Gravitational wave data analysis an overview

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Analyzing GW signals

- Detection, low-latency alerts
- Bayesian parameter estimation (PE) for each source
- Hierarchical Bayesian analyses (population, cosmology, ...)

Coalescence of compact binaries (BBH, NSBH, BNS): waveform templates for detection (digging in the noise) and PE

- Theoretical challenge
- Computational challenge

Left aside in this talk:

- unmodeled searches (for SN, and the unexpected)
- Stochastic backgrounds



Data analysis basics



Data analysis basic tools

Matched filter overlap:

$$(h_1|h_2) = 4\operatorname{Re} \int df \, \frac{\tilde{h}_1(f)\tilde{h}_2^*(f)}{S_n(f)}$$

Matched filter SNR: (h|d)

Likelihood (stationary, Gaussian): $\ln \mathcal{L}(d|\theta) = -\sum_{\text{channels}} \frac{1}{2}(h(\theta) - d|h(\theta) - d)$ $d = h(\theta_0) + n_0 \quad \text{(+ calibration uncertain.)}$

Instrument and noise

Noise PSD, idealized: $\langle \tilde{n}(f)\tilde{n}^*(f')\rangle = \frac{1}{2}S_n(f)\delta(f-f')$

- stationarity
- Gaussianity

Detector Characterization (DetChar):

- calibration, noise removal (lines)
- PSD estimation
- glitch identification/removal



Bayesian parameter estimation, posterior:

$$p(\theta|d) = rac{\mathcal{L}(d|\theta)p_0(\theta)}{p(d)}$$
 $p_0(\theta)$ prior $p(d)$ evidence

LIGO/Virgo detections: methods

Template bank searches

Precomputed bank of templates optimally placed in parameter space

Modified detection statistic: take into account SNR and chi-square of the residuals

Trade-off between sensitivity of the search and false alarm rate

Signal / background, p_astro

Time slides: generate large amount of background data, defines false alarm rate (**FAR**)

Poisson model for rates (assume population): astrophysical or instrumental origin Probability of astrophys. origin (**p_astro**)



Main pipelines

- PyCBC: [Usman&al 2016]
- GstLAL: [Cannon&al 2012]
- **MBTA**: [Adams&al 2016]
- SPIIR: [Chu&al 2020]

Online (low-latency) and offline versions

LIGO/Virgo detections: results

First detection: GWI509I4

Luckily, loud and clear $!~>5\sigma$





GWTC-3

- 90 detections in total
- 82 BBH
- 6 NSBH
- 2 BNS
- large number of sub-threshold triggers

LIGO/Virgo parameter estimation: methods

Sampling



- MCMC: evolve a chain with proposal and Metropolis-Hastings acceptance — parallel tempering, tuned proposals
- Nested sampling: sample uniformly from inside isolikelihood contours (and get evidence)
- + Machine learning ?

Pipelines

- LALInference: MCMC (PT), Nest (nested sampling chain proposals) [Veitch&al 2015]
- **Bilby**: nested sampling Dynesty [Singer&al 2015]
- **RIFT**: (expensive waveforms) two stages, first model marginal intrinsic likelihood [Singer&al 2015]



[LIGO-Virgo 2017]

LIGO/Virgo parameter estimation: results

GWTC-3





GWTC-3b

[LVK 2021]

q

 $\chi_{\rm eff}$

 $\mathcal{M}[M_{\odot}]$

0.5

 $\chi_{\rm p}$

 $D_{\rm L}$ [Gpc]

GW200322_091133

The future of Gravitational Wave Astronomy



Third-generation detectors: data analysis challenges



[Cosmic Explorer]

Number of detections

Events/yr (low-median-high):

- BBH: 60k-90k-150k
- BNS: 300k-1000k-3000k

Detections (2 CE + I ET):

- BBH: 93%
- BNS: 35%

(from [Samajdar&al 2021])

Computational challenge !

Third-generation detectors: data analysis challenges

mergers



LISA sources



LISA data - LDC-2 Sangria



- **MBHBs**: chirping signals, emerging from low-f noise
- **GBs**: quasi-monochromatic, horizontal lines

Contrasting LIGO/Virgo and LISA



Low-f approximation: **two LIGO-type detectors** in motion [Cutler 1997]



High-f: **three channels** frequency-dependence

LISA response

Laser frequency shift, spacecrafts s to r through link I: $y = \Delta \nu / \nu$ $y_{slr} = \frac{1}{2} \frac{1}{1 - \hat{k} \cdot n_l} n_l \cdot (h(t_s) - h(t_r)) \cdot n_l$

Chirping signals Fourier-domain:

$$\mathcal{T}_{slr} = \frac{i\pi fL}{2} \operatorname{sinc} \left[\pi fL \left(1 - k \cdot n_l\right)\right] \exp\left[i\pi f \left(L + k \cdot \left(p_r + p_s\right)\right)\right] n_l \cdot P \cdot n_l(\boldsymbol{t_f})$$

Time and frequency-dependency Time: motion of LISA on its orbit Frequency: departure from long-wavelength + Time-delay interferometry (TDI)

linear combinations with more delays

Galactic binaries: signals and challenges

- Mostly WD-WD, some other compact objects
- Full galaxy: ~20 million systems !
- About ~20000 individually resolvable
- Form a (non-stationary) background
- Verification binaries
- Quasi-monochromatic GW emitters
- Modulation by LISA motion (sidebands in Fourier-domain)
- Superposition of signals in Fourier-domain



Galactic binaries: example results



Massive black holes: signals and challenges

- Very loud sources, SNRs of several thousands !
- Detection of merger easy, but detection as early as possible ?
- Advance localization for multimessenger observations ?
- Signals can be short (< Iday) and degenerate
- Waveform model systematics for such loud signals ? Biases, residuals for other sources ?
- Subdominant features in the signal are important



Massive black holes: example results





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Stellar-mass BHs

- Quiet signals: a few detections in the LISA band
- Inspiral regime far from merger, very large number of cycles $(10^5 10^6)$
- Challenge of detection: template banks impossible
- Multiband analysis, archival searches ?



LISA Data Challenge

https://lisa-ldc.lal.in2p3.fr/

LDC-I (Radler):

- Noise known in advance
- Not blind, sources known
- Single class of sources

The full LISA problem

- Noise reduction (TDI)
- Noise properties must be inferred
- Data artifacts (gaps, ...)
- Global fit required, updated regularly

LDC-2 (Sangria):

- Noise unknown
- Blind
- MBHBs + full GBs
 LDC-2 (Spritz):
- MBHB + data gaps
- Glitches

LDC-2 (Yorsh):

- EMRI
- SBHB



Machine learning applications

Reviews (rapidly changing!): [Cuoco&al 2019] https://iphysresearch.github.io/Survey4GWML/

DetChar, glitches

- Glitch classification (CNN, image recognition in time-frequency)
- Supervised Learning: GravitySpy, citizen science
- Denoising

Waveform models

- Remnant properties
- Coefficients on a reduced basis
- Interpolating waveforms (GPRs)

Detection

• CNNs for detection

Low-latency

• Sky map generation

Parameter estimation

Variational inference, likelihood-free

- Variational auto-encoder (CVAE)
- Normalizing flows

Machine learning example - fast PE

[Dax&al 2021]

- Normalizing flows
- Embedding network
- Guided neural parameter estimation

(b)

LISA data - LDC-2 Sangria Time-Domain

- **MBHBs**: loudest ones clearly visible by eye above the noise
- **GBs**: superposed signals, annual modulation due to the LISA motion

LISA data - LDC-2 Sangria Frequency-Domain

- **MBHBs**: loudest ones visible in the spectrum, subdominant
- **GBs**: signals local in frequency, both individually resolvable and building up a background

LISA: simulated catalog for MBHB astrophysical models

LISA: simulated catalog for MBHB astrophysical models

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[Barausse 2012] Astrophysical models:

- Heavy seeds delay
- Light seeds no delay
- PopIII seeds delay

Mass and t(SNR=10)

MBHB detected signals: Bulk shorter than ~10days Tail extending to ~3months

Mass and t(SNR=I)

SMBH PE: accumulation of information with time

Method

- Represent a cut in time-tomerger by a cut in frequency, becomes inaccurate at merger
- Use Multinest and PTMCMC with and without higher harmonics

only broken shortly before merger 2-maxima sky degeneracy

survives after merger

One-arm frequency observables

From spacecraft s to spacecraft r through link s: $y = \Delta \nu / \nu$

$$y_{slr} = \frac{1}{2} \frac{1}{1 - \hat{k} \cdot n_l} n_l \cdot (h(t_s) - h(t_r)) \cdot n_l$$

$$t_s = t - L - \hat{k} \cdot p_s, \quad t_r = t - \hat{k} \cdot p_r$$

$$h = h_+ P_+(\hat{k}) + h_\times P_\times(\hat{k}) \quad \text{GW at SSB}$$

Time-delay interferometry (TDI)

- Crucial to cancel laser noise
- First generation: unequal arms
- Second generation: propagation and flexing
- Michelson X,Y,Z Uncorrelated noises A,E,T

Approximations

- Long-wavelength approximation: two moving LIGOs rotated by $\,\pi/4\,$ + orbital delay $\,$
- Rigid approximation (order of the delays does not matter, delay=L simple in Fourier domain)

 $\mathbf{X}_{1}^{\mathrm{GW}} = \underbrace{\left[(y_{31}^{\mathrm{GW}} + y_{13,2}^{\mathrm{GW}}) + (y_{21}^{\mathrm{GW}} + y_{12,3}^{\mathrm{GW}})_{,22} - (y_{21}^{\mathrm{GW}} + y_{12,3}^{\mathrm{GW}}) - (y_{31}^{\mathrm{GW}} + y_{13,2}^{\mathrm{GW}})_{,33} \right]}_{X^{\mathrm{GW}}(t)} - \underbrace{\left[(y_{31}^{\mathrm{GW}} + y_{13,2}^{\mathrm{GW}}) + (y_{21}^{\mathrm{GW}} + y_{12,3}^{\mathrm{GW}})_{,22} - (y_{21}^{\mathrm{GW}} + y_{12,3}^{\mathrm{GW}}) - (y_{31}^{\mathrm{GW}} + y_{13,2}^{\mathrm{GW}})_{,33} \right]_{,2233}}_{X^{\mathrm{GW}}(t-2L_2-2L_3) \simeq X^{\mathrm{GW}}(t-4L)}.$

Extreme mass ratio inspirals: example results

SBHB signal: Fourier-domain signal and response

PE results: SBHBs

