

Gravitational Waves

Frank Ohme

Max Planck Institute for Gravitational Physics
Albert Einstein Institute
Hannover, Germany

Astroparticle School September 2025

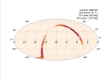
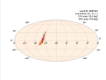
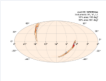
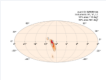


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FOR GRAVITATIONAL PHYSICS
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Public Alerts

SORT: EVENT ID (A-Z) ▾

Event ID	Possible Source (Probability)	Significant	UTC	GCN	Location	FAR
S250904cv	BBH (>99%)	Yes	Sept. 4, 2025 13:49:52 UTC	GCN Circular Query Notices VOE		1 per 100.04 years
S250904br	BBH (>99%)	Yes	Sept. 4, 2025 10:22:08 UTC	GCN Circular Query Notices VOE		1 per 100.04 years
S250904ae	BBH (>99%)	Yes	Sept. 4, 2025 03:33:07 UTC	GCN Circular Query Notices VOE		1 per 100.04 years
S250901cb	BBH (>99%)	Yes	Sept. 1, 2025 18:59:41 UTC	GCN Circular Query Notices VOE		1 per 9.9253e+05 years

Topics



1. Interpreting the Signal

- The Inverse Problem
- Waveform Models
- Bayesian Parameter Estimation
- Stochastic Sampling

2. Gravitational-Wave Observations

- Observing Runs

■ Masses and Spins

■ Stellar Graveyard

■ Select Highlights

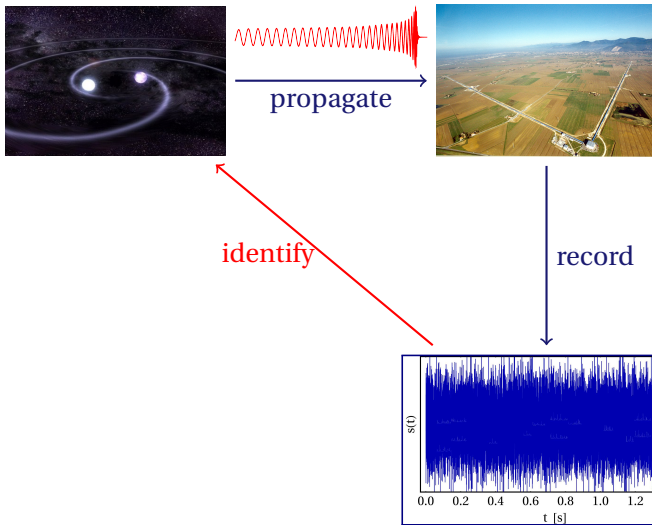
3. The Future

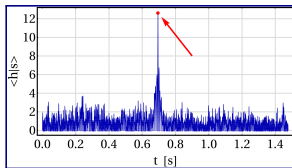
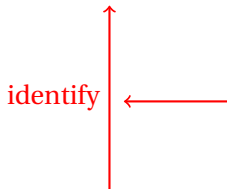
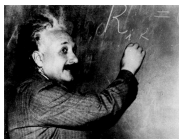
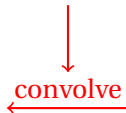
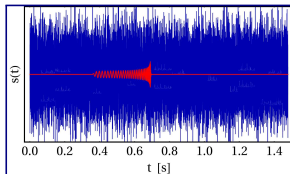
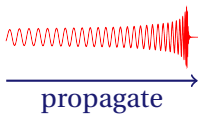
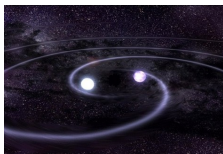
■ LISA

■ Einstein Telescope

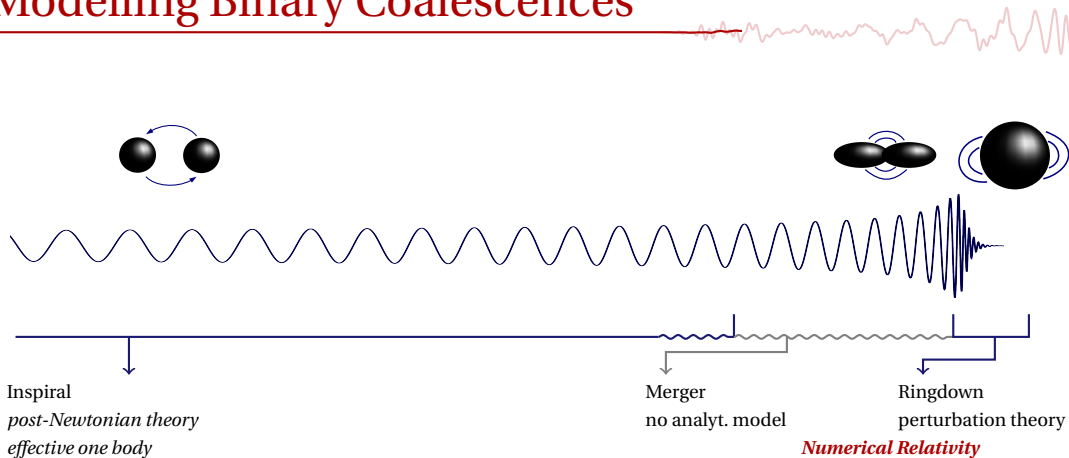
■ Conclusions

Interpreting the Signal





Modelling Binary Coalescences

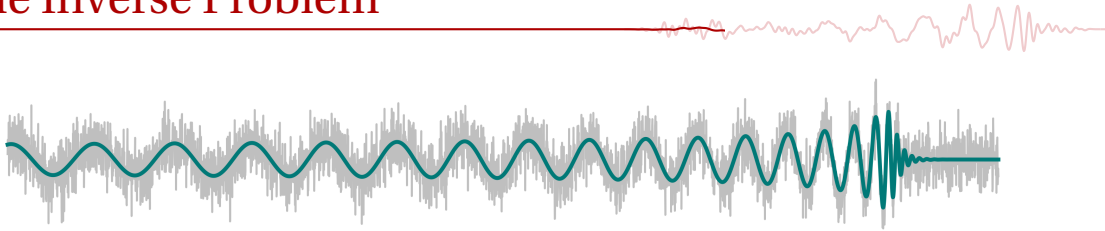


[FO, CQG 29 124002 (2012)]

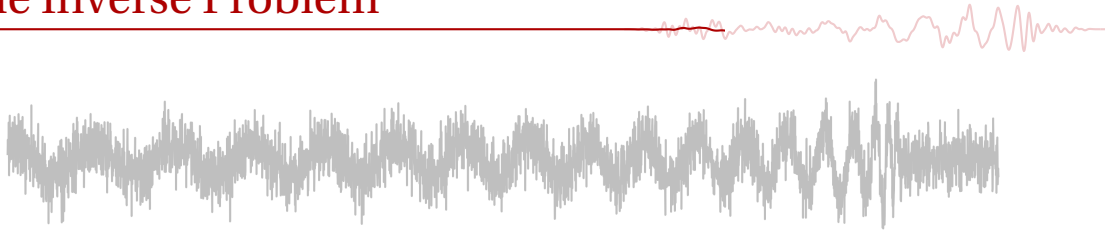
The Inverse Problem



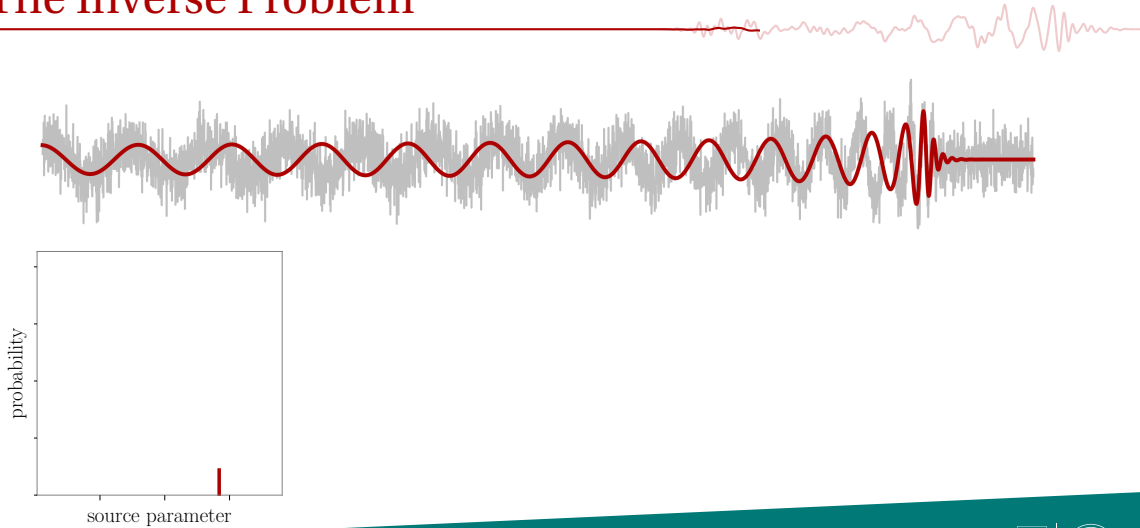
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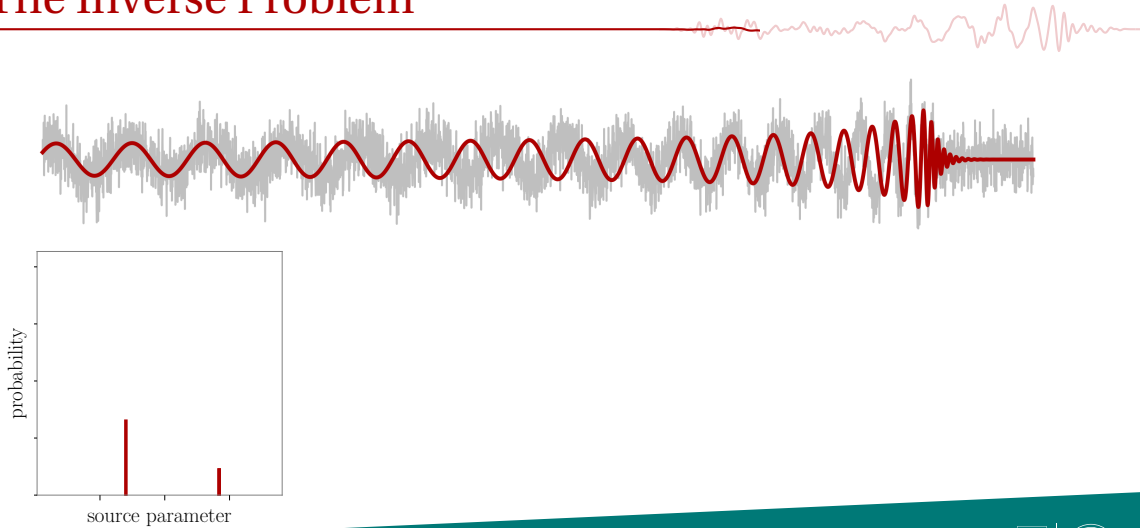
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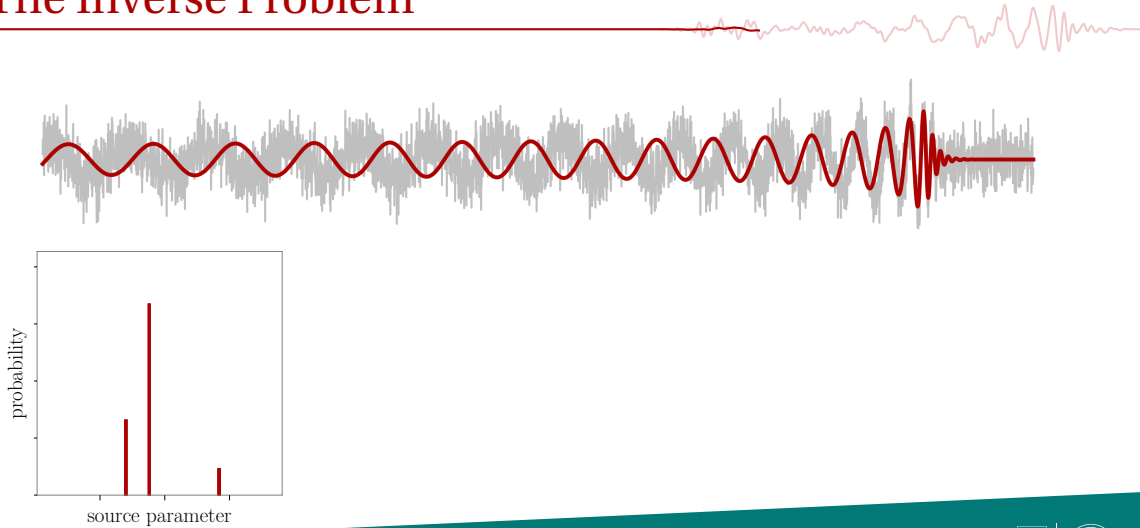
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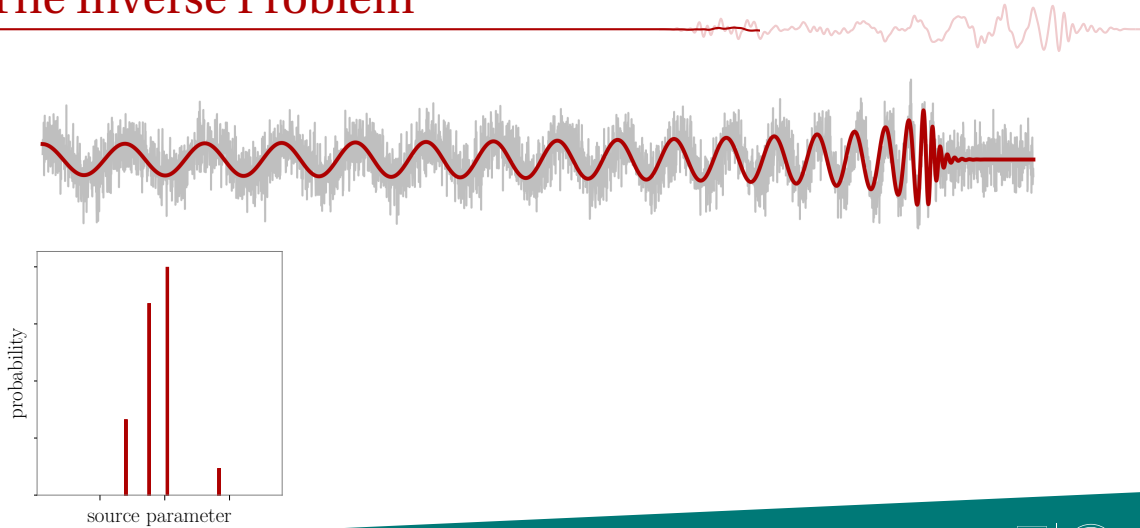
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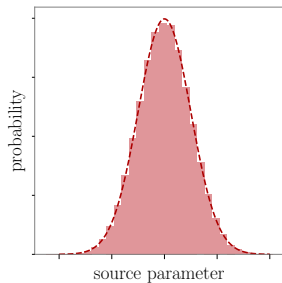
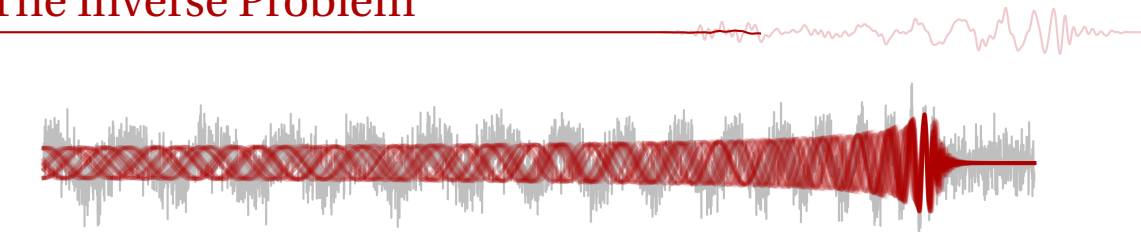
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The Inverse Problem



Likelihood & Posterior

Quantities d : data, h : GW model, θ : source parameters

Inner Product $\langle d, h \rangle = 4 \operatorname{Re} \int \frac{\tilde{d}(f) \tilde{h}^*(f)}{S_n(f)} df$

Likelihood $\Lambda(d|h(\theta)) \sim \exp(-\|d - h(\theta)\|^2/2)$

Posterior $p(\theta|d) = \frac{\Lambda(d|\theta)\pi(\theta)}{p(d)}$

Stochastic sampling

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Stochastic sampling

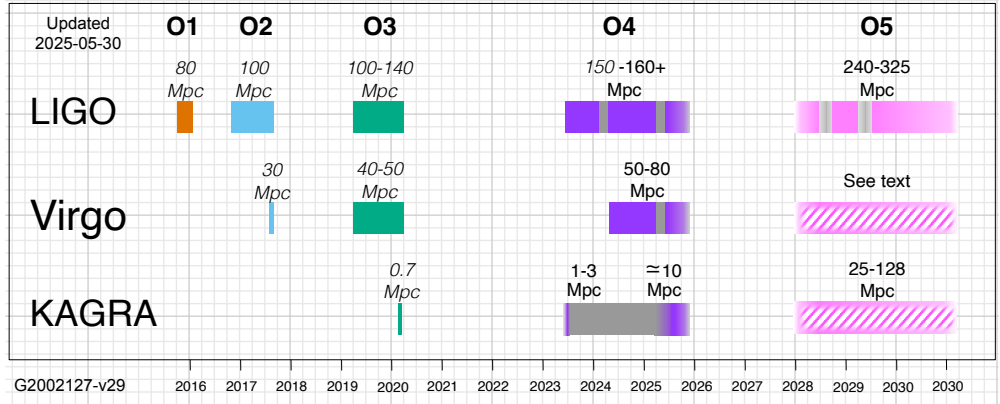
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- ▶ Integrals become simple operations on set Θ , e.g.,

$$E[\theta_1] = \int \theta_1 p(\theta|s, I) d\vec{\theta} \approx \text{median}(\theta_1 \in \Theta)$$

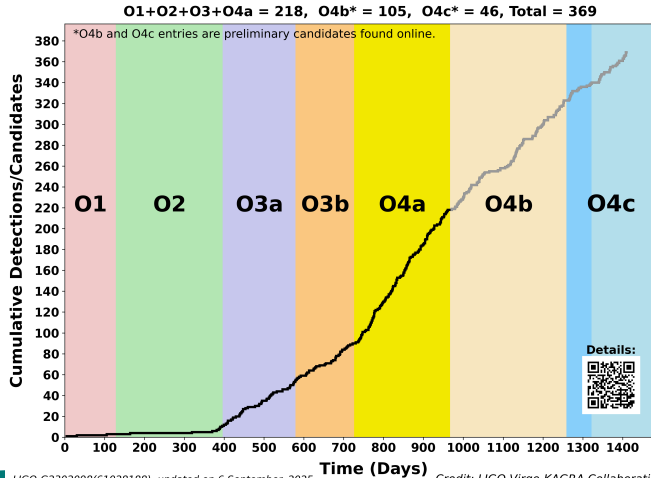
Gravitational-Wave Observations

Timeline

[<https://observing.docs.ligo.org/plan/>]



Observations



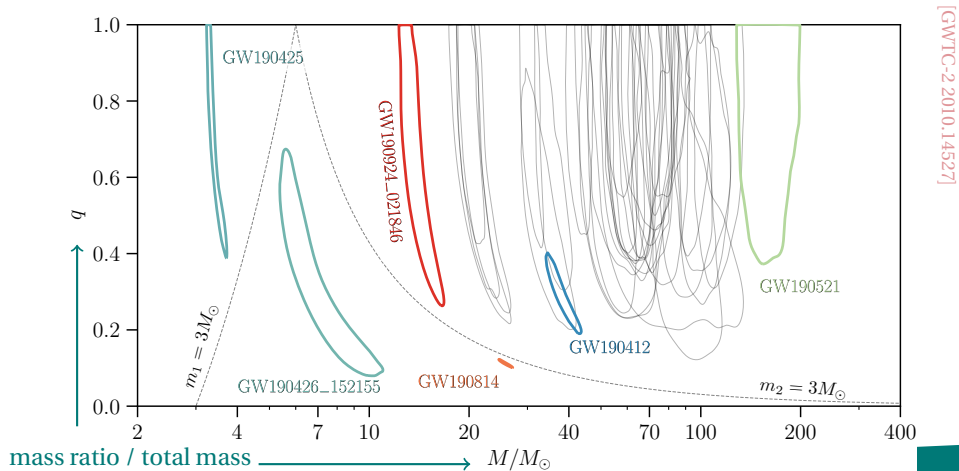
LIGO-G2302098(61028188), updated on 6 September, 2025

Credit: LIGO-Virgo-KAGRA Collaboration

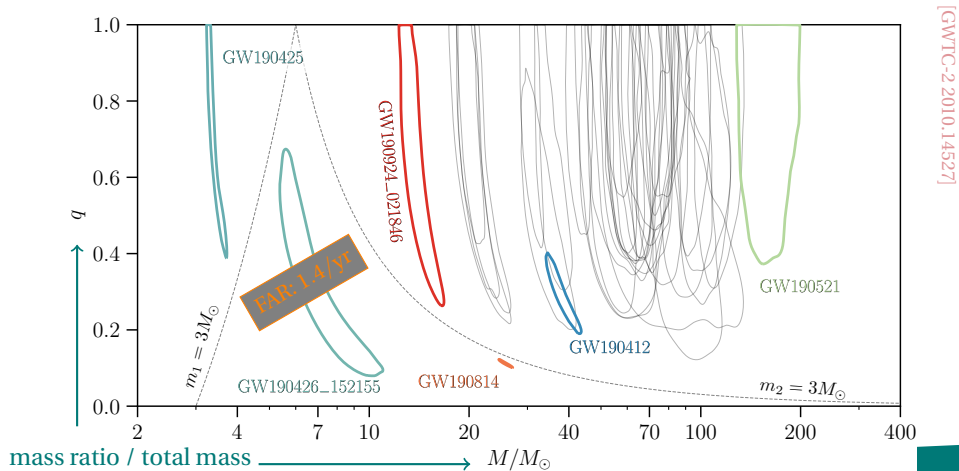
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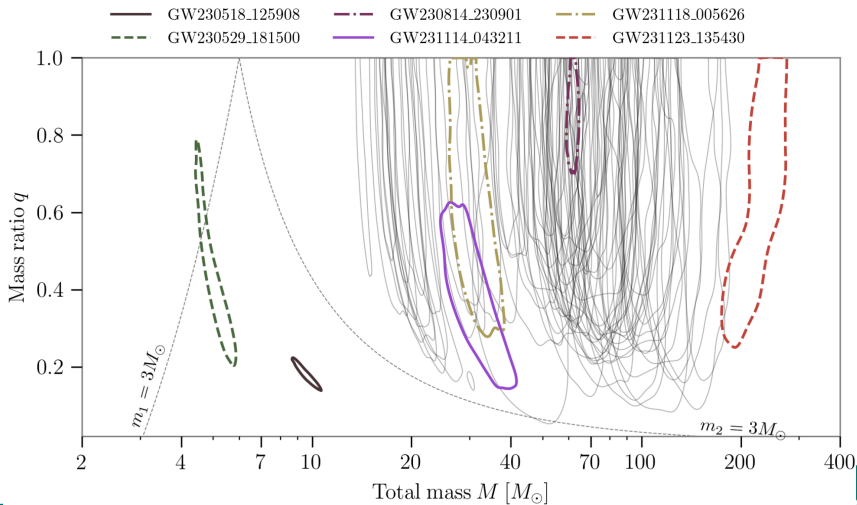
Mass Posteriors



Mass Posteriors

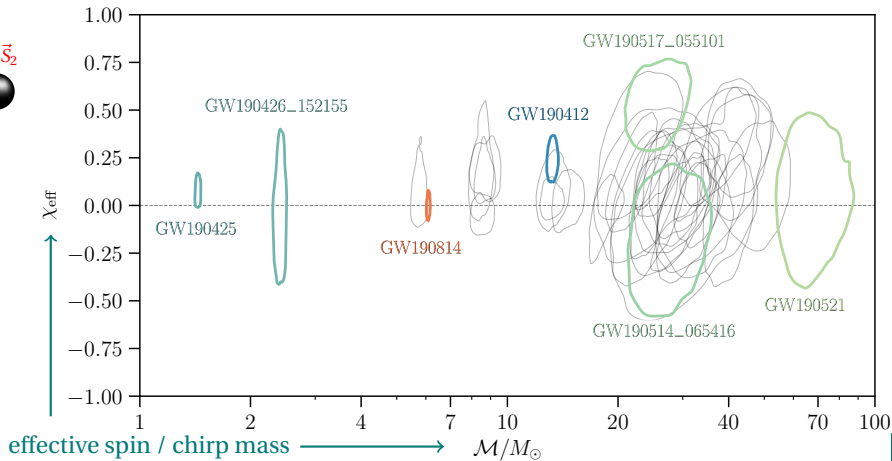
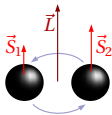


Mass Posteriors



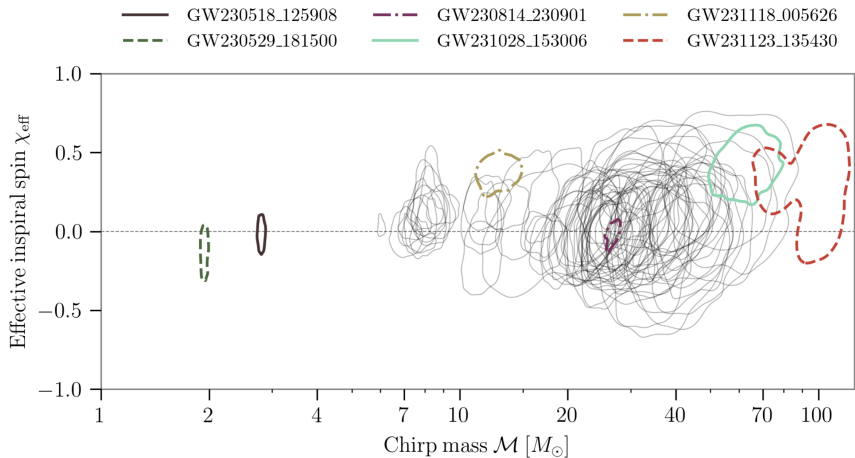
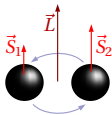
[GWTC-4 2508.18082]

Spins



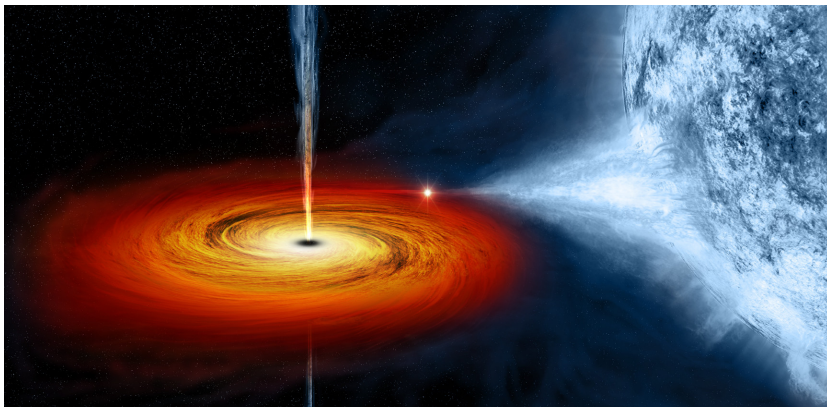
[GWTC-2 2010.14527]

Spins



[GWTC-4 2508.18082]

Binary Evolution

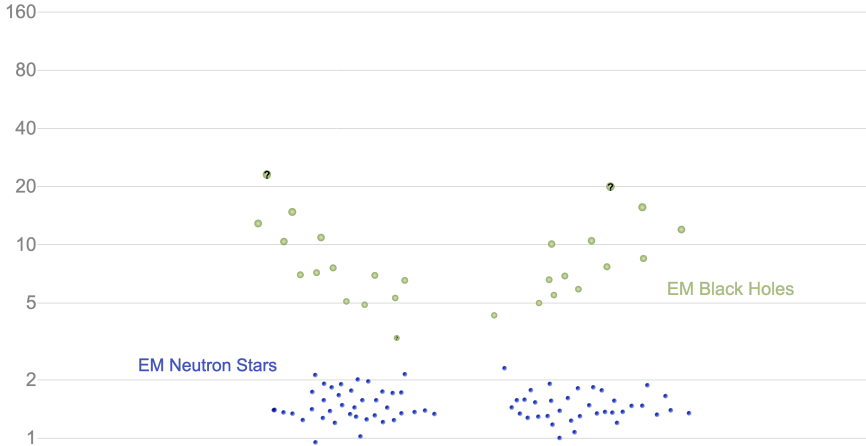


Material to Download



<https://s.gwdg.de/th635p>

Masses in the Stellar Graveyard



GWTC-2 plot v1.0

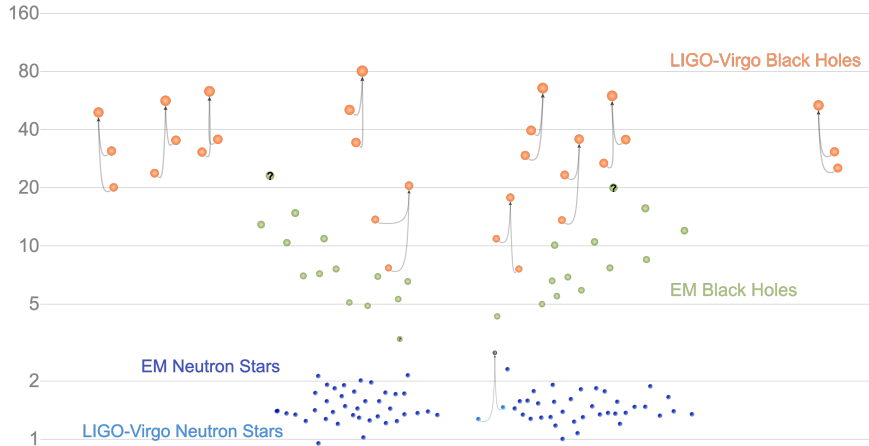
LIGO-Virgo | Frank Elavsky, Aaron Geller | Northwestern

15/24

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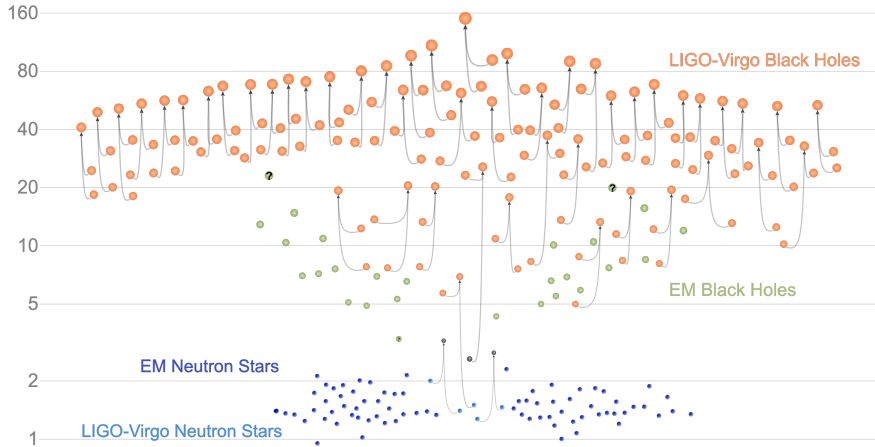
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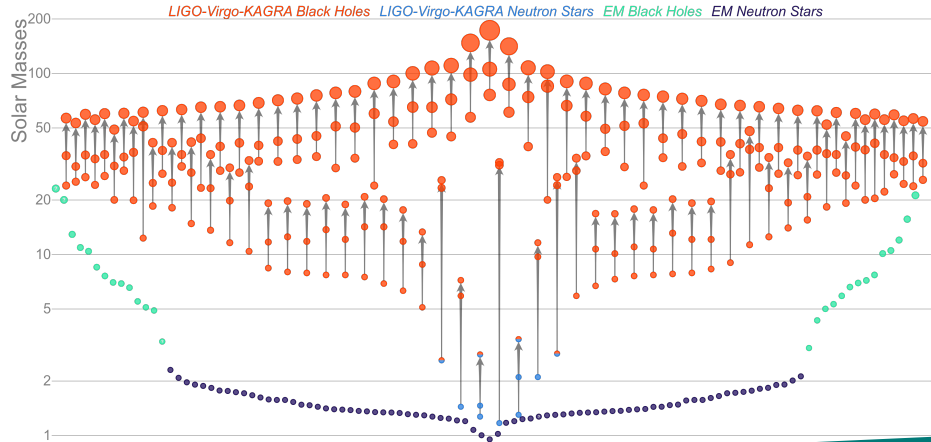
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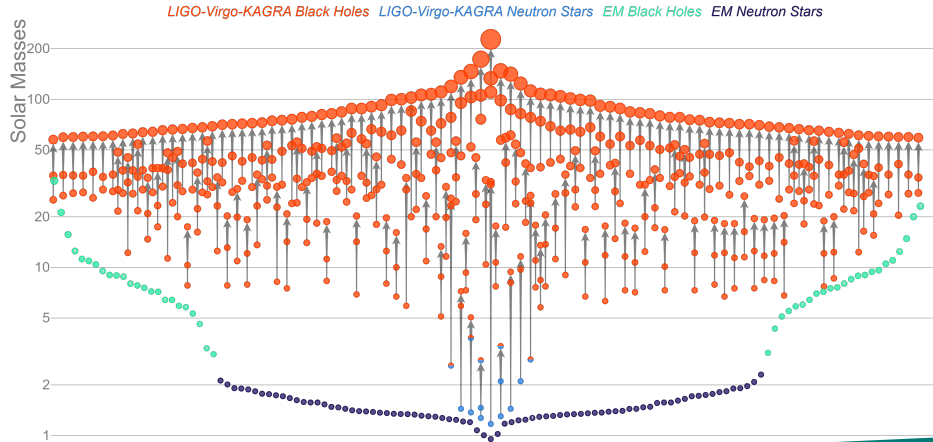


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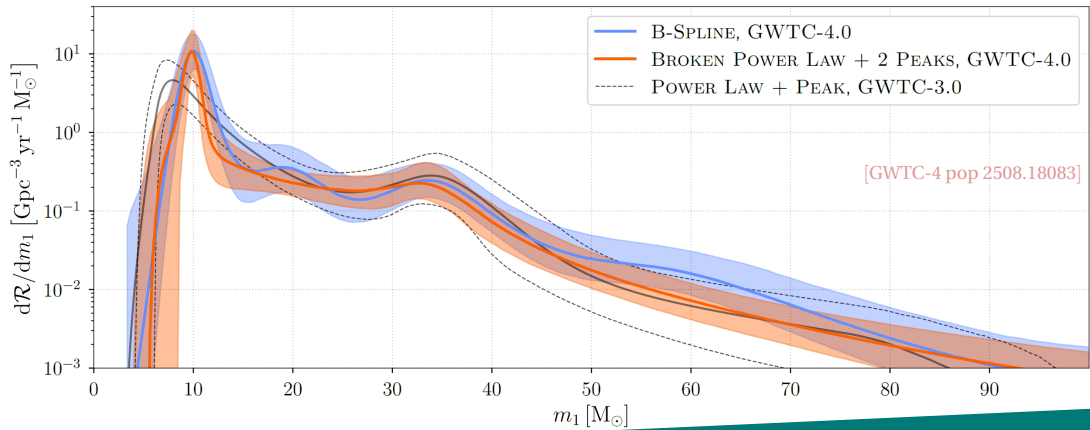
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Masses in the Stellar Graveyard



Population Properties



My Personal Highlights



GW150914 The first GW signal. Loud, but now considered typical.

GW170817 The first (and only) GW-multi-messenger event. Merger of two neutron stars.
Largest coordinated global observation campaign across the EM spectrum.

GW190412 Black holes of unequal masses. First higher harmonics.

GW190521 Black holes with masses that shouldn't have formed in a supernova. Where did they come from?

GW190814 Very unequal masses. The less massive object is a ...?

GW200129 Precessing black holes and potentially a remnant with a kick.

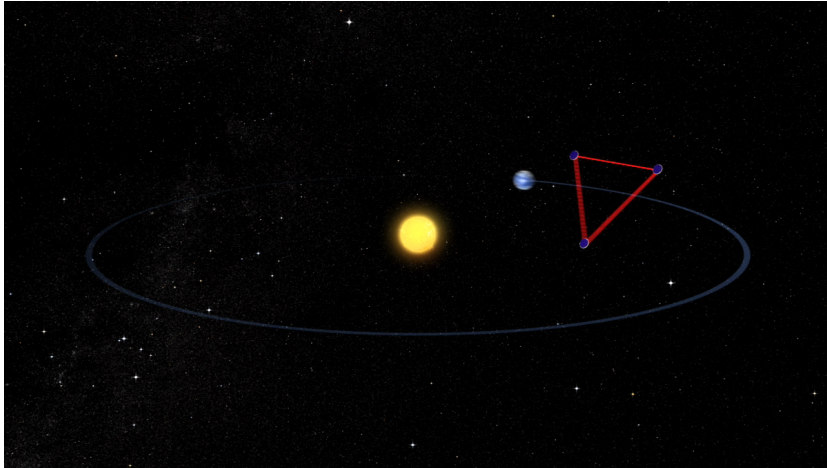
GW200115_042309 A black holes swallowing (or disrupting) a neutron star.

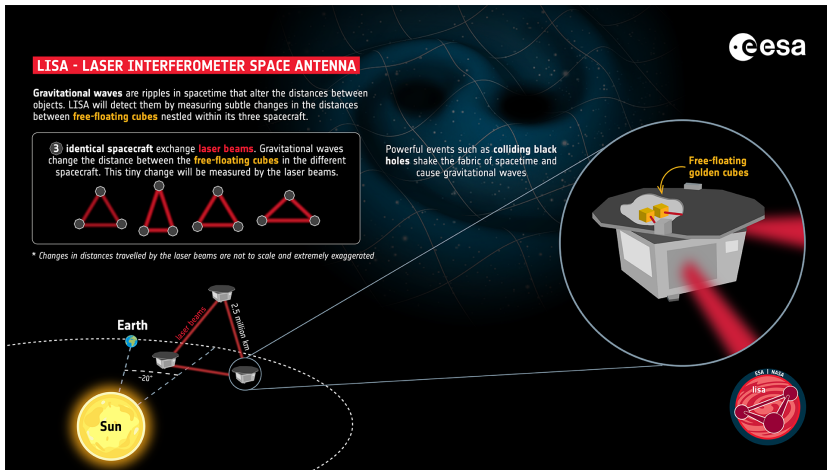
GW25xxxx



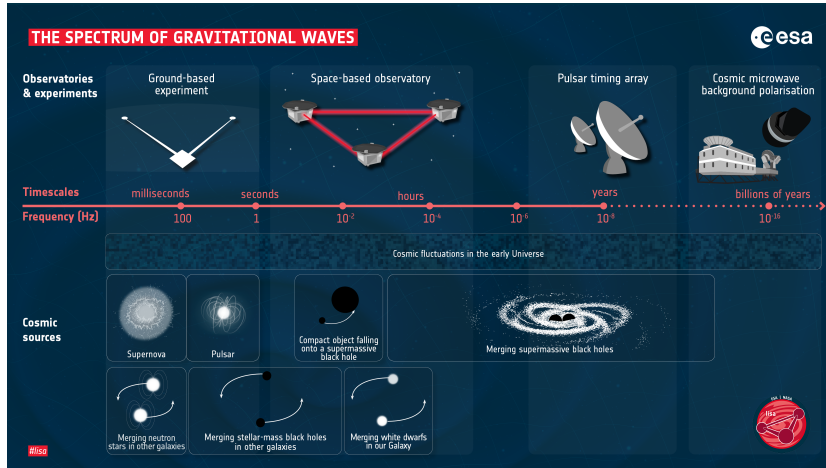
The Future

LISA

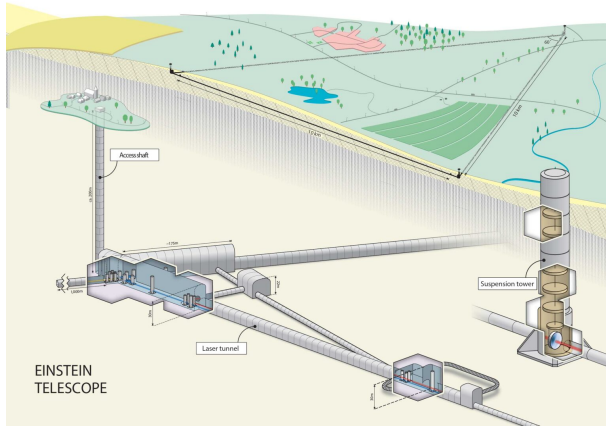




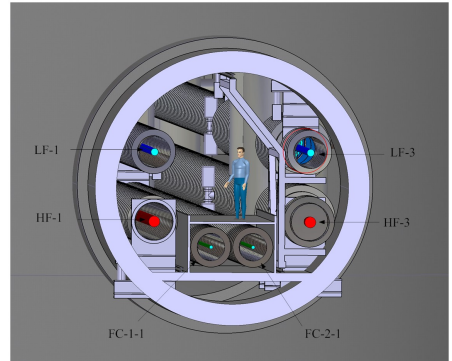
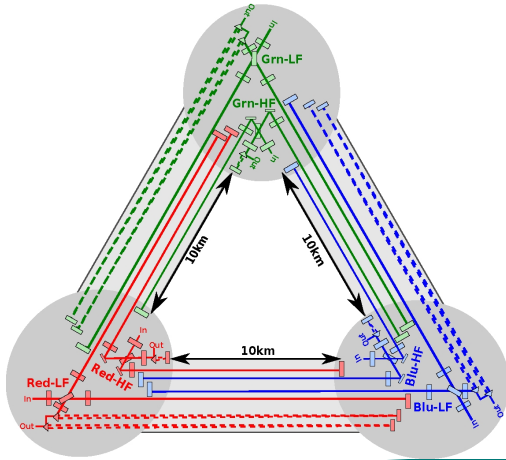
The Gravitational-Wave Spectrum



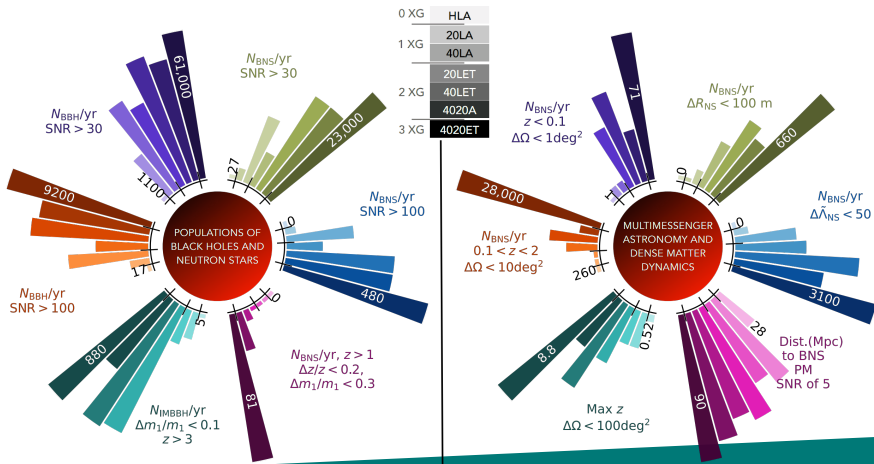
Einstein Telescope



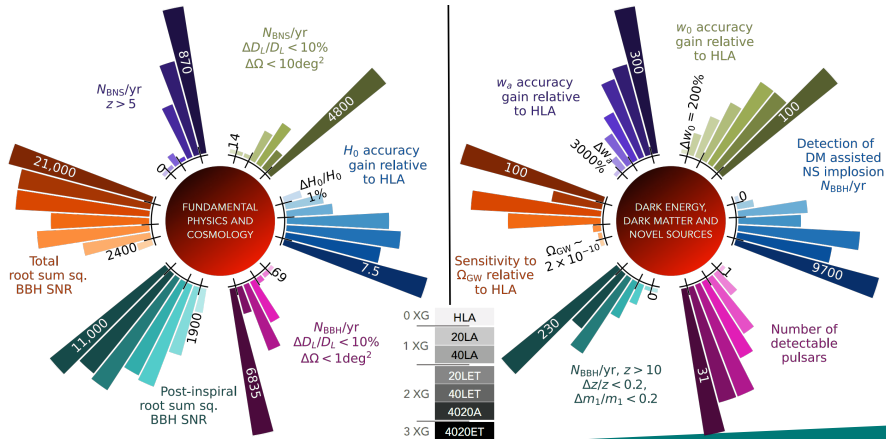
Einstein Telescope



Science Prospects



Science Prospects



Conclusions



- ▶ Gravitational-wave astronomy is reality
- ▶ The Universe is full of black holes with (surprisingly) diverse properties
→ diverse formation channels?
- ▶ More observations to come and needed to unravel big questions
(formation, environment, evolution, expansion, early Universe, ...)
- ▶ So far, no signs of Einstein's theory failing
- ▶ Continued work needed on possible systematics in the models and analyses
- ▶ Exciting future: LIGO Indida, LISA, Einstein Telescope
- ▶ Multimessenger astronomy will become the norm, not the exception