

#### A new beginning for transient Gravitational-wave astrophysics

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LIGO Scientific Collaboration



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#### Gravitational-wave astrophysics

Fundamentally new way to learn about the Universe:

- Is General Relativity in the correct theory of Gravity?
- What happens when matter is compressed to nuclear densities?
- What are the properties of the population(s) of compact objects?
- Is the mechanism that generates gamma-ray bursts a compact binary coalescence?

#### The Gravitational Wave Spectrum





• Before Einstein: Newtonian gravity



 1915: Einstein's General Relativity, gravitation due to spacetime curvature



 1916: Albert Einstein predicts the existence of gravitational waves

 The wave travels at the speed of light, is transverse, and has two polarisations:



Weak coupling with matter

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- High-precision length measurement: Laser Interferometers
- Dense masses moving fast: merging compact objects

#### ~30 years ago: Laser Interferometer Gravitational-wave Observatory

- Two sites 10 light-milliseconds apart
- Measurement of **space-time** deformations with  $\Delta L/L$ : ~10<sup>-21</sup> !





### Overview (or how can we study transient GWs?)

- Introduction
  - Compact Binary Coalescence
  - · LIGO
- Extracting astrophysics
  - Waveform models
  - Parameter Estimation
- Beyond aLIGO first observing run:
  - Astrophysics with multiple events



#### **Compact Binary Coalescence**

 Intrinsic parameters: primary and secondary masses and spins (and eccentricity)





 Extrinsic: time, sky-position, distance, orientation, reference phase

# LIGO measurement technique



#### Parameter Inference: GW150914 observation



How do we extract the scientific content?

#### Gravitational waveform models

- 2 models of the signal as a proxy for systematic errors:
  - Double-aligned-spin model (SEOBNRv2\_ROM, [Taracchini, et al., 2014; Pürrer, 2014])
  - Single-precessing-spin model (IMRPhenomPv2, [Hannam et al. Phys. 2014])



#### Gravitational waveform models

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#### Masses from the inspiral and ringdown

- Chirp mass:  $\mathcal{M} = \frac{(m_1 m_2)^{3/5}}{(m_1 + m_2)^{1/5}}$
- Total mass:
   ringdown

• Mass ratio: q =





#### Effects of spins

• 2 spin vectors



- Magnitude: orbital hang-up
- Mis-alignment: precession and modulations



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  - Magnitude: orbital hang-up



Mis-alignment: precession and modulations



• We want the **posterior** probability of parameters  $\vec{\lambda}$ , given the data  $\vec{x}$ . With **Bayes'** theorem:

$$p(\vec{\lambda}|\vec{x}, M) = \frac{p(\vec{\lambda}|M) p(\vec{x}|\vec{\lambda}, M)}{p(\vec{x}|M)}$$

- Fit a model to the data (noise and signal models)
- Build a likelihood function
- Specify **prior** knowledge
- Numerically estimate the resulting distribution (sampling algorithms)

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#### Likelihood



- How close is the **remainder** to the **mean**?
  - Assumptions: gaussianity and stationarity

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- Specify **prior** knowledge
- Numerically estimate the resulting distribution (efficient sampling algorithms) [Raymond, et al. 2010]



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#### Markov-Chain Monte Carlo

- High dimensional parameter space
- Slow waveform computation

#### Efficient sampling critical (especially with precession)



# Gravitational-wave observations in the first observing run (O1)



<sup>[</sup>LIGO-Virgo Collaboration, 2016]

#### GW150914: masses

- 2 models as a proxy for systematic errors:
  - Double-precessing-spin
     model (SEOBNRv3)
  - Single-precessing-spin model (IMRPhenomP)

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 $\cdot$  Errors:



### **2.3** GW150914: remnant black hole

- Final values fitted from Numerical Relativity simulations
  - Final mass:

$$M_f = 62.2^{+3.7}_{-3.4} \,\mathrm{M}_{\odot}$$

• Final (dimensionless) spin:

 $a_f = 0.68^{+0.05}_{-0.06}$ 

• ~3 solar mass radiated !



#### GW150914: location



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### CBC LIGO sky maps Electromagnetic counterpart

- Bayestar O(minutes)
- LALInference-lite O(hours)
  - Includes spin effects
  - Sub-threshold triggers in part of a **network**
- Full LALInference O(daysweeks)
- Sky localisation degeneracies with only 2 detectors
   [Raymond, et al., 2009]



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• **Degeneracies** in **extrinsic** parameters, strain *h*:

$$h = -\frac{1 + \cos^2(\iota)}{2 D_L} F_{j+}(\mathbf{R}.\mathbf{A}., \det, \psi) H_+$$
$$+ \frac{\cos \iota}{D_L} F_{j\times}(\mathbf{R}.\mathbf{A}., \det, \psi) H_{\times}$$

3 angles for the orientation: (R.A., dec,  $\psi$ ) Intrinsic waveform:

 $H_{+,\times}(m_1, m_2, \vec{S_1}, \vec{S_2})$ 

• Sampling in LALInference [Raymond, Farr, 2014]



#### GW150914: spins

- Weak constraints on spin magnitude
  - Very weak constraints on spin orientation
- Due to Almost equalmass, face-off binary
   [Raymond, 2012]
   [LIGO-Virgo Collaboration, 2013]





#### Were the black-holes spinning?



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#### Some results of the first observing run (O1)

- Observational medium delivers heavy stellar mass black-holes
- Merging binary black holes exist in a broad mass range
- New access to black holes spins (GW151226 at least one black-hole spinning)
- Measured **masses** and **spins** consistent with both:
  - Isolated binary evolution (more aligned spins)
  - **Dynamical formation** (more misaligned spins)
- Statistical errors dominate waveform systematical errors

#### Ongoing work in Gravitational-wave astrophysics

- Joint analysis of electromagnetic and gravitational-wave data
  - Understanding of **extreme** astrophysical phenomena
  - Higher probability of astronomical origin, better estimations
- Testing General Relativity (with black-hole ringdowns)
- Waveform modelling:
  - Reduced Order Modelling [Canizares, Field, Gair, Raymond, et al., 2015]
  - Calibration of waveform models against Numerical
     Relativity [Bohé, Shao, Taracchini, Buonanno, Babak, Harry, Hinder, Ossokine, Pürrer, *Raymond*, et al., 2016]
- Towards automated interferometers control [Driggers, Raymond, et. al., 2014]
- Combining Observations [Raymond, Price, 2015; Raymond, Price, Gendler, in prep]

### **Towards Automated** Control

Mass

Photodetector

Signal Recycling



- Test Mass More **sensitive** detector Higher duty cycle • 4 km Inform design of **future instruments** • • Test Mass Power Beam 4 km ——— Recycling Splitter 100 kW Circulating Power Laser Source Test Test
- Improving gravitational-wave observatories:

**Optimize** for specific **astrophysical sources** 

Mass

#### Towards Automated Control



# Beyond the first observing run (O1)

More Binary Black Holes

•

- Better **spin** constraints (magnitude AND orientation)
- Neutron stars in binaries

- New tests of General Relativity
- Neutron stars equation of state
- Population of compact objects



# Combining detections

- New tests of General Relativity
- Neutron stars equation of state
- Mass gap

•

- Field and cluster populations
  - Star formation parameters



#### For instance:

- Neutron-star mass distribution:
  - Iron-core collapse supernovae

 $\approx 1.35\,M_{\odot}$ 

Electron-capture supernovae

 $\approx 1.25\,M_{\odot}$ 

[Knigge, et al., 2011, Schwab, et al., 2010]



### Parametrisation of a population

Neutron-star mass distribution:

Parameters:

$$\mu_1 = 1.246$$
  
 $\sigma_1 = .008$   
 $\mu_2 = 1.345$   
 $\sigma_2 = 0.025$   
 $h_{12} = .293$ 

Model inspired by [Schwab, et al., 2010]

Typical Neutron Star mass estimation from 1 observation [Rodriguez, Farr, *Raymond*, et al., 2014]



#### Framework to combine observations

- There is a dense literature on how to use gravitational waves from compact binary coalescence to:
  - **distinguish** Source populations [Stevenson, et al. 2015; Littenberg, et al. 2015, Mandel et al. 2015]
  - mitigate detection and observation bias [Gair, Moore, 2015; Messenger, Veitch, 2012]
  - measure source distribution meta-parameters, [Lackey, Wade 2014]

All of the above in a common treatment [Raymond, Price, 2015; Raymond, Price, Gendler, in prep]

 example with N~1000 (optimistic end of O3), we could resolve the distribution

#### Future outlook:

- What are the properties of gravitational waves? Is General Relativity still valid under strong-gravity conditions?
- How does matter behave under extremes of density and pressure?
- How abundant are stellar-mass binary black holes? And what are the mass distributions of coalescing compact objects?
- How are compact binaries that coalesce formed, what is their accretion history and what has been their effect on star formation rates?
- Is the mechanism that generates gamma-ray bursts a compact binary coalescence?
- Where and when do **massive black holes** form, and what role do they play in the **formation** and **evolution of galaxies**?
- And the unexpected !

