#### **Gravitational Wave Project**

"Matched filtering"

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BND Graduate School 2024 Blankenberge, 2024 Sep 02-12

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# Gravitational-wave sources







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## Matched filtering



# In this project ...

- Code up a simple matched filtering algorithm
   Newtonian chirp waveform
   Gaussian white noise
- Search for a "fake" signal (with above assumptions)
   When?? [time] How heavy?? [mass]
- Look at stretches of LIGO-Virgo-KAGRA data (gwpy)
   Determine its noise properties
- Try to generalize above algorithm for real data analysis Discuss complications

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#### Pros and cons ...

Project likely on the easier side :)

In line with the GW lectures by Elena Cuoco :)

Limited help available until until the very end :(

Python, SciPy, conda (gwpy)  $\dots$  | cluster / laptop

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## Reference slides

# Gravitational-wave sources







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## Gravitational-wave data analysis

Typically signal buried in noise significantly louder than itself.

In order to search for / study any signal, one needs a thorough understanding of the detector noise.

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## Stationary Gaussian noise

Detector noise is **random**: detector output in absence of signal is a random number at every instant of time.

Stationary?

Its statistical properties do not change over time!

Caveat: they do!

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#### Stationary Gaussian noise

Correlation of noise with itself at an earlier (or later) time

$$\langle n(t)n(t+\tau)\rangle \equiv \frac{1}{T}\int_{-T/2}^{T/2} dt n(t) n(t-\tau) = \kappa(\tau)$$

is a such a statistical property | "ergodicity"

In Fourier space

power spectral density

 $\langle \tilde{n}(f)\tilde{n}^*(f')\rangle = S_n(f)\delta(f-f'), \text{ where } S_n(f) \equiv \int_{-\infty}^{\infty} dt \,\kappa(t)e^{i2\pi ft}$  $\kappa(\tau) = \kappa(-\tau) \Rightarrow S_n(f) \text{ is real } \text{ dimensions of time } \text{ unit: } \text{Hz}^{-1}$ 

## Stationary Gaussian noise

$$S_n(f) = ext{constant}$$
 : white noise  
 $S_n(f) \propto rac{1}{f}$  : flicker noise  
 $S_n(f) \propto rac{1}{f^2}$  : random walk

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#### Gravitational-wave detector noise



## Non-stationary effects

#### PSD slowly changing with time

transient "glitches"

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#### Searching for modelled signal in detector data

detector data = possible signal + noise:  $x(t) = h(t - t_{arr}) + n(t)$ 

Correlation of template q(t) with detector output.

$$c(\tau) \equiv \int_{-\infty}^{\infty} dt \, x(t) \, q(t+\tau)$$

lag  $\tau$ : effectively concentrates all information in signal in one place

**optimal filter** q(t) maximizes  $c(\tau)$  when h(t) in detector output

In frequency domain

$$c( au) = \int_{-\infty}^{\infty} df \, ilde{x}(f) \, ilde{q}^*(f) e^{-i2\pi f au}$$

Searching for modelled signal in detector data

$$S \equiv \langle c(\tau) \rangle = \int_{-\infty}^{\infty} df \, \tilde{h}(f) \, \tilde{q}^*(f) e^{-i2\pi f(\tau - t_{arr})}$$
$$N^2 \equiv \langle |c - \langle c \rangle|^2 \rangle = \int_{-\infty}^{\infty} df \, S_n(f) \, |\tilde{q}(f)|^2$$

Noise-weighted inner product

given a(t), b(t)

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$$\langle a|b\rangle \equiv 2\int_0^\infty \frac{df}{S_n(f)} \left[ \tilde{a}(f)\tilde{b}^*(f) + \tilde{a}^*(f)\tilde{b}(f) \right]$$

Signal-to-noise ratio (SNR):

$$\rho \equiv \frac{S}{N} = \frac{\langle h e^{i2\pi ft} | S_n q \rangle}{\langle S_n q | S_n q \rangle^{\frac{1}{2}}}$$

## Searching for modelled signal in detector data

Signal-to-noise ratio (SNR):

$$\rho \equiv \frac{S}{N} = \frac{\langle h e^{i2\pi ft} | S_n q \rangle}{\langle S_n q | S_n q \rangle^{\frac{1}{2}}}$$

**Optimal template** that maximizes  $\rho$  is

$$ilde{q}(f) \sim rac{ ilde{h}(f) e^{i 2 \pi f(t-t_{\mathsf{arr}})}}{S_n(f)}$$

- SNR maximized when  $\tau = t_{arr}$
- Optimal filter = signal weighted down by PSD (not just copy of signal)

Optimal SNR

$$\rho_{\rm opt} = \frac{\langle h|h\rangle}{\langle h|h\rangle^{\frac{1}{2}}} = \langle h|h\rangle^{\frac{1}{2}} = 2\left[\int_0^\infty df \frac{|h(f)|^2}{S_n(f)}\right]^{\frac{1}{2}}$$

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## Matched filtering



# Searching for unmodelled signal in detector data

• Wavelet reconstruction: look for coherent excess power







#### Detector strain from an astrophysical signal

- Location of source w.r.t (arms of) the detector:  $(\theta, \phi)$
- Polarization angle:  $\psi$

$$F_{+}(\theta,\phi,\psi) = -\frac{1}{2}(1+\cos^{2}\theta)\cos 2\phi\cos 2\psi - \cos\theta\sin 2\phi\sin 2\psi \quad (A.10)$$
$$F_{\times}(\theta,\phi,\psi) = +\frac{1}{2}(1+\cos^{2}\theta)\cos 2\phi\sin 2\psi - \cos\theta\sin 2\phi\cos 2\psi \quad (A.11)$$

These beam pattern functions are shown in Figure A.1.



**Figure A.1** The beam pattern functions  $F^2_+(\theta, \phi, \psi = 0)$  (a) and  $F^2_\times(\theta, \phi, \psi = 0)$  (b) for an interferometric gravitational-wave detector with orthogonal arms along the *x*- and *y*-axes.

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#### detector strain $h = F_{+}(\theta, \phi, \psi)h_{+}(t - t_{arr}) + F_{\times}(\theta, \phi, \psi)h_{\times}(t - t_{arr})$

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#### Searching for unmodelled signal in detector data

Multiple reconstructions: which reconstruction to choose?

Bayesian Occam's razor: simpler reconstruction; fewer params

detector strain = signal + glitch + gaussian noise

signal model:  $F^+_{detector}(\theta, \phi, \psi)h_+(t - t_{arr}) + F^{\times}_{detector}(\theta, \phi, \psi)h_{\times}(t - t_{arr})$ glitch model: independent sum of wavelets in each detector noise model: gaussian noise of given (or measured) PSD



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Bonus: model selection & glitch removal!

## Compact binary coalescences

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## Two-body problem in GR

Note:  $f_{GW} = 2f_{orb}$ 

### Two-body problem in GR

$$L_{\rm GW} = \frac{1}{5} \frac{G}{c^5} \langle \vec{I}_{ij} \vec{I}^{ij} \rangle = \frac{32}{5} \frac{c^5}{G} \eta^2 \left( \frac{\pi GM f_{\rm GW}}{c^3} \right)^{10/3}$$
  
Larmor formula

$$E = \frac{1}{2}\mu(\pi GMf_{\rm GW})^{2/3} \qquad \qquad L_{\rm GW} = -\frac{dE}{dt}$$

Newtonian energy

energy conservation

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$$\Rightarrow \qquad \frac{df_{\rm GW}}{dt} = \frac{96}{5} \pi^{8/3} \left(\frac{G\mathcal{M}_c}{c^3}\right)^{5/3} f_{\rm GW}^{11/3}$$
with chirp mass  $\mathcal{M}_c \equiv \frac{(m_1 m_2)^{3/5}}{(m_1 + m_2)^{1/5}}$ 

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Leading order of the **post-Newtonian expansion**.



## Post-Newtonian expansion

Expansion in powers of 
$$\left(\frac{\nu}{c}\right)^2$$
  

$$P_{gw} = \frac{32c^5}{5G}\nu^2 x^5 \left\{ 1 + \left( -\frac{1247}{336} - \frac{35}{12}\nu \right) x + 4\pi x^{3/2} + \left( -\frac{44711}{9072} + \frac{9271}{504}\nu + \frac{65}{18}\nu^2 \right) x^2 + \left( -\frac{8191}{672} - \frac{583}{24}\nu \right) \pi x^{5/2}$$

$$+ \left( -\frac{6643739519}{69854400} + \frac{16}{3}\pi^2 - \frac{1712}{105}C - \frac{856}{105}\log(16x) + \left( -\frac{134543}{7776} + \frac{41}{48}\pi^2 \right) \nu - \frac{94403}{3024}\nu^2 - \frac{775}{324}\nu^3 \right] x^3 + \left( -\frac{16285}{504} + \frac{214745}{1728}\nu + \frac{193385}{3024}\nu^2 \right) \pi x^{7/2} + \mathcal{O}\left(\frac{1}{c^8}\right) \right\}$$

GW150914: LVC: Abbott+, PRL 116, 061102 (2016)



# Waveform approximants

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Post-	Newtor	nan

TaylorT, TaylorF

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Hybrid, phenomenological

IMRPhenom, IMRPhenomP, IMRPhenomX

Effective one-body

SEOBNR

reduced-order-model

ROM

## Intrinsic parameters

- Component masses, *m*<sub>1,2</sub>
- Component dimensionless spin angular momenta,  $\vec{s}_{1,2}$
- Tidal deformability parameters for neutron stars,  $\lambda_{1,2}$

NS in binaries deform in gravitational field of companion deformation depends on NS composition (equation-of-state) leaves imprint on GW signal (waveform)

• Any residual eccentricity?

radiated away Peters & Matthews (1963)

 $\vec{s} \equiv \frac{J}{m^2}$ 

## Two polarizations

Dominant (2, 2)-mode to leading order:

$$h_{+}(t) = \frac{2M\eta}{d_L} (\pi M f_{\rm GW})^2 (1 + \cos^2 \iota) \cos 2\varphi(t)$$
$$h_{\times}(t) = \frac{4M\eta}{d_L} (\pi M f_{\rm GW})^2 \cos \iota \sin 2\varphi(t)$$



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- Distance,  $d_L$
- Inclination angle,  $\iota$
- Phase at coalescence,  $\phi_c$
- Time of coalescence,  $t_c$

#### Antenna beam pattern functions

- Two angles in the sky  $( heta,\phi)\longrightarrow(lpha,\delta)$ , right ascension and declination
- Polarization angle,  $\psi$

$$F_{+}(\theta,\phi,\psi) = -\frac{1}{2}(1+\cos^{2}\theta)\cos 2\phi\cos 2\psi - \cos\theta\sin 2\phi\sin 2\psi \quad (A.10)$$
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## **CBC** parameters

Intrinsic parameters:  $\{m_1, m_2, \vec{s_1}, \vec{s_2}, \lambda_1, \lambda_2, \ldots\}$ Extrinsic parameters:  $\{\alpha, \delta, d_L, \iota, \psi, \varphi_c, t_c\}$ 

$$\textit{F}_{\text{detector}}^{+}(\theta,\phi,\psi)\textit{h}_{+}(t-\textit{t}_{\text{arr}}) + \textit{F}_{\text{detector}}^{\times}(\theta,\phi,\psi)\textit{h}_{\times}(t-\textit{t}_{\text{arr}})$$

**masses** of the components **spin** angular momenta **tidal deformability** for neutron stars **eccentricity distance**  $d_L$ **sky position**  $\alpha, \delta$ 

euler angles  $\{\theta, \phi, \psi\} = \text{fn.}(\{\iota, \psi, \varphi_c\})$ time of arrival

### Bayesian parameter estimation

Obtain the posterior probability distribution on the parameter space given the data and a prior probability distribution.

$$\textit{Posterior}(\vec{\Omega}|\texttt{data}, I) = \frac{\textit{Prior}(\vec{\Omega}|I) \, \mathcal{L}(\texttt{data}|\vec{\Omega}, I)}{\textit{Evidence}(\texttt{data}, I)}$$

$$\vec{\Omega} = \{\mathcal{M}, \boldsymbol{q}, \vec{\boldsymbol{s}_1}, \vec{\boldsymbol{s}_2}, \lambda_1, \lambda_2, \alpha, \sin \delta, \boldsymbol{d_L}, \cos \iota, \psi, \varphi_c, t_c\}$$

$$\mathcal{L}(\mathsf{data}|\vec{\Omega}, I) = P(\mathsf{data}|\mathsf{signal}(\vec{\Omega}), I) \qquad \mathsf{data} = \mathsf{signal}(\vec{\Omega}) + \mathsf{noise}, n$$

$$= \exp\left(-\frac{1}{2}\left\langle\mathsf{data} - \mathsf{signal}(\vec{\Omega})|\mathsf{data} - \mathsf{signal}(\vec{\Omega})\right\rangle\right)$$

$$\stackrel{\mathsf{adta}}{=} \operatorname{signal}(\vec{\Omega}) = \exp\left(-\frac{1}{2}\left\langle\mathsf{data} - \mathsf{signal}(\vec{\Omega})|\mathsf{data} - \mathsf{signal}(\vec{\Omega})\right\rangle\right)$$

$$\stackrel{\mathsf{adta}}{=} \operatorname{signal}(\vec{\Omega}) = \operatorname{signal}($$

Bayesian parameter inference algorithms

Markov Chain Monte Carlo (MCMC)

Metropolis-Hastings

Parallel tempering (MCMC)

Nested sampling

1024 live points

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## Data analysis workflow of CBCs



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### Parameter estimation results





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