

Equation of state and gravitational waves

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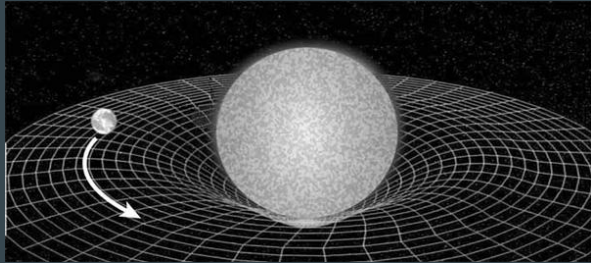
TALENT/ECT*24
18 July 2024



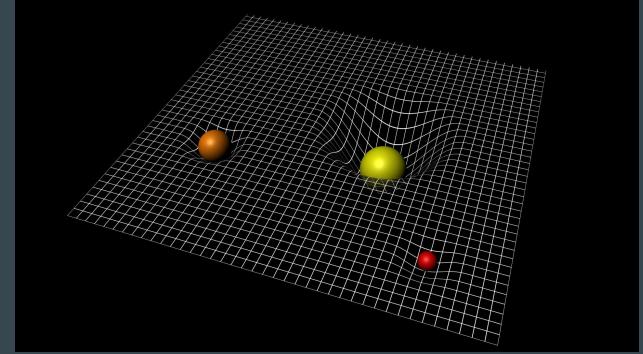
Tim Dietrich , Henrique Gieg, Vsevolod Nedora, Adrian Abac, Edoardo Giangrandi, Hauke Koehn, Nina Kunert, Ivan Markin, Anna Neuweiler, Henrik Rose, Federico Schianchi, Matthew Beaudoin, Ramon Jaeger, Sahil Jhavar, Hannes Kroke, Ranjith Mudimadugula, Karol Pawel Peters, Ashwin Shirke, Sophia Ve Dahm

General Relativity and gravitational waves

General Relativity: masses deform the spacetime geometry producing **curvature**, and this geometry is what determines how massive objects move



$g_{\alpha\beta}$: spacetime metric



Credit:ESA-C.Carreau

Einstein Field Equations

$$R_{\alpha\beta} - \frac{1}{2}Rg_{\alpha\beta} = \frac{8\pi G}{c^4}T_{\alpha\beta}$$

curvature

source of the curvature
(mass and energy)

General Relativity and gravitational waves

Weak field limit

$$g_{\alpha\beta} = \eta_{\alpha\beta} + h_{\alpha\beta}$$

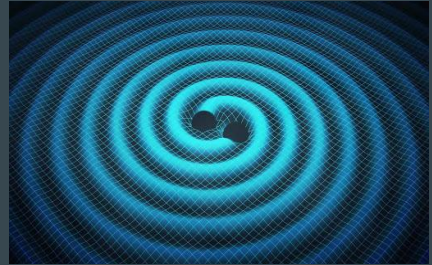
$$|h_{\alpha\beta}| \ll 1$$

↳ linearized theory of gravity (expansion of Einstein's equation around flat metric to linear order in $h_{\alpha\beta}$)

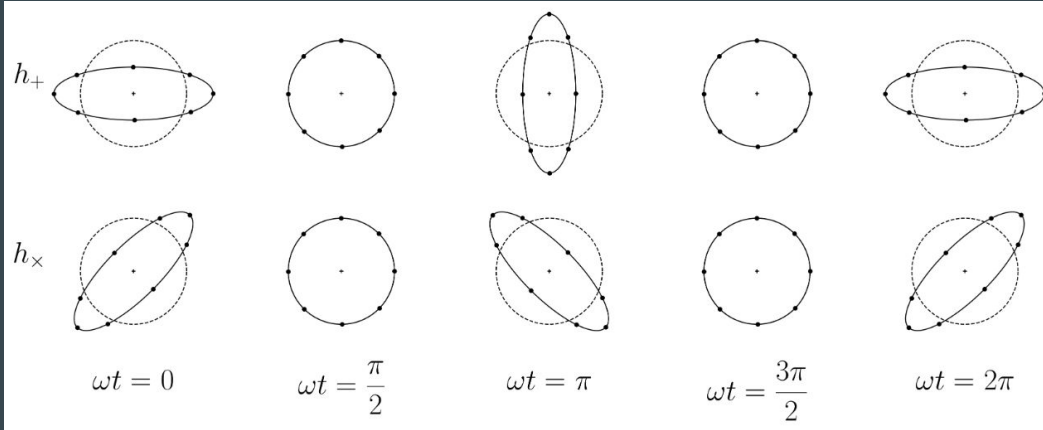
Linearized Einstein equations: $\square \bar{h}_{\alpha\beta} = -\frac{16\pi G}{c^4} T_{\alpha\beta}$

in vacuum $\square \bar{h}_{\alpha\beta} = 0 \implies$ wave-equation!

Perturbations of spacetime propagate on the flat spacetime background as plane waves \implies **gravitational waves.**

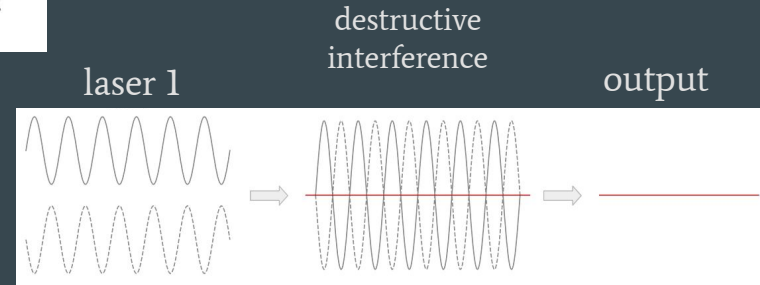
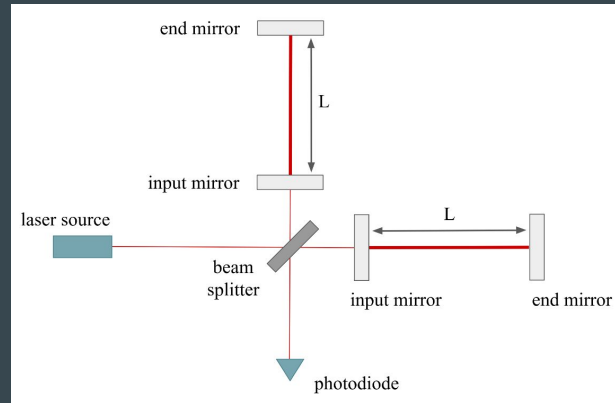


Gravitational waves detection



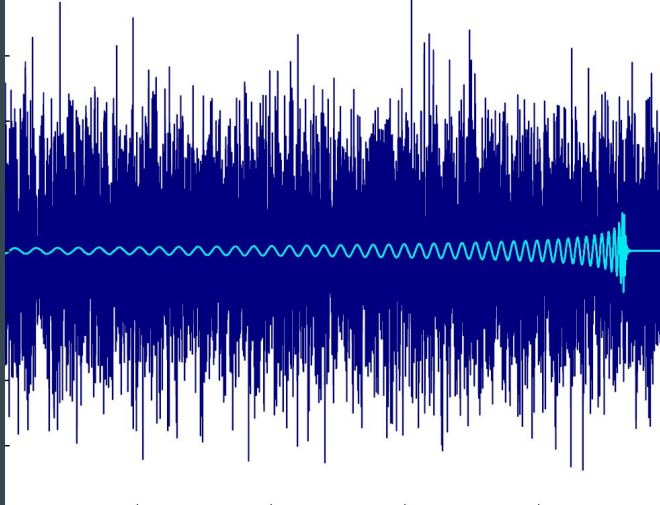
Gravitational waves:
stretch and compress
spacetime

Laser interferometers



$$h(t) = \frac{\delta L}{L} \sim \mathcal{O}(10^{-20})$$

Identifying signals



If you want to know more:
tutorial [here](#)

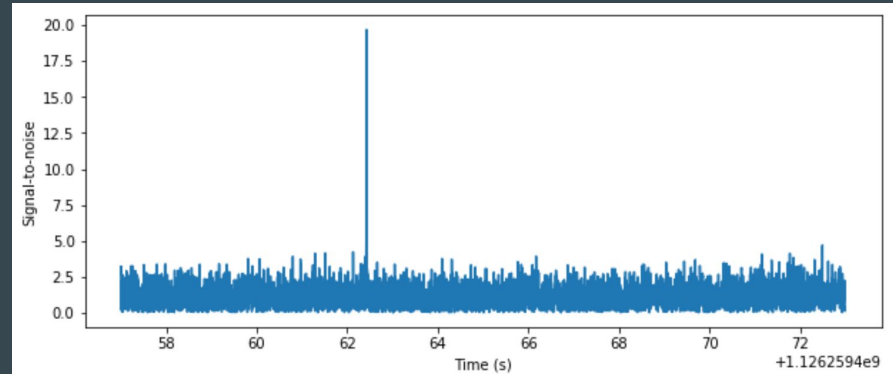


Matched filtering

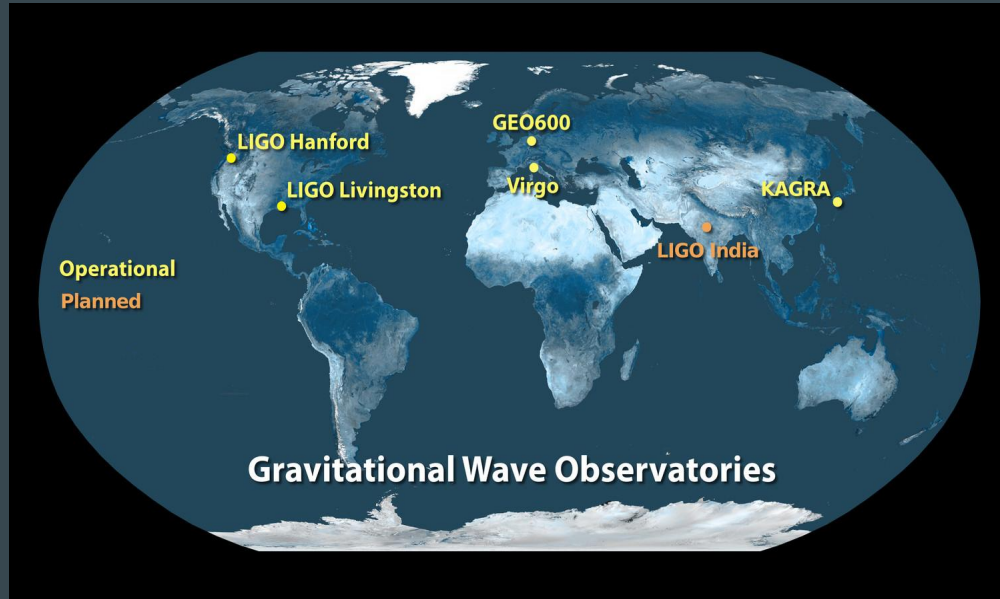
Compare data with *template banks*
(only main parameters)



Find *trigger time*

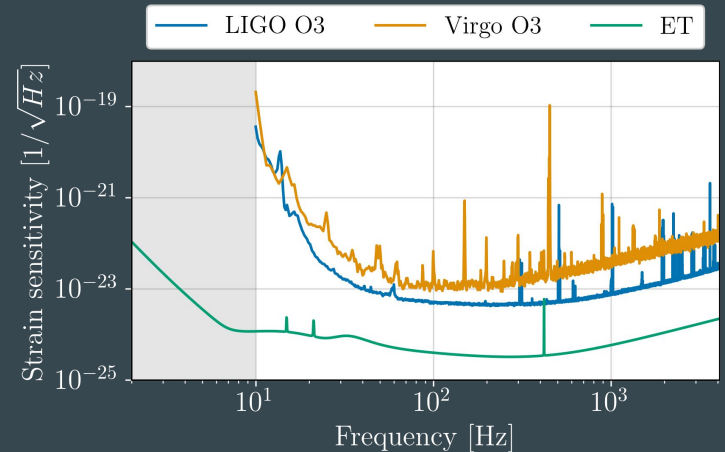


Gravitational waves detectors

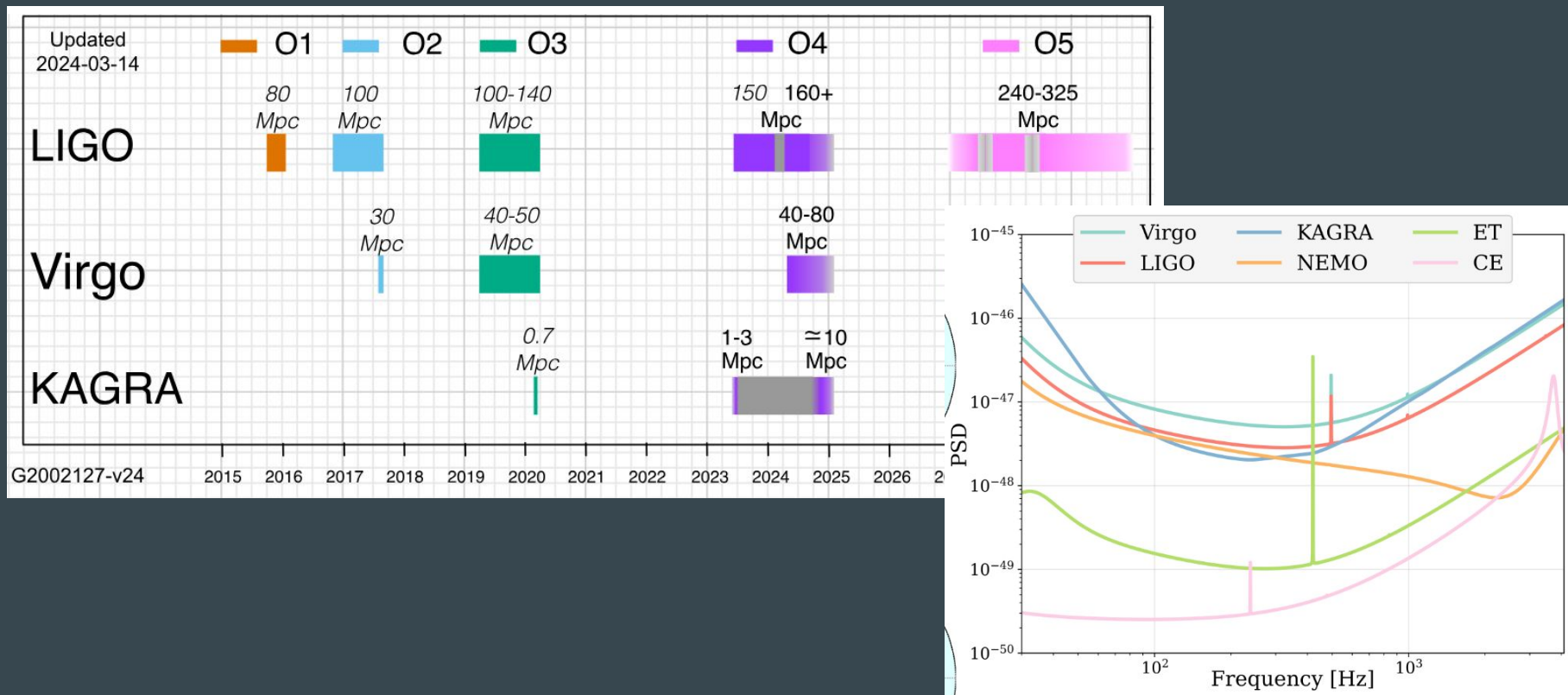


Next-generation detectors:

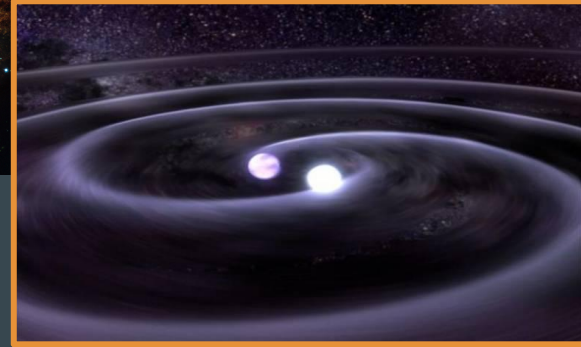
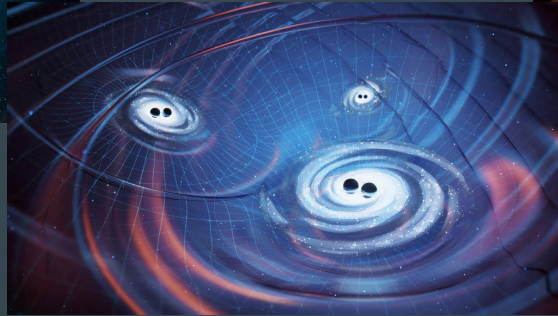
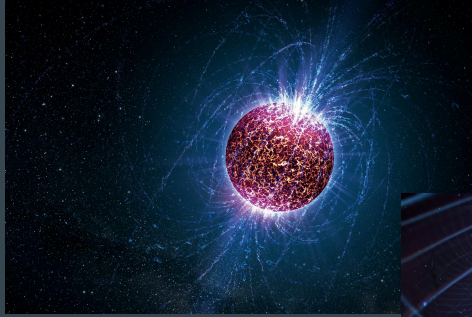
- Einstein Telescope (ET) in Europe
- Cosmic Explorer (CE) in the US



Gravitational waves detectors



Gravitational waves sources



Leading order: *mass quadrupole moment*

What we have

Sept '15 - Jan '16

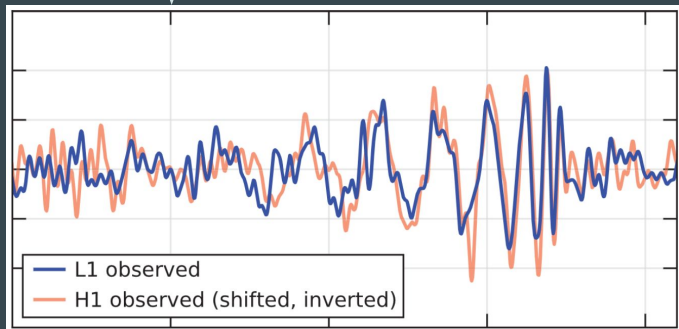
Nov '16 - Aug '17

Apr '19 - Mar '20

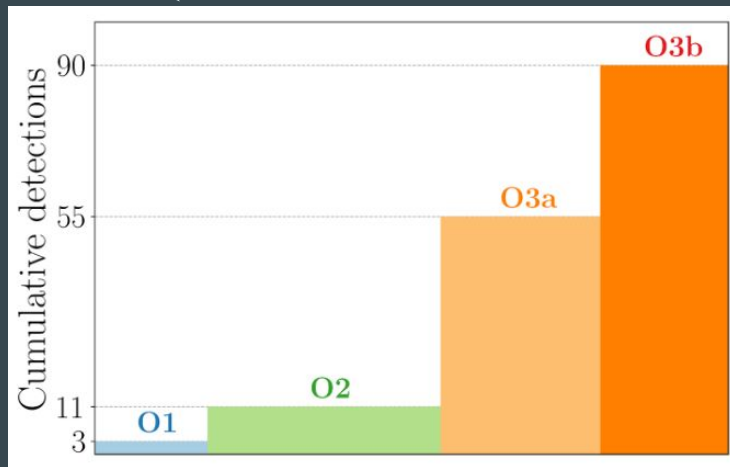
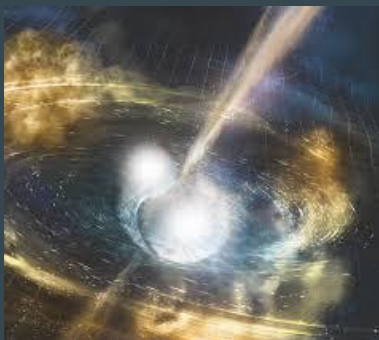
May '23 - ongoing



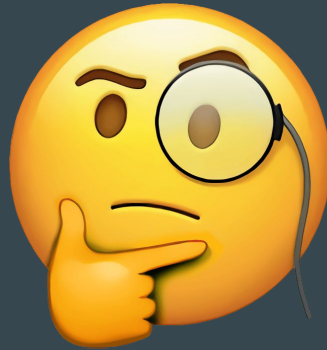
GW150914



GW170817

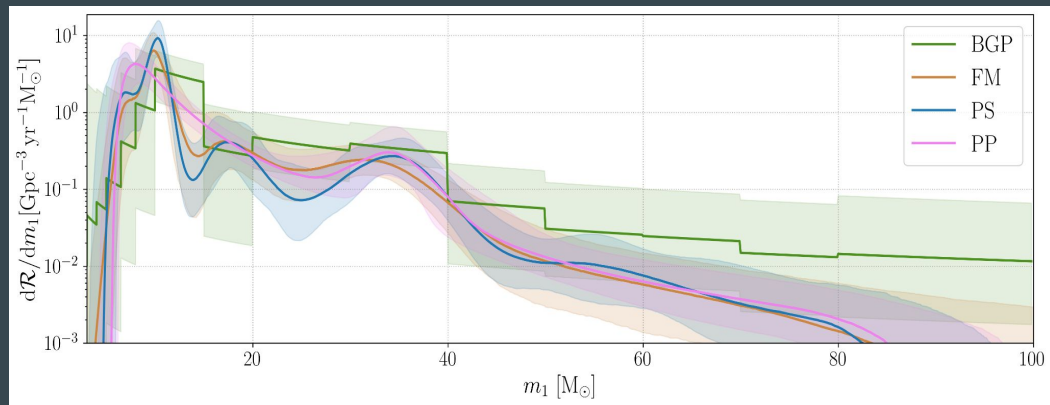


*What science can we do with
gravitational waves?*



Population properties

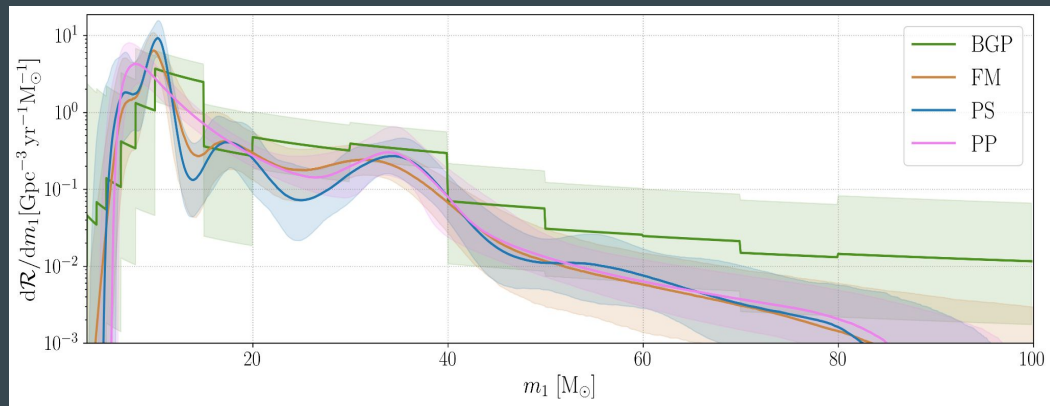
MASSES



LVK, *Phys.Rev.X* 13 (2023) 1, 011048

Population properties

MASSES



LVK, *Phys.Rev.X* 13 (2023) 1, 011048

SPINS

Binaries formation channels:

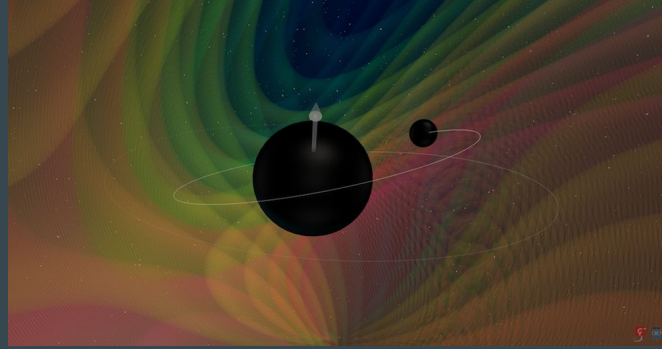
- Isolated binary evolution: aligned spins
- Dynamical interaction: precession, misaligned spins

Spin properties \iff Formation channels

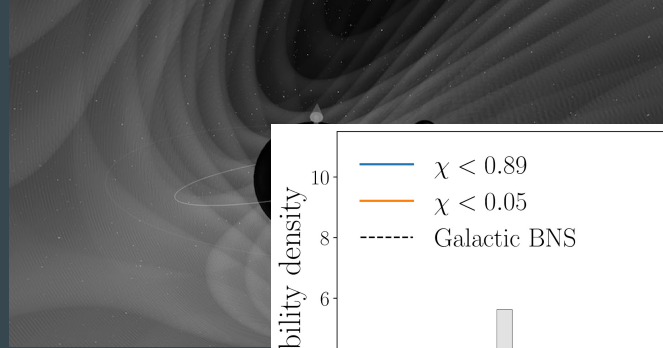
Investigate and model
different mechanisms

Special events

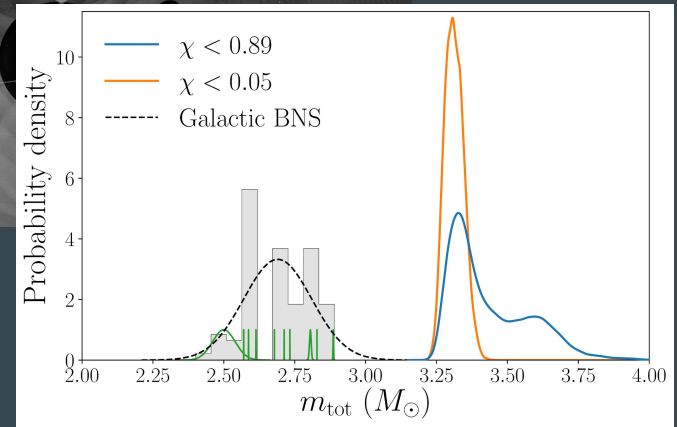
- GW190412: unequal masses



Special events

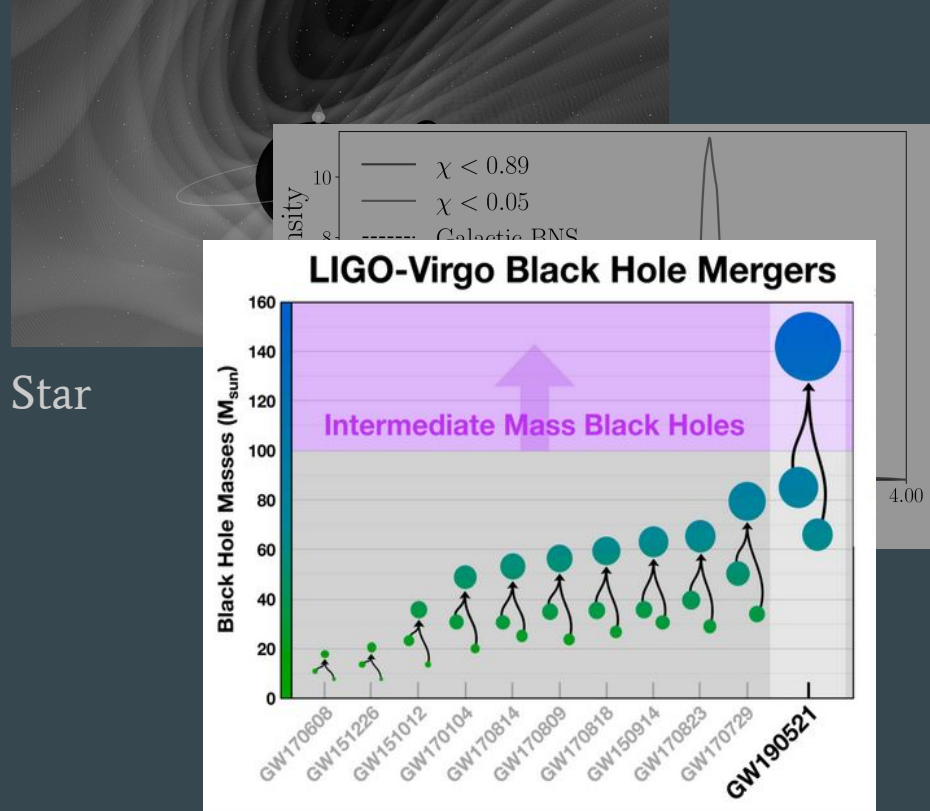


- GW190412: unequal masses
- GW190425: heavy Binary Neutron Star



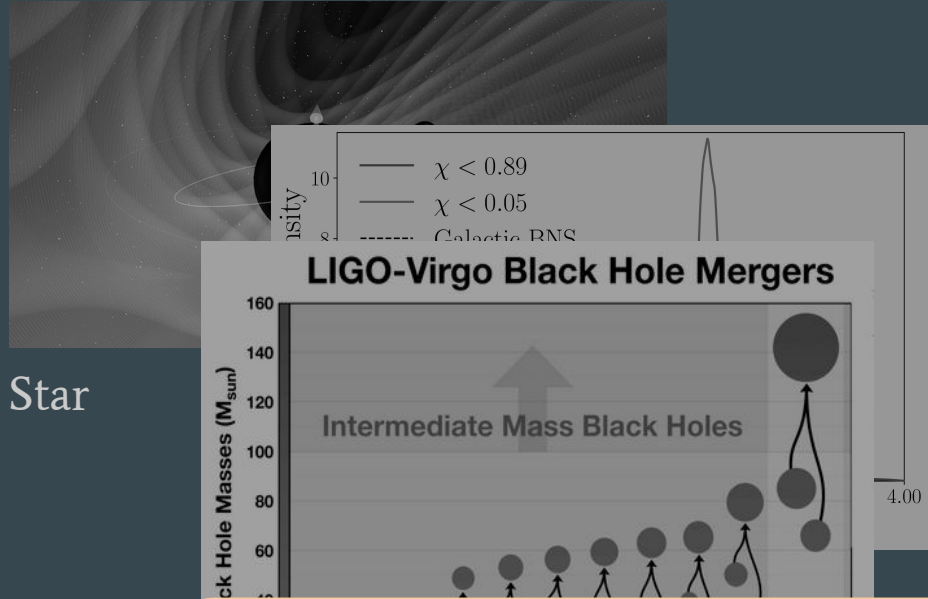
Special events

- GW190412: unequal masses
- GW190425: heavy Binary Neutron Star
- GW190521: the most massive



Special events

- GW190412: unequal masses
- GW190425: heavy Binary Neutron Star
- GW190521: the most massive
- GW190814: a mysterious compact object

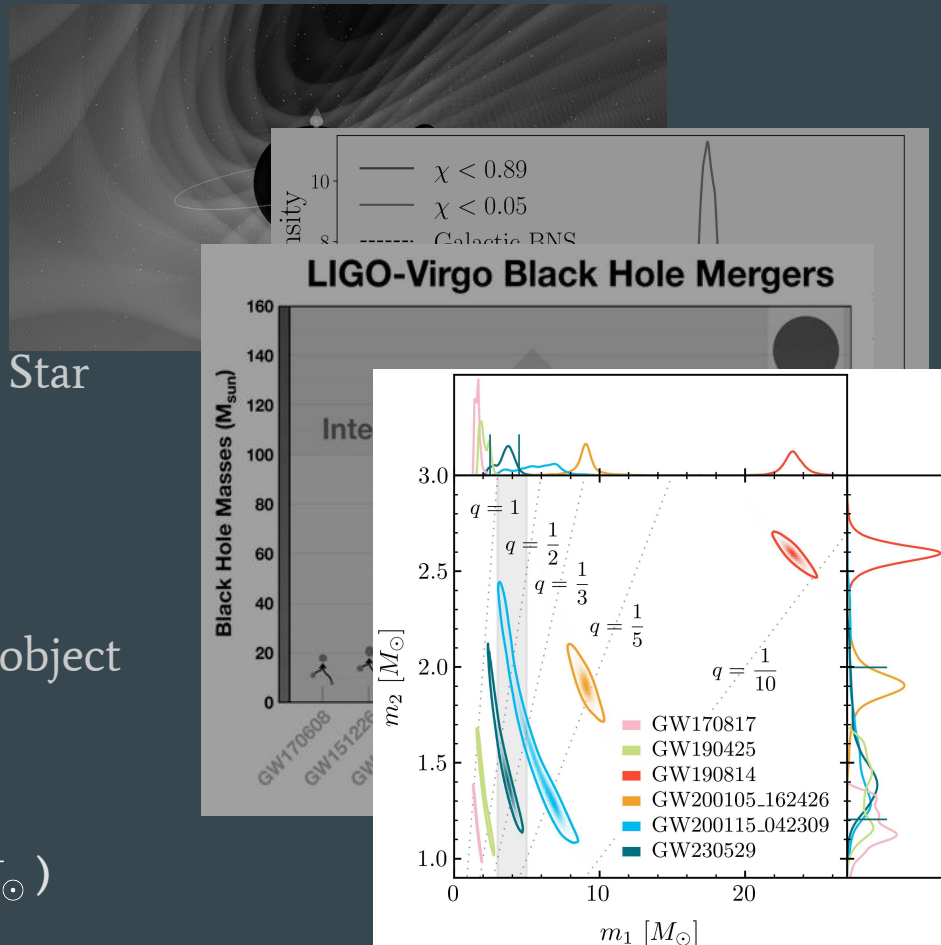


$m_1 \sim 23 M_{\odot} \Rightarrow$ Black Hole

$m_2 : 2.50 - 2.67 M_{\odot} \Rightarrow ???$

Special events

- GW190412: unequal masses
- GW190425: heavy Binary Neutron Star
- GW190521: the most massive
- GW190814: a mysterious compact object
- GW230529: mass-gap event
Primary object $m_1 = 3.6M_\odot$
BH in mass lower mass gap ($3-5 M_\odot$)



Testing general relativity

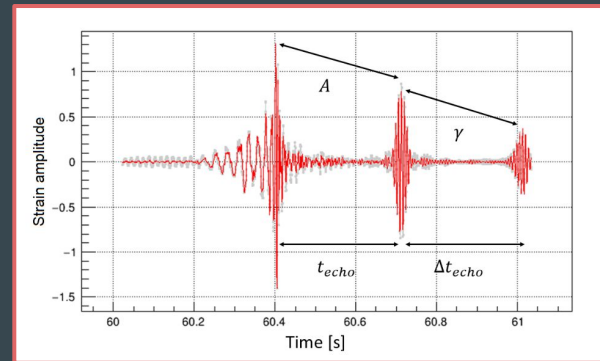
➤ Dynamical, strong-field regime

➤ Comparing data with general relativity predictions

- Parametrized deviations from the phase evolution
- Test the nature of compact objects: echoes (no horizon)
- Propagation of gravitational waves

...

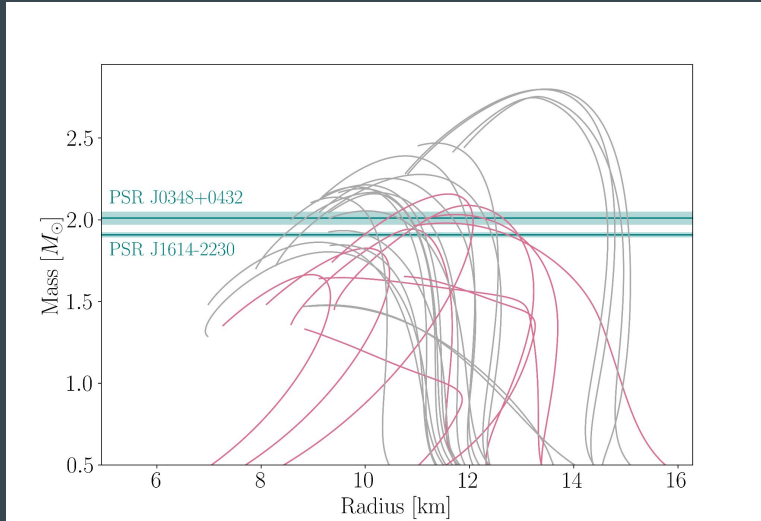
No violations of general relativity
found until now!



Phys.Rev.D 108 (2023) 6, 064018

Most recent LVK results:
arXiv:2112.06861

Nuclear matter properties



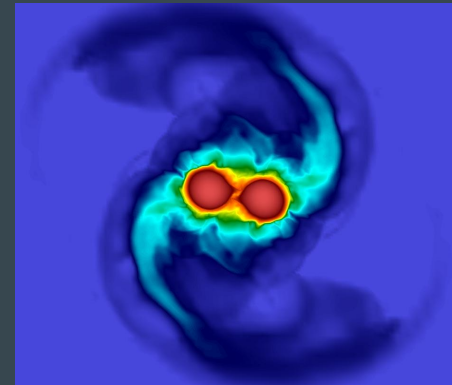
Neutron stars: supranuclear-dense matter

Equation of state :

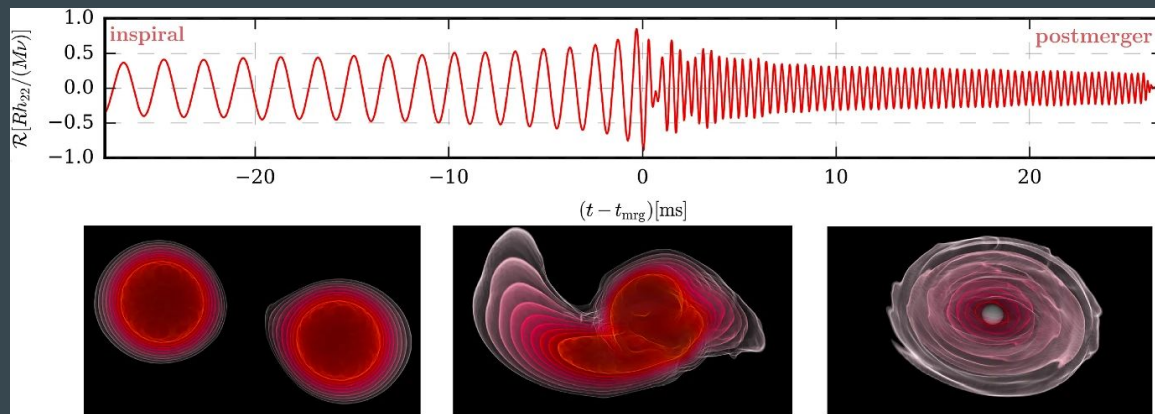
relation between pressure and density

↔ mass - radius

↔ mass - tidal deformability parameter



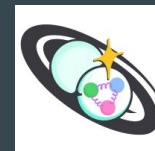
Nuclear matter properties



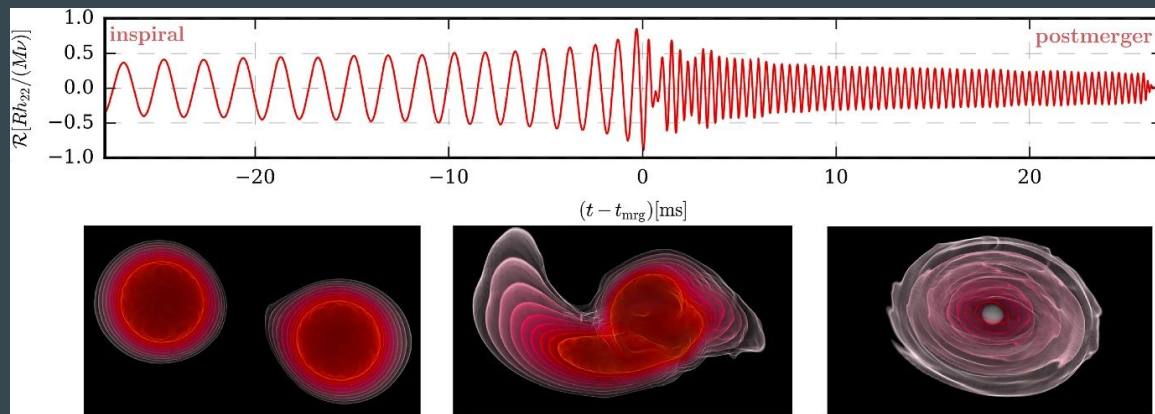
Gen. Rel.Grav. 53, 27 (2021)

- measure the system parameters (masses and tidal deformabilities)
- study the postmerger (future)

- multimessenger astrophysics
NMMA
- combine with nuclear information
[Nature 606, 276 (2022)]



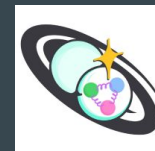
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Gen. Rel.Grav. 53, 27 (2021)

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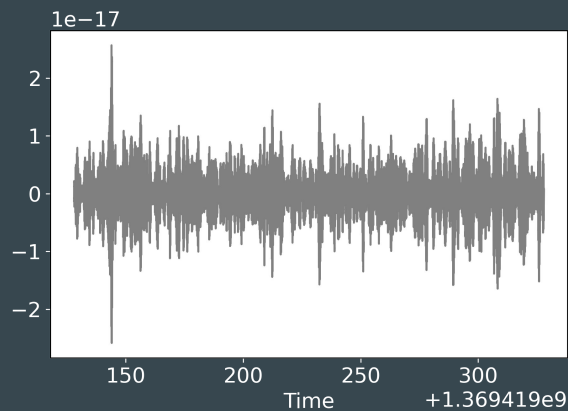
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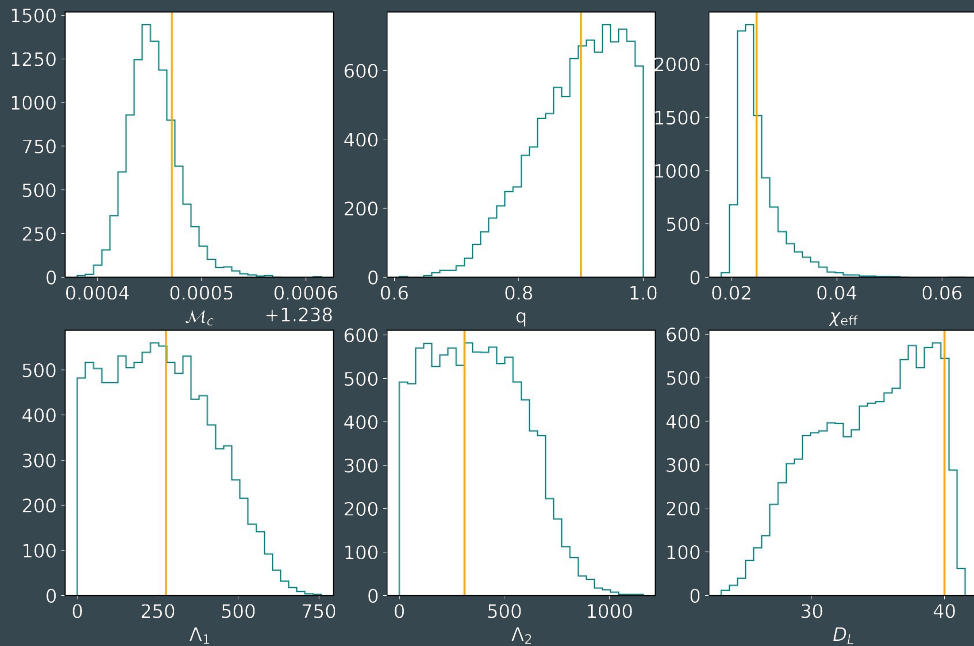
*How do we estimate the source
parameters?*

Parameter estimation

Source parameters \longleftrightarrow GW signal



Parameter estimation



We assume that from match filtering we know there is a signal

Parameter estimation

We need:

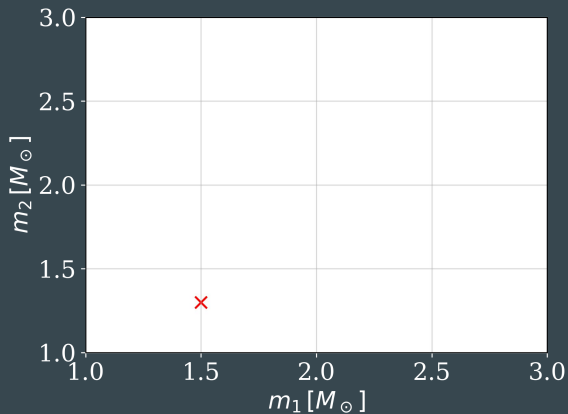
- The measured GW data
- A model H to describe our data, whose parameters we want to estimate (ex: the masses)
- An idea of the range in which the parameters lie (ex: for a NS, $m_1, m_2 \in [1, 3]M_\odot$)

Parameter estimation

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1. Pick some values for the parameters

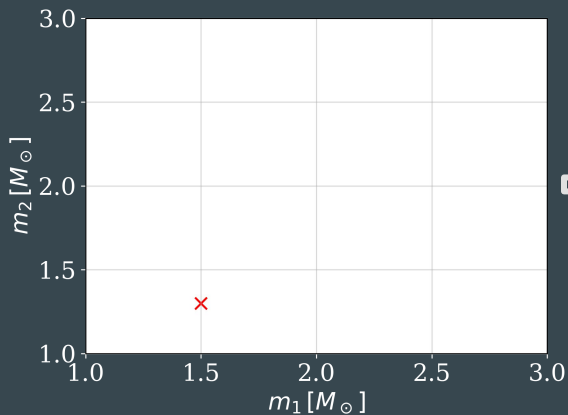


Parameter estimation

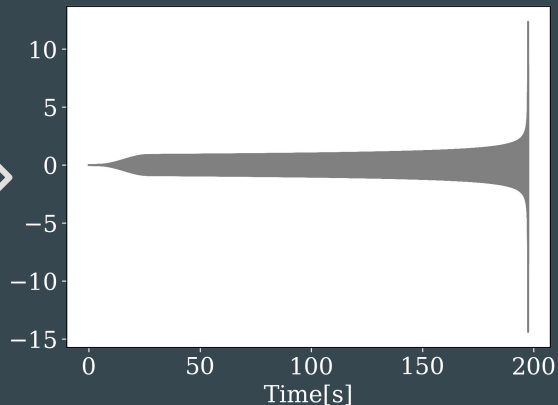
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2. Use the model to compute the GW waveform for those parameters values

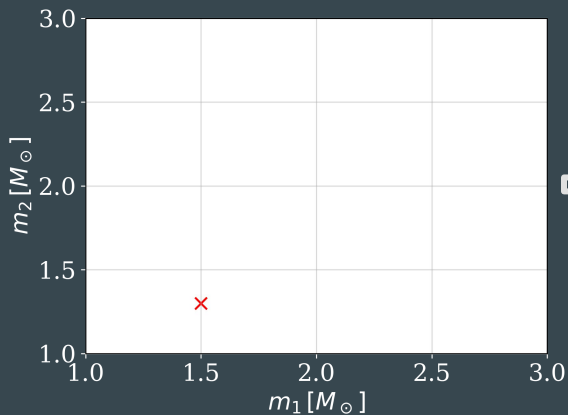


Parameter estimation

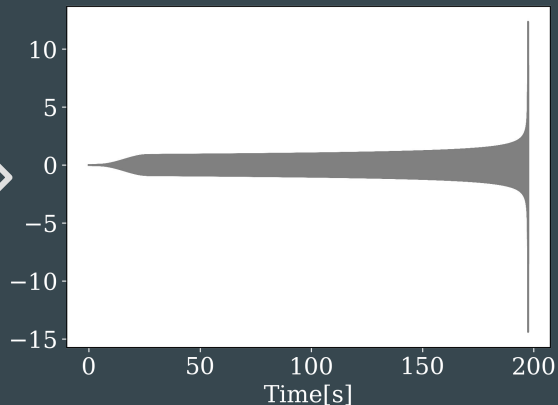
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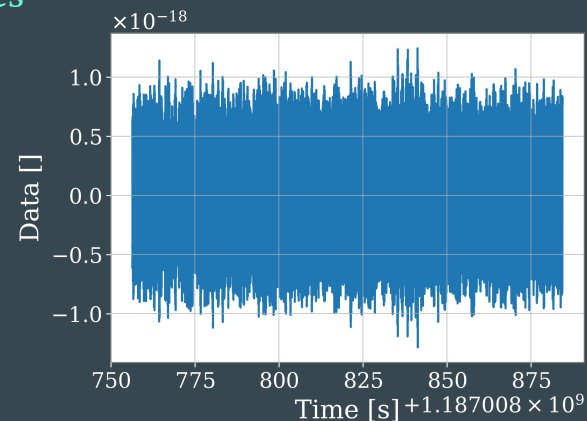
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3. Compare to GW data

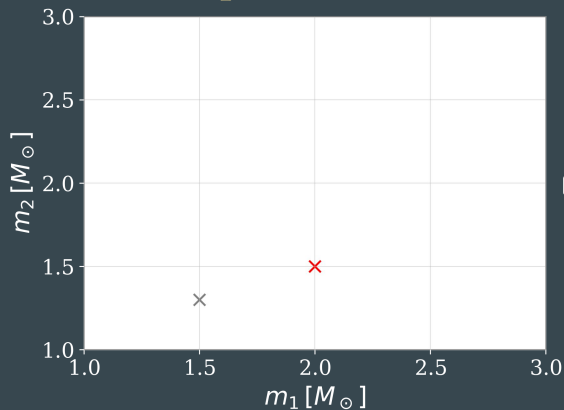


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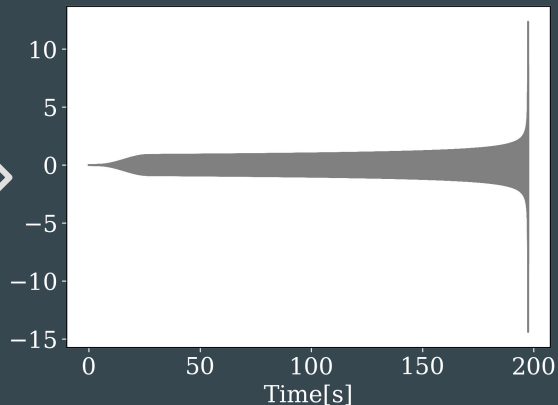
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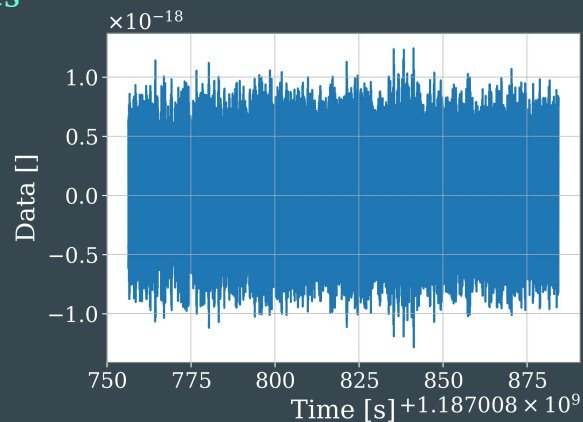
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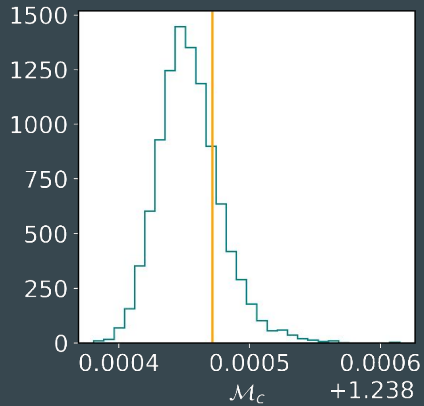
2. Use the model to compute the GW waveform for those parameters values



3. Compare to GW data



Parameter estimation - Bayesian inference



Bayes theorem:

$$p(d|\vec{\theta}, \mathcal{H}_s) \propto \exp\left[-\frac{1}{2}\langle d - h(\vec{\theta})|d - h(\vec{\theta})\rangle\right]$$

likelihood

prior

$$p(\vec{\theta}|\mathcal{H}_s, d) = \frac{p(d|\vec{\theta}, \mathcal{H}_s)p(\vec{\theta}|\mathcal{H}_s)}{p(d|\mathcal{H}_s)}$$

Posterior probability density

evidence

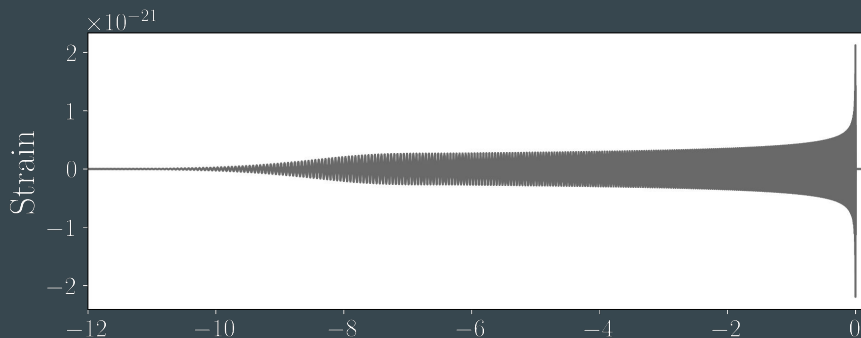
Evidence → Model comparison

$$\mathcal{B}_A^B = \frac{p(d|H_B)}{p(d|H_A)} = \frac{Z_B}{Z_A}$$

Parameter estimation - waveform models

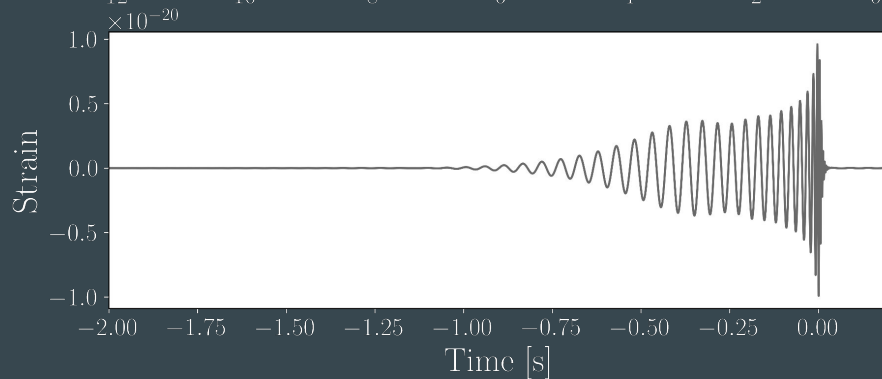
$$m_1 = 10 M_\odot$$

$$m_2 = 8 M_\odot$$



$$m_1 = 45 M_\odot$$

$$m_2 = 36 M_\odot$$



Parameters: masses, distance,
spins, inclination, sky location...

Accurate models needed: include tidal effects, spins, precession,
higher-order modes, eccentricity...

Waveform models

GW signal \Rightarrow solve Einstein Field Equations: $R_{\alpha\beta} - \frac{1}{2}Rg_{\alpha\beta} = \frac{8\pi G}{c^4}T_{\alpha\beta}$

Numerical Relativity \Rightarrow Computationally expensive

NRsurrogate models: directly interpolated on a set of pre-computed NR waveforms

✓ very accurate

✗ limited parameters space

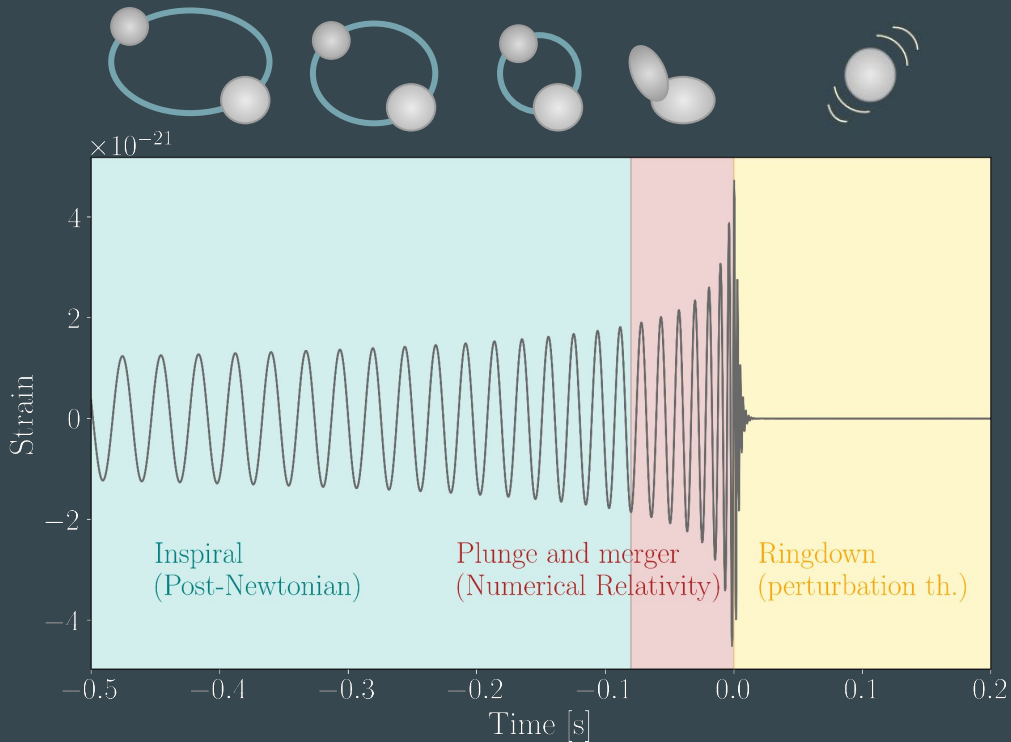


Approximations

Waveform models

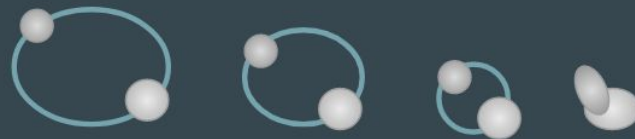
GW signal \Rightarrow solve Einstein Field Equations: $R_{\alpha\beta} - \frac{1}{2}Rg_{\alpha\beta} = \frac{8\pi G}{c^4}T_{\alpha\beta}$

Numerical Relativity \Rightarrow Computationally expensive \Rightarrow **Approximations**



Waveform models - Post Newtonian expansion

- GWs carry energy \rightarrow remove energy from the system \rightarrow orbit shrinks

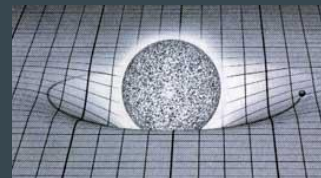


Energy balance:

$$\mathcal{F} = -\frac{dE}{dt}$$

GW luminosity (flux) Orbital energy loss

Waveform models - Post Newtonian expansion



➤ Curvature induced by the source

Inspiral: Assume binary components at large distance and with small velocities

Energy balance:
$$\mathcal{F}(v) = -\dot{E}(v)$$

expand both sides in power series of (v/c) : **post-Newtonian expansion**

Post-Newtonian (PN) expansion = *expansion in (v/c)* n-th PN order $\iff \mathcal{O}\left(\frac{v^{2n}}{c^{2n}}\right)$

compute energy and flux up to needed order

\implies compute the GW phase $h(t) = B(t)e^{i\Phi(t)}$

$$\dot{\phi}(t) = v^3 / M_{\text{tot}}$$

$$\dot{v} = -\mathcal{F}(v) / E'$$

Taylor family models

$$(E' = dE/dv)$$

Waveform models - Post Newtonian expansion

Currently: energy up to 4 PN, flux up to 4.5 PN

$$\begin{aligned} \mathcal{E}_4(v) = & -\frac{1}{2}\eta v^2 \left\{ 1 - \left(\frac{3}{4} + \frac{1}{12}\eta \right) v^2 - \left(\frac{27}{8} - \frac{19}{8}\eta + \frac{1}{24}\eta^2 \right) v^4 \right. \\ & + \left[-\frac{675}{64} + \left(\frac{34445}{576} - \frac{205}{96}\pi^2 \right) \eta - \frac{155}{96}\eta^2 - \frac{35}{5184}\eta^3 \right] v^6 \\ & + \left[-\frac{3969}{128} + \left(-\frac{123671}{5760} + \frac{9037}{1536}\pi^2 + \frac{896}{15}\gamma_E + \frac{448}{15}\ln(16v^2) \right) \eta \right. \\ & \left. \left. + \left(-\frac{498449}{3456} + \frac{3157}{576}\pi^2 \right) \eta^2 + \frac{301}{1728}\eta^3 + \frac{77}{31104}\eta^4 \right] v^8 + \mathcal{O}(v^{10}) \right\} \end{aligned}$$

Characteristic velocity of the binary

$$v = (\pi f_{\text{gw}} M_{\text{tot}})^{1/3}$$

Symmetric mass ratio

$$\eta = \frac{m_1 m_2}{(m_1 + m_2)^2}$$

$$\begin{aligned} \mathcal{F}_{4.5}(v) = & \frac{32}{5}\eta^2 v^{10} \left\{ 1 - \left(\frac{1247}{336} + \frac{35}{12}\eta \right) v^2 + 4\pi v^3 \right. \\ & - \left(\frac{44711}{9072} - \frac{9271}{504}\eta - \frac{65}{18}\eta^2 \right) v^4 - \left(\frac{8191}{672} + \frac{583}{24}\eta \right) \pi v^5 \\ & + \left[\frac{6643739519}{69854400} + \frac{16}{3}\pi^2 - \frac{1712}{105}\gamma_E + \left(\frac{41}{48}\pi^2 - \frac{134543}{7776} \right) \eta - \frac{94403}{3024}\eta^2 - \frac{775}{324}\eta^3 \right. \\ & \left. - \frac{856}{105}\ln(16v^2) \right] v^6 - \left(\frac{16285}{504} - \frac{214745}{1728}\eta - \frac{193385}{3024}\eta^2 \right) \pi v^7 \\ & + \left[-\frac{323105549467}{3178375200} + \frac{232597}{4410}\gamma_E - \frac{1369}{126}\pi^2 + \frac{39931}{294}\ln 2 - \frac{47385}{1568}\ln 3 + \frac{232597}{8820}\ln(v^2) \right. \\ & \left. + \left(-\frac{1452202403629}{1466942400} + \frac{41478}{245}\gamma_E - \frac{267127}{4608}\pi^2 + \frac{479062}{2205}\ln 2 + \frac{47385}{392}\ln 3 + \frac{20739}{245}\ln(v^2) \right) \eta \right. \\ & \left. + \left(\frac{1607125}{6804} - \frac{3157}{384}\pi^2 \right) \eta^2 + \frac{6875}{504}\eta^3 + \frac{5}{6}\eta^4 \right] v^8 \\ & + \left[\frac{265978667519}{745113600} - \frac{6848}{105}\gamma_E - \frac{3424}{105}\ln(16v^2) + \left(\frac{2062241}{22176} + \frac{41}{12}\pi^2 \right) \eta \right. \\ & \left. - \frac{133112905}{290304}\eta^2 - \frac{3719141}{38016}\eta^3 \right] \pi v^9 + \mathcal{O}(v^{10}) \left. \right\}, \end{aligned}$$

Post Newtonian expansion - GW phase

In frequency domain: $\tilde{h}(f) = A(f)e^{i\Psi(f)}$

GW phase: $\Psi(f) = 2\pi f t_c - \phi_c - \frac{\pi}{4} + \frac{3}{128\eta} v^{-5} \left[\sum_{k=0}^7 \phi_k(\eta) v^k \right]$

$$\begin{cases} v = (\pi f_{\text{gw}} M_{\text{tot}})^{1/3} \\ \eta = \frac{m_1 m_2}{(m_1 + m_2)^2} \end{cases}$$

➤ OPN: $\phi_0 = 1$ $\Psi(f)_{\text{OPN}} = \frac{3}{128\eta} v^{-5} \propto (\mathcal{M}_c \pi f)^{5/3}$

⇒ phase evolution dominated by chirp mass

Post Newtonian expansion - GW phase

In frequency domain: $\tilde{h}(f) = A(f)e^{i\Psi(f)}$

GW phase: $\Psi(f) = 2\pi f t_c - \phi_c - \frac{\pi}{4} + \frac{3}{128\eta} v^{-5} \left[\sum_{k=0}^7 \phi_k(\eta) v^k \right]$

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⇒ phase evolution dominated by chirp mass

$$\psi = \psi_{\text{PP}} + \psi_{\text{SO}} + \psi_{\text{SS}} + \psi_{\text{T}}$$

point-particle $\Psi(f)$

spin-orbit

spin-spin

tidal effects

Post Newtonian expansion - spins

➤ lowest spin correction at 1.5 PN (k=3):

$$\propto \underbrace{\frac{m_1 \chi_1 + m_2 \chi_2}{M_{\text{tot}}}}_{\chi_{\text{eff}}} - \frac{76 \eta (\chi_1 + \chi_2)}{226}$$

χ_i = dimensionless spin aligned with orbital angular momentum

changing mass ratio can mimic spin effects -> degeneracy

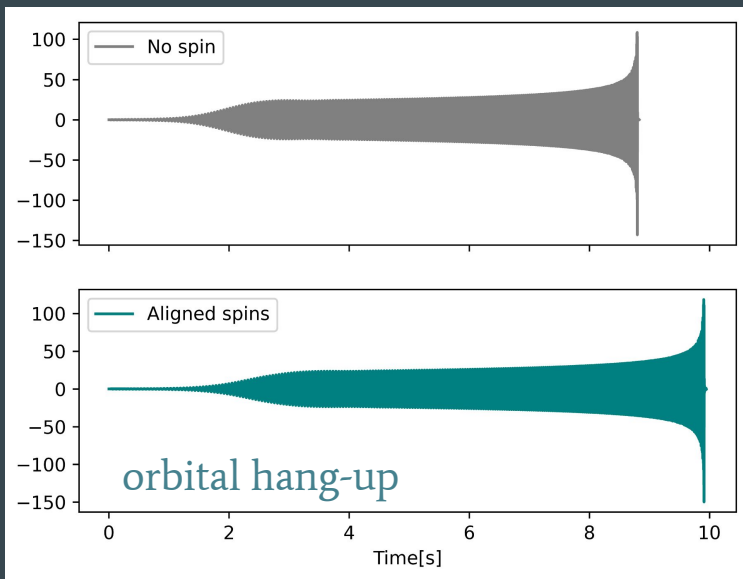
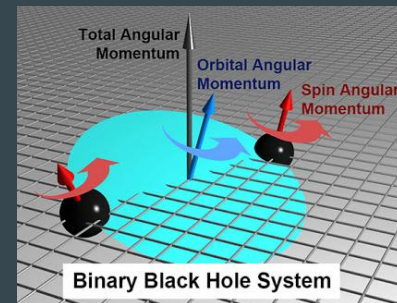
Post Newtonian expansion - spins

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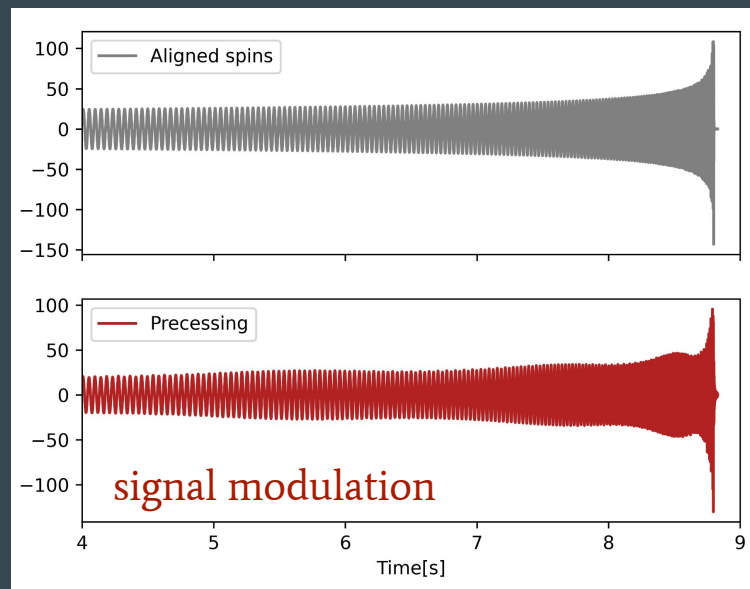
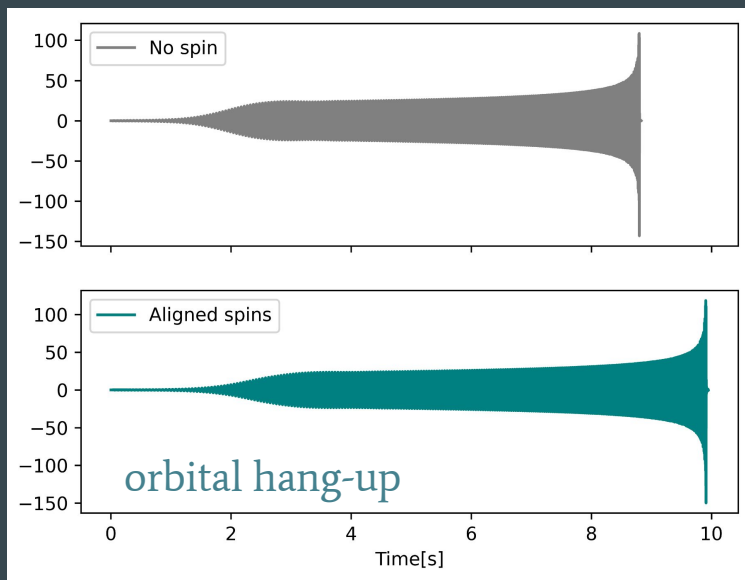
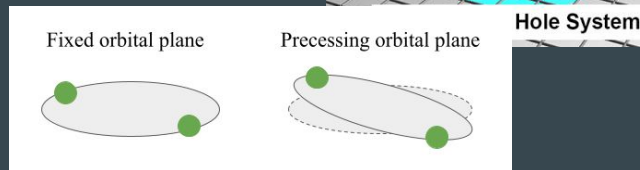
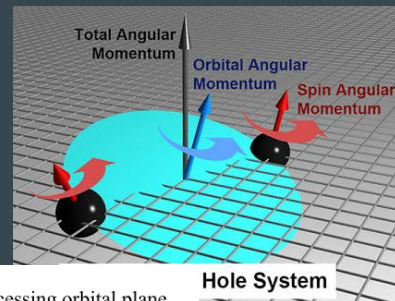


Post Newtonian expansion - spins

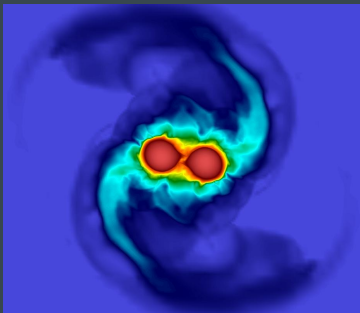
➤ lowest spin correction at 1.5 PN (k=3):

$$\propto \underbrace{\frac{m_1 \chi_1 + m_2 \chi_2}{M_{\text{tot}}}}_{\chi_{\text{eff}}} - \frac{76 \eta (\chi_1 + \chi_2)}{226}$$

changing mass ratio can mimic spin effects -> degeneracy



Post Newtonian expansion - tidal corrections



University of Birmingham

$$Q_{ij} = -\Lambda m^5 \mathcal{E}_{ij}$$

\downarrow quadrupole moment \rightarrow tidal field
 \searrow tidal deformability parameter:

$$\Lambda = \frac{2}{3} k_2 \left(\frac{R}{m} \right)^5$$

Tidal deformability
 -> orbital energy loss
 -> faster inspiral

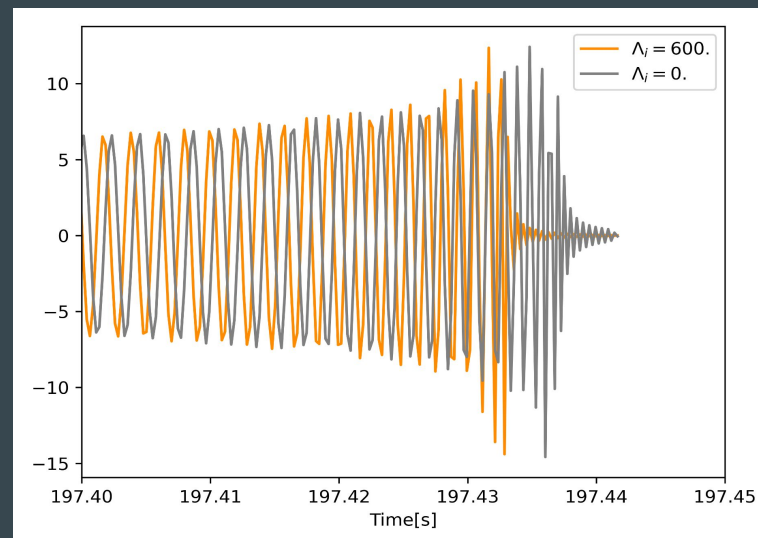
$$\psi(f) = \dots - \frac{39}{2} \tilde{\Lambda} v^{10}$$

From 5PN order ($\sim \mathcal{O}(100)$ Hz)

$$\tilde{\Lambda} = \frac{16}{13} \left[\left(1 + \frac{12m_2}{m_1} \right) \frac{\lambda_1^5}{M^5} + \left(1 + \frac{12m_1}{m_2} \right) \frac{\lambda_2^5}{M^5} \right]$$

mass-weighted tidal deformability

6PN order $\propto \Delta \tilde{\Lambda}$



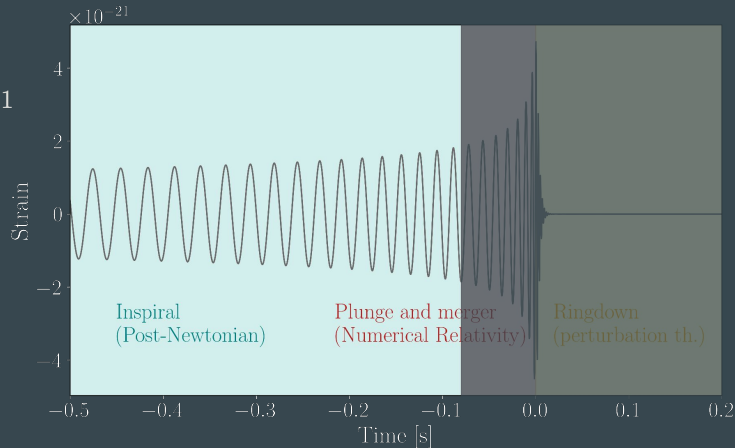
Waveform models - Post Newtonian

- PN approximation: expansion in (v/c)
- It explains why observe certain effects
- We can derive models directly from it (*TaylorF2*), analytical and fast

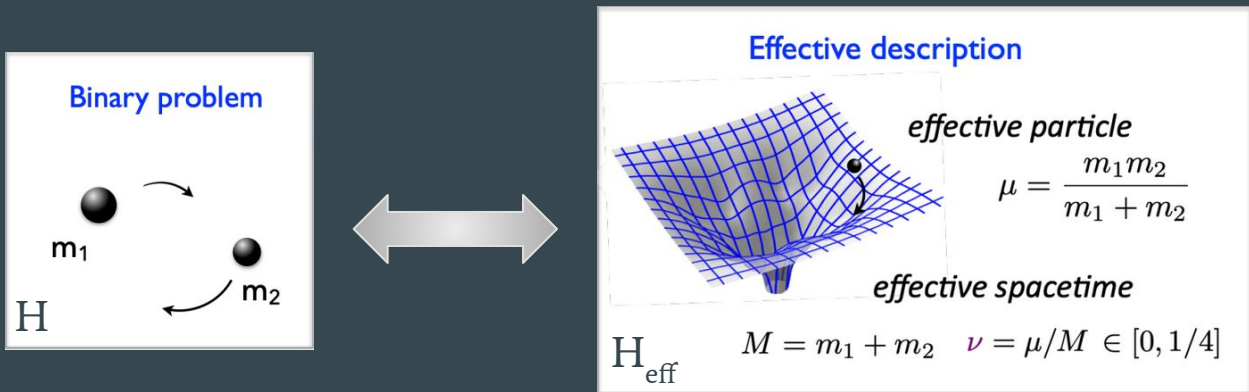
We measure chirp mass and mass-weighted tidal deformability, *not* radius!

but

- PN expansion valid until ISCO $f_{\text{ISCO}} = \left(6^{3/2}\pi M_{\text{tot}}\right)^{-1}$
- Taylor models for inspiral only
- Not extremely accurate (velocity increases)



Waveform models - Effective one body



Define a **map** between the Hamiltonian of the binary problem and the one of effective description:

$H_{\text{eff}} \Rightarrow$ **Concise resummation of PN results**

↓
Add terms corresponding to not-yet-computed PN orders and calibrate to NR simulations

Equations of motion

$H \Rightarrow$

$$\frac{dx^i}{dt} = \frac{\partial H}{\partial p_i}, \quad \frac{dp_i}{dt} = -\frac{\partial H}{\partial x^i} + \mathcal{F}_i$$

⇒ Gravitational wave signal

Waveform models - Effective one body

- Inspiral-merger-ringdown (IMR) model:

EOB dynamics until lighting + quasinormal ringdown modes

- Extend region of validity by calibrating to NR waveforms
- Examples: *SEOBNRv4PHM*, *SEOBNRv5*, *TEOBResumS*

✓ Very accurate
✗ Slow

Waveform models - Phenom

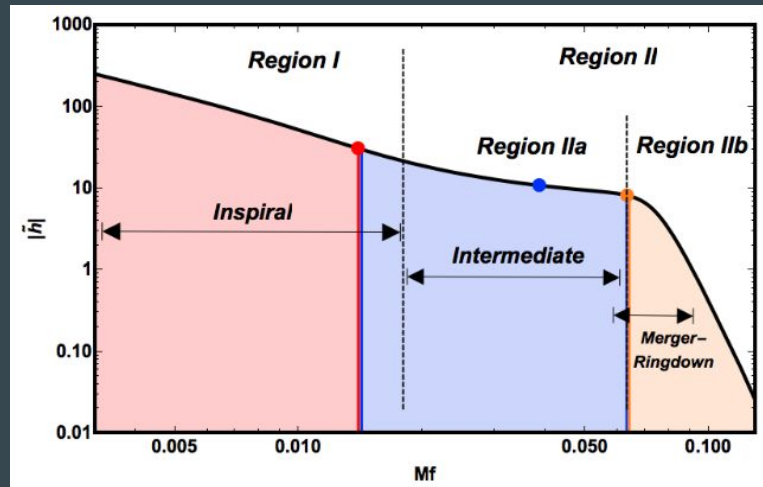
- Phenomenological ansatz with coefficients calibrated to NR waveforms
 - Three different regions
 - $\tilde{h}(f) = A(f)e^{i\Phi(f)}$
- Different phase and amplitude ansatz for each region

Example: phase *IMRPhenomD*

$$\Phi_{\text{Ins}} = \Phi_{\text{spin}}^{F2}(f; \Xi) + \frac{1}{\eta} \left(\sigma_0 + \sigma_1 f + \frac{3}{4} \sigma_2 f^{4/3} + \frac{3}{5} \sigma_3 f^{5/3} + \frac{1}{2} \sigma_4 f^2 \right)$$

$$\Phi_{\text{Int}} = \frac{1}{\eta} \left(\beta_0 + \beta_1 f + \beta_2 \ln(f) - \frac{\beta_3}{3} f^{-3} \right)$$

$$\Phi_{\text{MR}} = \frac{1}{\eta} \left[\alpha_0 + \alpha_1 f - \alpha_2 f^{-1} + \frac{4}{3} \alpha_3 f^{3/4} + \alpha_4 \tan^{-1} \left(\frac{f - \alpha_5 f_{\text{RD}}}{f_{\text{damp}}} \right) \right]$$



[Phys. Rev. D 93, 044007]

Full IMR waveform: require continuity

$\{\sigma_i, \beta_i, \alpha_i\}$ calibrated to hybrid waveforms

✓ Fast

Waveform models - effects of matter

$$\psi = \psi_{\text{PP}} + \psi_{\text{SO}} + \psi_{\text{SS}} + \psi_{\text{T}} \rightarrow \text{Tidal deformability}$$

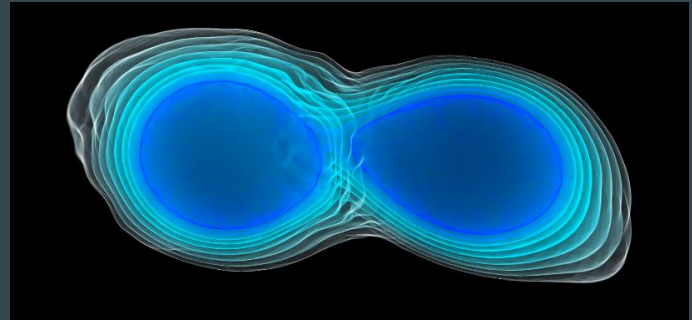
NRTidal: closed-form expression to describe tidal effects, calibrated to NR

Most recent: *NRTidalv3* [arXiv:2311.07456]

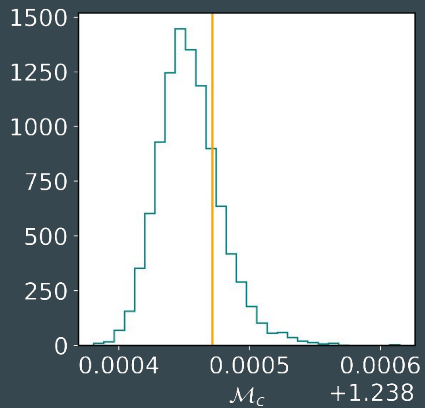
- different EoS
- also not-equal mass

Add phase corrections to existing BBH models

IMRPhenomD_NRTidalv3, *IMRPhenomXAS_NRTidalv3*, *IMRPhenomXP_NRTidalv3*,
SEOBNRv5_ROM_NRTidalv3



Parameter estimation - Bayesian inference



Bayes theorem:

$$p(\vec{\theta} | \mathcal{H}_s, d) = \frac{p(d | \vec{\theta}, \mathcal{H}_s) p(\vec{\theta} | \mathcal{H}_s)}{p(d | \mathcal{H}_s)}$$

Posterior probability density

likelihood

prior

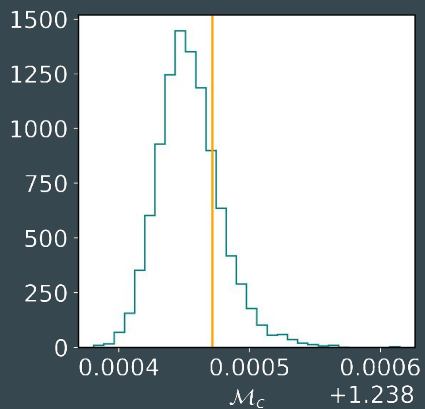
evidence

Likelihood and evidence are multidimensional integrals on the parameters space:

15 parameters for BBHs,
17 for BNSs



Parameter estimation - Bayesian inference



Bayes theorem:

$$p(\vec{\theta} | \mathcal{H}_s, d) = \frac{p(d | \vec{\theta}, \mathcal{H}_s) p(\vec{\theta} | \mathcal{H}_s)}{p(d | \mathcal{H}_s)}$$

Posterior probability density

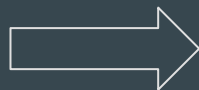
likelihood

prior

evidence

Likelihood and evidence are multidimensional integrals on the parameters space:

15 parameters for BBHs,
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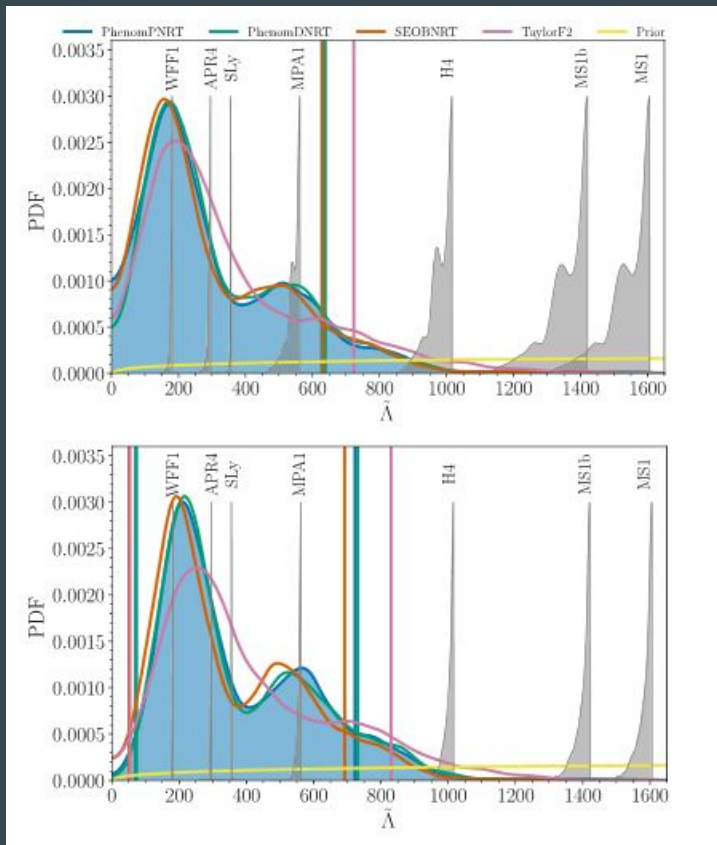


Stochastic sampling

- nested sampling (evidence)
- Markov Chain Monte Carlo (MCMC)

(see backup slides for details)

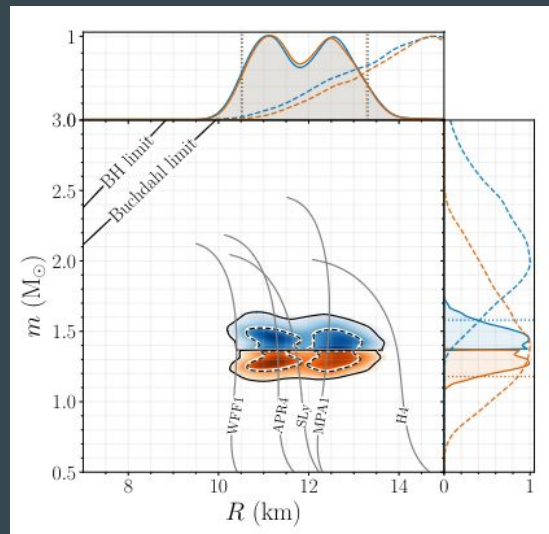
Neutron stars EOS with current detectors



Phys.Rev.X 9 (2019) 1, 011001

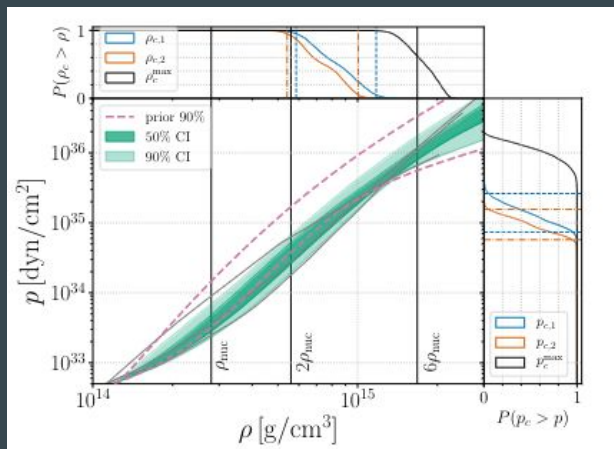
Two BNSs events:

- GW170817
(close, loud, multimessenger)
- GW190425 (far, no $\tilde{\Lambda}$ information)



Phys.Rev.Lett. 121 (2018) 16, 161101

Neutron stars EOS with current detectors

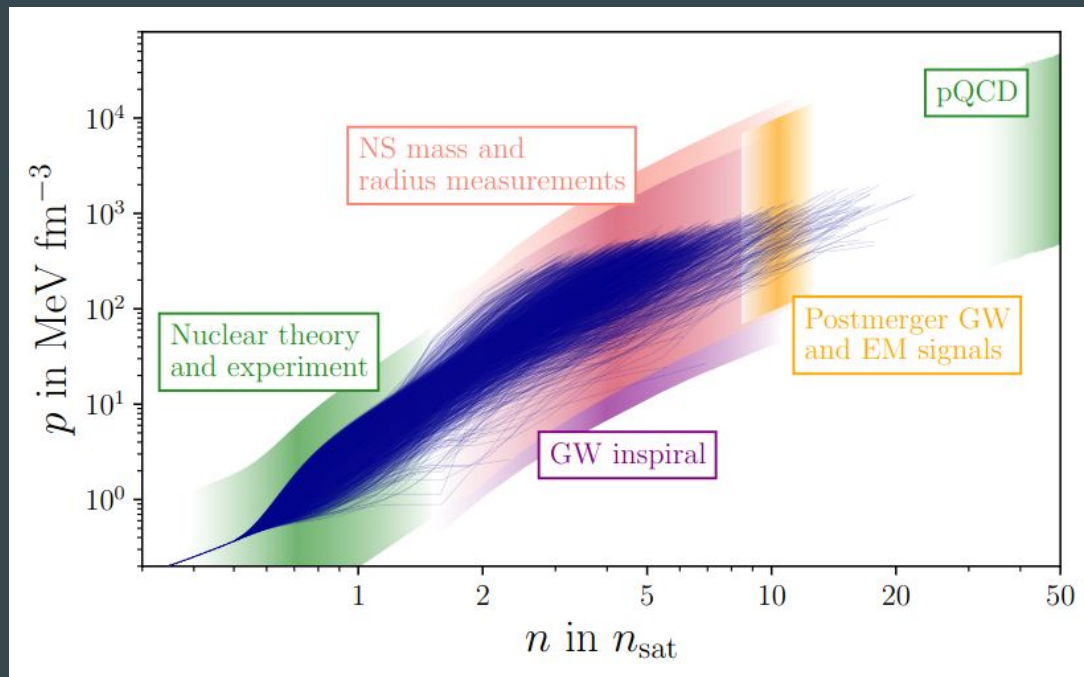


Phys.Rev.Lett. 121 (2018) 16, 161101

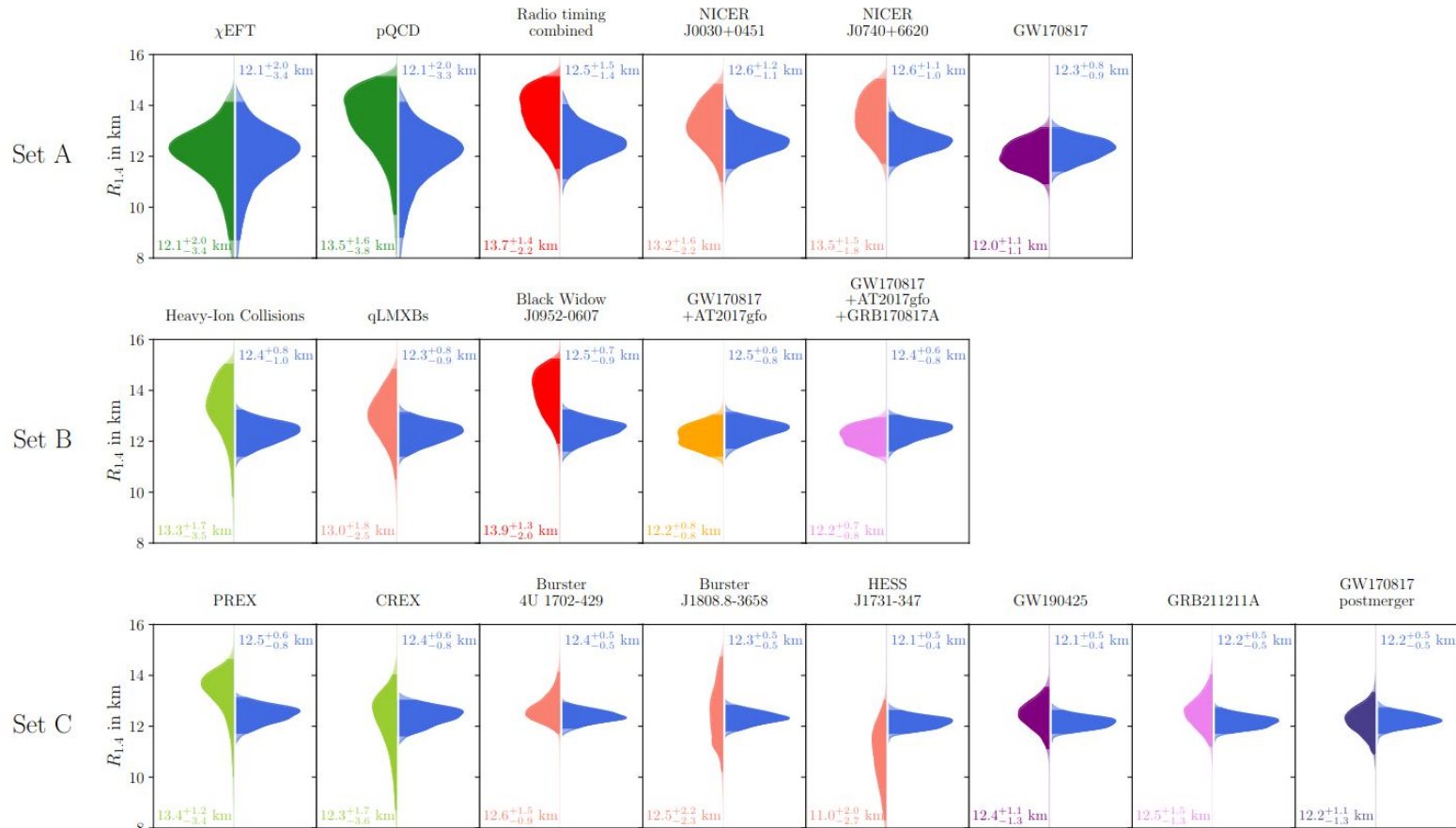
Sample on the EoS \rightarrow compute $\tilde{\Lambda}$
from mass and EoS \rightarrow use it in
the model

Combine with other sources of information!

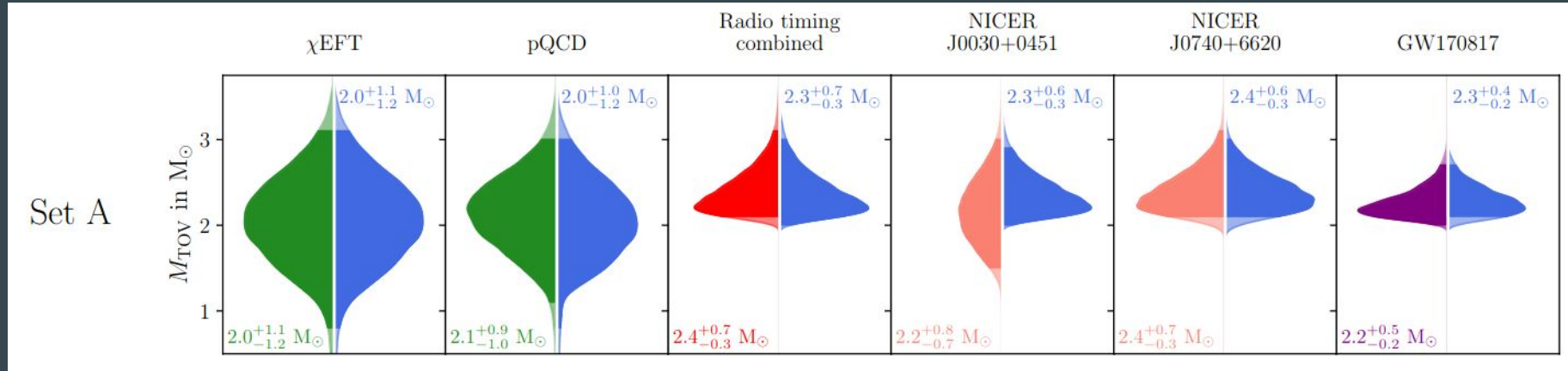
Overview: <https://arxiv.org/abs/2402.04172>



Combining information



Combining information



Web interface: https://multi-messenger.physik.uni-potsdam.de/eos_constraints/

Future detectors

Extended frequency band

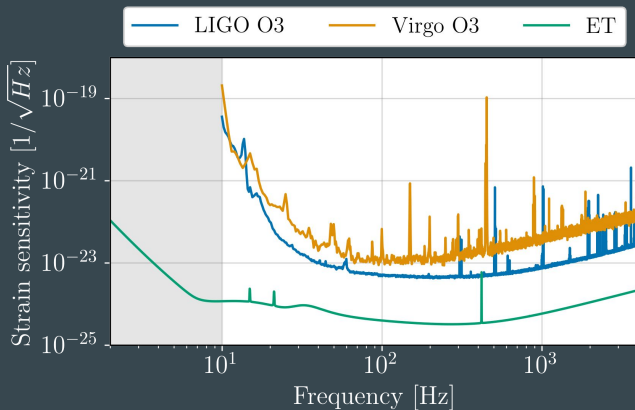
Problem: quantum noise

Radiation pressure noise at low frequencies

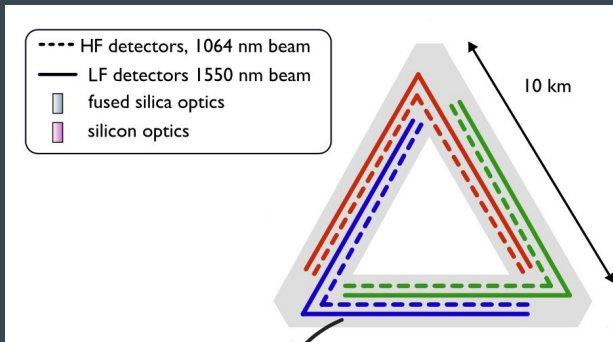
Shot noise at high frequencies

Decrease laser power

Increase laser power



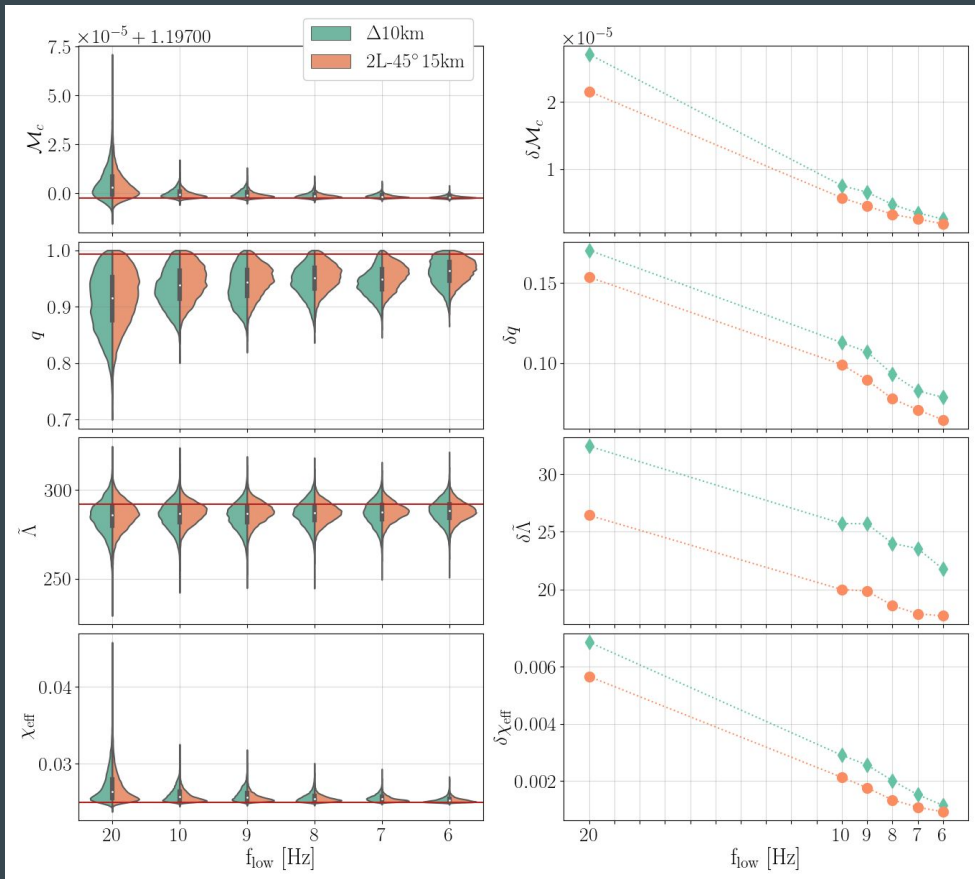
Einstein Telescope:
xylophone configuration



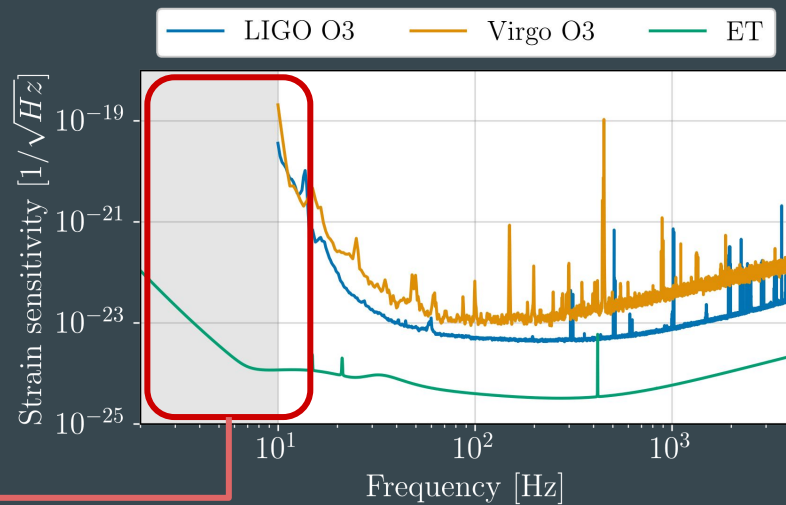
Plus (to improve low freqs):

- cryogenic
- underground

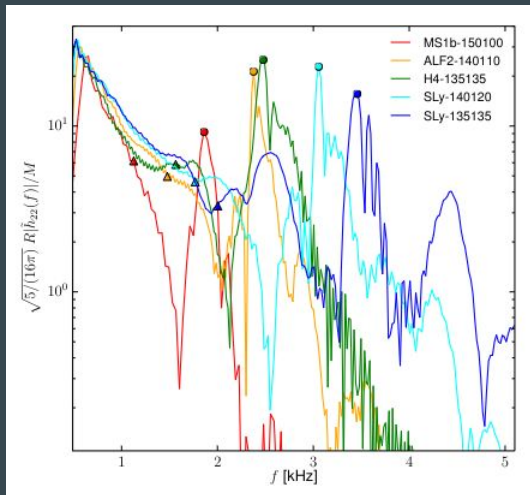
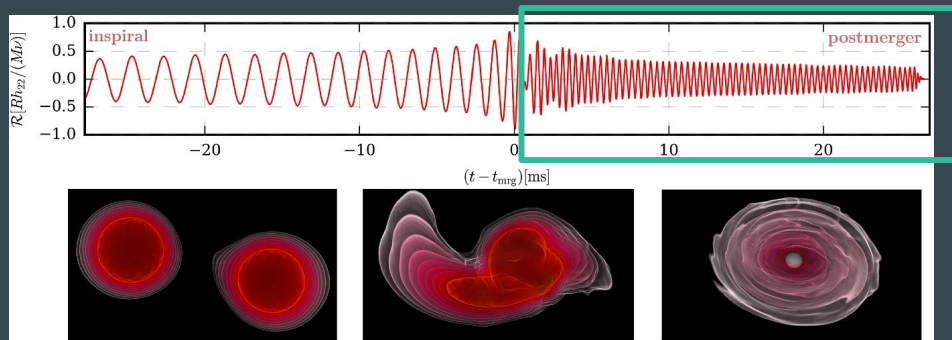
Future detectors



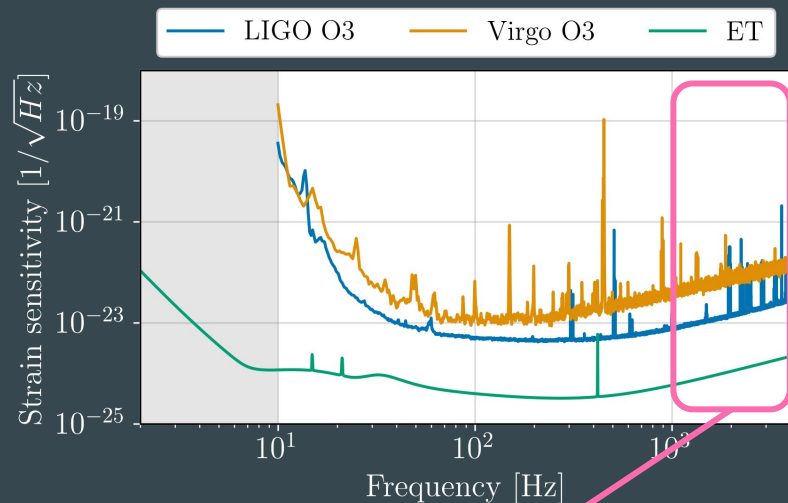
Extended frequency band



Future detectors

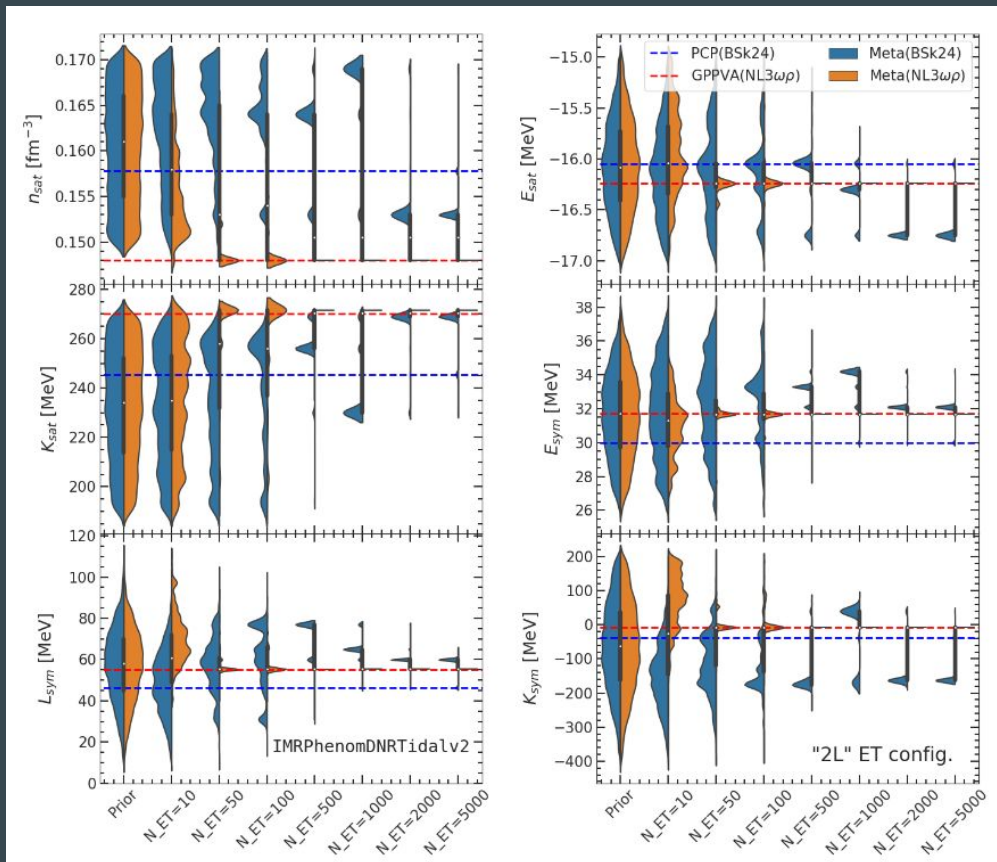


Extended frequency band



Postmerger
EoS in different temperature and
density regime

Future detectors - nuclear parameters

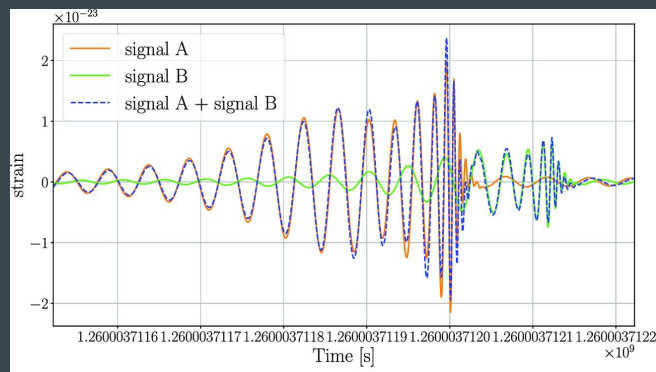
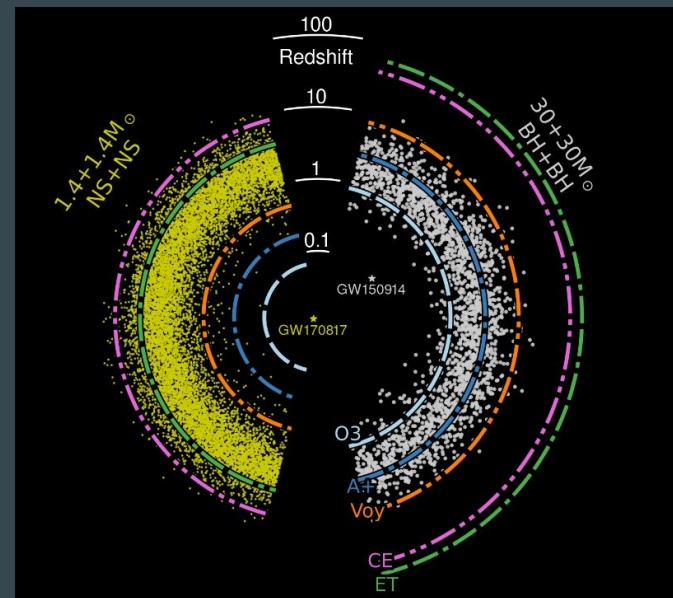


Posteriors for masses and lambdas + nuclear meta-model (expansion of energy per particle in density)

Future detectors - challenges

More events, louder, more time in band

- Overlapping signals
- *Systematics* (ex: waveform models)
- Computational issues!



Future detectors - challenges

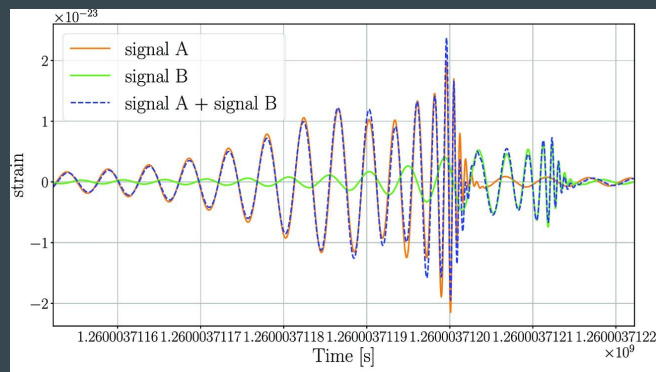
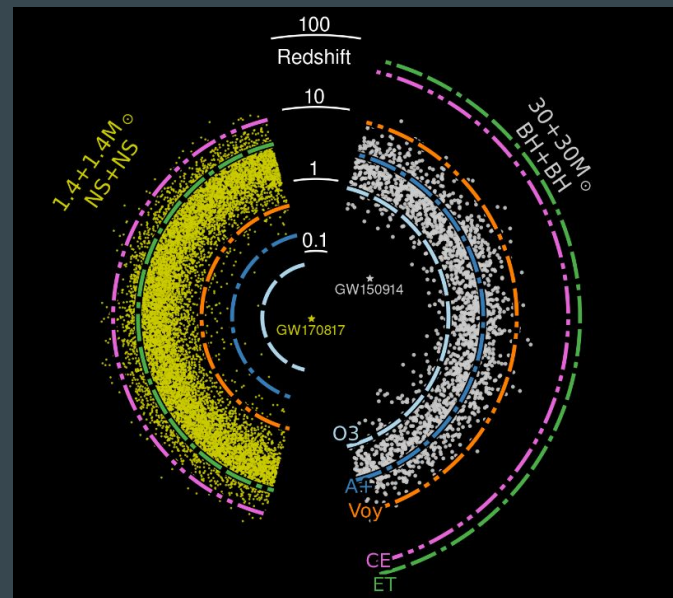
More events, louder, more time in band

- Overlapping signals
- *Systematics* (ex: waveform models)
- Computational issues!

BNS full parameter estimation with LIGO-Virgo takes weeks, with ET/CE months



- Faster waveform/samplers
- Machine learning to evaluate the posteriors
- **Approximate likelihood** (Reduced Order Quadratures, Relative Binning, Multibanding)



Take-away message

- From GWs we can measure chirp mass and mass-weighted tidal deformability
- We can use this information (possibly combined with other data) to get information about the EoS
- To get chirp mass and $\tilde{\Lambda}$ we use Bayesian inference:
 - Computationally very expensive -> nested or mcmc sampling
 - We need waveform models -> accurate but fast ($\mathcal{O}(10^6)$ evaluations per analysis)
- Next generation detectors:
 - More precise measurements
 - Postmerger (different temperature and density, but very complicated)
 - Various challenges

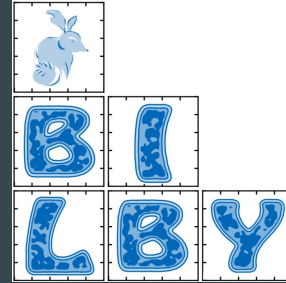


Software

Parameter estimation : bilby (python)

documentation: <https://lscsoft.docs.ligo.org/bilby/>

[google 'bilby software']



- nested: dynesty [documentation: <https://dynesty.readthedocs.io/en/latest/>]
- mcmc: bilby_mcmc

Bilby pipe, parallel bilby: automation and parallelization

Waveforms: LALsuite library (C), documentation

Useful tools: pyCBC (python), documentation: <https://pycbc.org/>

Tutorials

Python notebook:

- “*toy_model_sampling.ipynb*”: how PE works, example on a toy model (no GW)
- “*waveforms_and_mismatch.ipynb*”: how to generate different GW waveforms for different parameters, mismatch computation between different waveforms
- “*gw_pe.ipynb*”: example of PE on GW signals (simplified sampler settings)

Nested sampling

$$\mathcal{P}(\vec{\theta}) = \frac{L(\vec{\theta})\pi(\vec{\theta})}{Z(\vec{\theta})}$$



Estimate evidence, posterior as by-product

$$Z = \int L(\vec{\theta})\pi(\vec{\theta})d\vec{\theta}$$

Prior mass $X(\lambda) :=$ integral over the prior volume of all parameters with likelihood greater than a given value λ

$$X(\lambda) = \int_{\vec{\theta}:L(\vec{\theta})>\lambda} \pi(\vec{\theta})d\vec{\theta}$$

$$Z = \int_0^1 L(X)dX$$

discrete samples

$$Z \sim \sum_k g(L_k)\Delta X_k$$

Evidence \rightarrow **Model comparison**

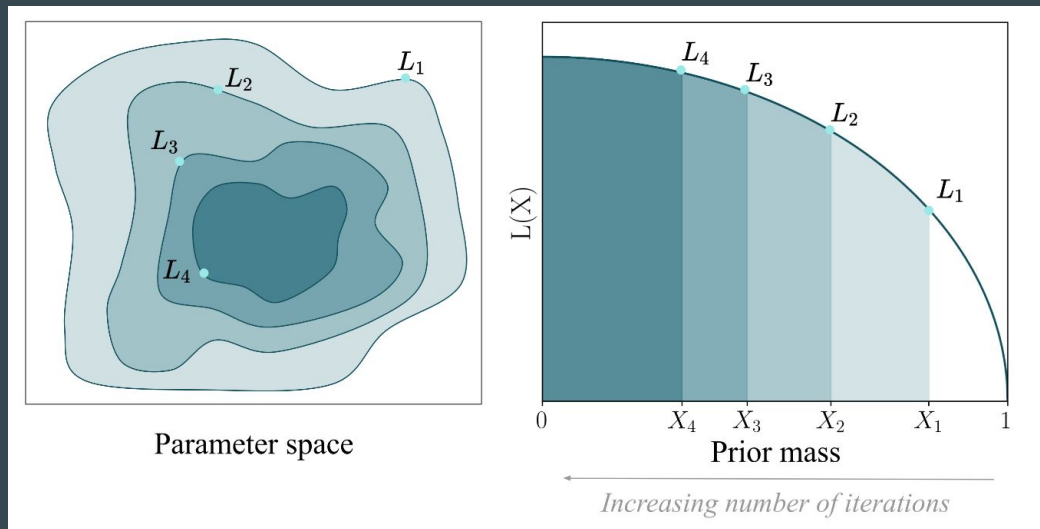
$$\mathcal{B}_A^B = \frac{p(d|H_B)}{p(d|H_A)} = \frac{Z_B}{Z_A}$$

Nested sampling

1. Sample N_{live} points $\{\vec{\theta}_1, \vec{\theta}_2, \dots, \vec{\theta}_{N_{\text{live}}}\}$ from the prior and compute their likelihood
2. For the point with the lowest likelihood L_k , estimate the corresponding prior mass X_k ; save $(\vec{\theta}_k, L_k, X_k)$, and remove $\vec{\theta}_k$ from the set of live points
3. Increase evidence by $Z_k = f(L_k, X_k)$
4. Randomly draw $\vec{\theta}_j$ from the prior, with $L(\vec{\theta}_j) > L_k$
5. Repeat until termination condition

We obtain a series of $(\vec{\theta}_k, L_k, X_k)$ with increasing L_k

$$\Rightarrow \mathcal{P}(\vec{\theta}_k) \simeq \frac{Z_k}{Z}$$



MCMC

Sample from the posterior distribution

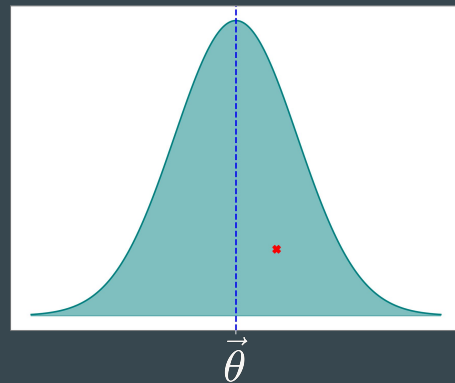
1. Sample $\vec{\theta}$ from the prior
2. Generate next point $\vec{\theta}'$ from a *proposal distribution* $Q(\vec{\theta}' | \vec{\theta})$
3. Accept the new point with probability

$$\alpha = \min \left\{ 1, \frac{L(\vec{\theta}')\pi(\vec{\theta}')Q(\vec{\theta} | \vec{\theta}')}{L(\vec{\theta})\pi(\vec{\theta})Q(\vec{\theta}' | \vec{\theta})} \right\}$$

accepted \rightarrow append $\vec{\theta}'$ to list of samples

rejected \rightarrow append $\vec{\theta}$ to list of samples

4. Repeat from the appended sample



Proposal distributions for GWs can be very complicated!