^2$$
(9)

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RZ MODELS AND QNM SPECTRA

Choose finite number of RZ parameters and available QNMs

RZ models

- model₁ $\equiv \{M, \varepsilon\}$
- $model_2 \equiv \{M, \varepsilon, a_0, b_0\}$
- model₃ $\equiv \{M, \varepsilon, a_1, b_1\}$
- $model_{K1} \equiv \{M, \varepsilon, K\}$
- $model_{K2} \equiv \{M, \varepsilon, a_0, b_0, K\}$

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QNM spectra

- spectrum₁ $\equiv \{l = 2, n = [0, 1]\}$
- spectrum₂ $\equiv \{l = [2, 3], n = [0, 1]\}$

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QNM spectra

- spectrum₁ $\equiv \{l = 2, n = [0, 1]\}$
- spectrum₂ $\equiv \{l = [2, 3], n = [0, 1]\}$

We also consider two different precision for both spectra

• relative errors for real and imaginary part of $\pm 10\%$ or $\pm 1\%$.

Methods

BAYESIAN ANALYSIS

Bayes theorem:

$$P(\theta|D) = \frac{P(D|\theta)P(\theta)}{P(D)}$$
(10)

with

• θ parameters of a model

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• D observed data

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with

- θ parameters of a model
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and

• **posterior** $P(\theta|D)$: probability of parameters given the data

J

- likelihood $P(D|\theta)$: probability of data given the parameters
- prior $P(\theta)$: probability of parameters before looking at data
- evidence P(D): probability of Data

Methods

MARKOV CHAIN MONTE CARLO

Markov chain Monte Carlo (MCMC) are a class of algorithms

- directly sample from the posterior distribution
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We use Python based framework PYMC3

- combine it with external C++ code
- use standard Metropolis Hastings algorithm

Methods



Results

RESULTS



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MODEL₁ WITH SPECTRUM₁ AT 1%



Results for model₁ obtained by using spectrum₁ with $\pm 1\%$ relative error. Left: MCMC parameter estimation. Right top: Exact (black lines) and reconstructed (color lines) potentials $V_2(r)$ and $V_3(r)$. Right bottom: Exact (black lines) and reconstructed (color lines) metric functions $g_{tt}(r)$ and $g_{rr}(r)$.

MODEL₂ WITH SPECTRUM₂ AT 1%



FIGURE 5: Results for model₂ obtained by using spectrum₂ with ± 1 % relative error. Left: MCMC parameter estimation. Right top: Exact (black lines) and reconstructed (color lines) potentials $V_2(r)$ and $V_3(r)$. Right bottom: Exact (black lines) and reconstructed (color lines) metric functions $g_{tt}(r)$ and $g_{rr}(r)$.

MODEL₃ WITH SPECTRUM₂ AT 1%



FIGURE 6: Results for model₃ obtained by using spectrum₂ with ± 1 % relative error. Left: MCMC parameter estimation. Right top: Exact (black lines) and reconstructed (color lines) potentials $V_2(r)$ and $V_3(r)$. Right bottom: Exact (black lines) and reconstructed (color lines) metric functions $g_{tt}(r)$ and $g_{rr}(r)$.

MODEL_{K1} WITH SPECTRUM₁ AT 1%



FIGURE 7: Results for model_{K1} obtained by using spectrum₁ with $\pm 1\%$ relative error. Left: MCMC parameter estimation. Right top: Exact (black lines) and reconstructed (color lines) potentials $V_2(r)$ and $V_3(r)$. Right bottom: Exact (black lines) and reconstructed (color lines) metric functions $g_{tt}(r)$ and $g_{rr}(r)$.

$MODEL_{K2}$ WITH SPECTRUM₂ AT 1%



FIGURE 8: Results for model_{k2} obtained by using spectrum₂ with $\pm 1\%$ relative error. Left: MCMC parameter estimation. Right top: Exact (black lines) and reconstructed (color lines) potentials $V_2(r)$ and $V_3(r)$. Right bottom: Exact (black lines) and reconstructed (color lines) metric functions $g_{tt}(r)$ and $g_{rr}(r)$.

OVERALL FINDINGS

General observation for models, spectra and precision

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General observation for models, spectra and precision

- low dimensional models work well with spectrum₁ (l = 2, n = [0, 1])
- higher dimensional models need spectrum₂ (l = [2, 3], n = [0, 1])
- benefits from 1 % precise QNMs outperform additional *l* = 3 QNMs
- · modified perturbation potential decreases reconstruction mildly
- · sampled potentials are suitable for WKB treatment

RELATION TO EXOTIC COMPACT OBJECTS



Left: Axial perturbations ultra compact stars, V. Ferrari and K. D. Kokkotas, Phys. Rev. D 62, 107504, 2000.

Right: Echoes from the abyss: Tentative evidence for Planck-scale structure at black hole horizons, Abedi, Dykaar and Afshordi, Phys. Rev. D 96, 082004 2017.

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RELATION TO EXOTIC COMPACT OBJECTS

- · Inverse problem for horizonless ultra compact objects is different
- WKB method and Bohr-Sommerfeld rules³ are powerful here (approximate, but easier to invert)

³here
$$E_n \equiv \omega_n^2$$

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$$\int_{x_0}^{x_1} \sqrt{E_n - V(x)} dx = \pi \left(n + \frac{1}{2} \right) - \frac{i}{4} \exp\left(2i \int_{x_1}^{x_2} \sqrt{E_n - V(x)} dx \right)$$
(11)

³here
$$E_n \equiv \omega_n^2$$

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INVERTING BOHR-SOMMERFELD RULES

- Known for single wells or single barriers (classical BS rule)⁴
- Extended to quasi-stationary states (3 or 4 turning points) ⁵
- Neutron star potentials with discontinuity (1 turning point)⁶

$$\mathcal{L}_{1}(E) = x_{1} - x_{0} = 2\frac{\partial}{\partial E} \int_{E_{\min}}^{E} \frac{n(E') + 1/2}{\sqrt{E - E'}} dE'$$
(12)

$$\mathcal{L}_{2}(E) = x_{2} - x_{1} = -\frac{1}{\pi} \int_{E}^{E_{\text{max}}} \frac{(\mathbf{d}T(E')/\mathbf{d}E')}{T(E')\sqrt{E'-E}} \mathbf{d}E'$$
(13)

⁴Wheeler [6]; Cole and Good [7]
 ⁵Völkel and Kokkotas [8]; Völkel [9]; Völkel and Kokkotas [10]
 ⁶Völkel and Kokkotas [11]

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INVERTING BOHR-SOMMERFELD RULES



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Outlook

Several extensions planned:

- inverse problem for non GR QNMs
- · address the rotating case
- combine with other approaches (BH shadows)



As a proof of principle addressed in this work

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- Bayesian framework to construct BH metrics from QNMs via MCMC
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Black hole spectroscopy can be used for inverse spectrum problems!

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$$\tilde{A}(x) = \frac{a_1}{1 + \frac{a_2 x}{1 + \frac{a_3 x}{1 + \frac{a_3 x}{1 + \dots}}},$$

$$\tilde{B}(x) = \frac{b_1}{1 + \frac{b_2 x}{1 + \frac{b_3 x}{1 + \frac{b_3 x}{1 + \dots}}}.$$
(15)

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