Black hole astrophysics with gravitational-wave catalogs

IPAM Workshop: Source Inference and Parameter Estimation in Gravitational Wave Astronomy November 16 2021



CENTER FOR INTERDISCIPLINARY EXPLORATION AND RESEARCH IN ASTROPHYSICS

Image Credit: Aurore Simonnet/LIGO-Caltech-MIT-Sonoma State

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Observing Binary Black Holes

How big is each black hole?



Where and when did they merge?



How fast are they spinning? Where are the spin axes pointing?





How are black holes made?





The origins of LIGO-Virgo's black holes

- What were their progenitors? (Likely massive stars)
- Where and when did these massive stars live?
- How did these stars die?
- How did the stellar deaths affect their environments?
- How did these stellar remnants pair up into merger partners?

All of these pieces affect the masses, spins, and merger rates of gravitational-wave events

See Michela's talk



Measuring the parameters of gravitational-wave events

 $\mathcal{M}[M_{\odot}]$ 100 0.02 50 ♦

GW191103_012549 GW191105_143521 GW191109_010717 GW191113_071753 GW191126_115259 GW191127_050227 GW191129_134029 GW191204_110529 GW191204_171526 GW191215_223052 GW191216_213338

Chirp mass





Parameter estimation

Prior Likelihood Posterior $p(\text{data}_i \mid \theta_i) p_0(\theta_i \mid \mathscr{H})$ $p(\theta_i \mid \text{data}_i) =$ $p(\text{data}_i \mid \mathcal{H})$ Evidence

θ_i : event i's parameters, like masses, spins

Parameter estimation assumes a default prior. Population inference finds the "best" prior, common to all systems



From Single Events to a Population

- Introduce a population model \mathcal{H} with a set of population hyperparameters that describe the distributions of masses, spins, redshifts across multiple events
- Example: Fit a power law to black hole masses. Hyper-parameters: power-law slope, minimum black hole mass, maximum black hole mass.
- Take into account measurement uncertainty and selection effects



Population analysis Find the "best" prior to use for individual events $p_{\text{pop}}(\theta \mid \lambda, \mathcal{H})$ Parameter estimation likelihood for event *i* $p(\text{data} \mid \lambda, \mathcal{H}) =$ Likelihood given i population model and hyperparameters

See, e.g., Gair+ 2019, Thrane & Talbot 2019, Vitale+ 2020 for derivation

- $\int p(\text{data}_i \mid \theta_i) p_{\text{pop}}(\theta_i \mid \lambda, \mathcal{H}) d\theta$

$$\beta(\lambda, \mathcal{H})$$

Selection effects: fraction of detectable systems in the population



Example of selection effects: Big black holes are louder than small black holes





Population inference also updates our knowledge of the parameters of individual events



Default parameter estimation prior is arbitrary! The likelihood is the same, but different priors different posteriors

Default priors are flat in (detectorframe) component masses.

Population prior is the result of fitting the *population model* to all events





Three Astrophysical Lessons

- Mass distribution 1.
- Spin distribution 2.
- **Evolution with redshift** 3.



Three Astrophysical Lessons

- Mass distribution: transition between neutron stars and black holes
- Spin distribution 2.
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Masses in the Stellar Graveyard

LIGO-Virgo-KAGRA Black Holes LIGO-Virgo-KAGRA Neutron Stars EM Black Holes EM Neutron Stars



LIGO-Virgo-KAGRA | Aaron Geller | Northwestern



Can we distinguish neutron stars and black holes based on mass? **GWTC-1 suggested YES**



MF, Essick, & Holz 2020 ApJL 899 L8

New O3 events are not really filling in the "gap"...



Farah, MF, Essick, & Holz 2021 arXiv:2111.03498



There is a clear deviation from a power law at ~2.4 solar masses **Coincides with neutron star maximum mass?**







Three Astrophysical Lessons

- **Mass distribution** 1.
 - Spin distribution: slowly spinning black holes and large spin tilts
- **Evolution with redshift** 3.



Effective spin parameters

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- The gravitational-wave signal can be parameterized by two "effective" spins:
 - The effective inspiral spin measures the total spin along the orbital angular momentum axis,

$$\chi_{\text{eff}} = \frac{m_1 \chi_1 cos\theta_1 + m_2 \chi_2 \cos\theta_2}{m1 + m2}$$

• The effective precessing spin measures the spin in the orbital plane, perpendicular to orbital angular momentum axis $\chi_p \propto \chi_1 \sin \theta_1$





- $\chi_{\rm eff} < 0$ implies there are spin tilts > 90 degrees
- Current distribution consistent with symmetric χ_{eff} centered at zero, implies the distribution of spin tilts may be isotropic
- Favors dynamical origin or mixture between dynamical and isolated(?)

LVK arXiv:2111.03634, methods papers by Miller, Callister & Farr 2020, Roulet & Zaldarriaga 2019





Black holes in gravitational-wave systems spin slower than those in X-ray binaries



Fishbach & Kalogera 2021, arXiv:2111.02935



Three Astrophysical Lessons

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- Mass distribution 1.
- Spin distribution 2.

Evolution with redshift: Merger rate evolution matches star formation rate



Binary black hole merger rate density across cosmic time



LVK arXiv:2111.03634, methods paper by MF+ 2018, gwpopulation by Talbot+





Comparing merger rate evolution to star formation + time delays



MF & Kalogera 2021, ApJL 914 L30





Data favor short time delays between formation and merger



MF & Kalogera 2021, ApJL 914 L30



Predicted time delay distributions









Ongoing and future population inference with larger catalogs

- Model checking and outlier tests
- Towards nonparametric population models
- spin correlation with masses)

• Population-level correlations between parameters (e.g. mass evolution with redshift,





Ongoing and future population inference with larger catalogs

- Model checking and outlier tests: when do we have to upgrade our simple models?
- Towards nonparametric population models
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Posterior Predictive Checks "Goodness of fit:" Unlike with model selection, we don't have to specify an alternative model. But there is also not a single statistic



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MF, Doctor, Callister, Edelman, Ye, Essick, Farr, Farr & Holz 2021 ApJ 912 98

we can use to reject a model.



Quantify deviations between predicted and observed events



MF, Doctor, Callister, Edelman, Ye, Essick, Farr, Farr & Holz 2021 ApJ 912 98



Outlier tests with coarse-grained likelihoods

- Typical outlier tests in gravitational-wave population inference do a "leave-one-out" analysis
- This usually involves selecting a potential "outlier" event in advance, which leads to bias
- One solution is to repeat "leave-one-out" for every event in the catalog —> computationally expensive
- Another solution: define a "coarse-grained" likelihood
- Also see Roulet+ Phys. Rev. D 104, 083010 (2021)





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Semiparametric population fits in GWTC-3



model deviations from parameterized model as spline

LVK arXiv:2111.03634, methods papers by Edelman+ 2021, Tiwari+ 2020, Mandel+ 2017

BPG: Binned Gaussian Process on 2D mass plane, FM: flexible mixture model, PS: See Bruce's poster



Ongoing and future population inference with larger catalogs

- Model checking and outlier tests
- Towards nonparametric population models

Population-level correlations between parameters (e.g. mass evolution with redshift, spin correlation with masses)



Examples of population-level correlations





- Astrophysical Lessons

 - Mass distribution: Transition between neutron stars and black holes • Spin distribution: Slowly spinning black holes and large spin tilts • Evolution with redshift: Merger rate evolution and small time delays
- Future prospects and challenges
 - Model checking and outlier tests: posterior predictive checks, coarse-grained likelihoods

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- Nonparametric population models: Gaussian mixtures, Gaussian process regularized histograms, splines
- Look for correlations between population distributions

Summary



Future prospects and challenges, continued

- Simultaneously model signal and noise population
- How do we interpret the population fits in the context of theoretical models? (See Michela's talk)
- Include cosmology, theory of gravity, neutron star equation of state in population model (see Rachel's talk)
- Leverage information from stochastic background (see Arianna's talk) • Turn systematic uncertainties into statistical uncertainties (e.g. waveform models,

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- calibration)
- Smarter parameter estimation, sensitivity estimation

