

Modeling Compact Mergers in the Era of Regular Gravitational-Wave Observations

Frank Ohme

Computational Challenges in Gravitational Wave Astronomy

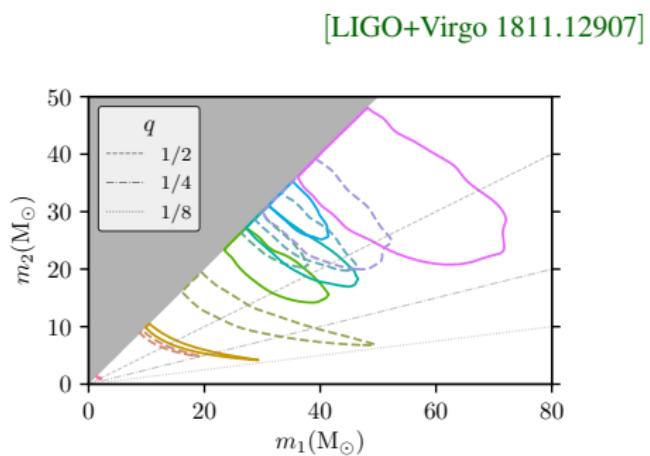
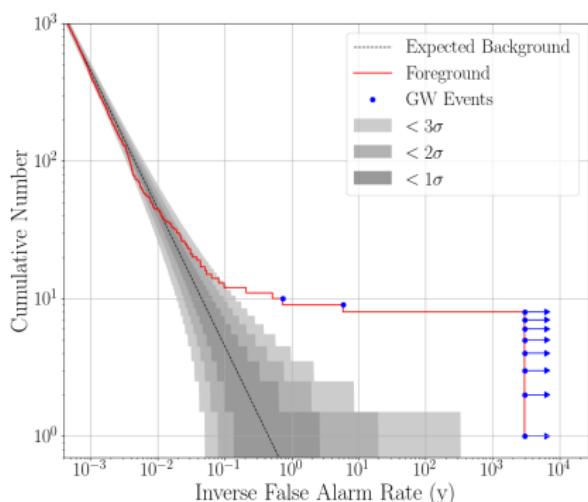


January 28, 2019

LIGO-G1900140-v2



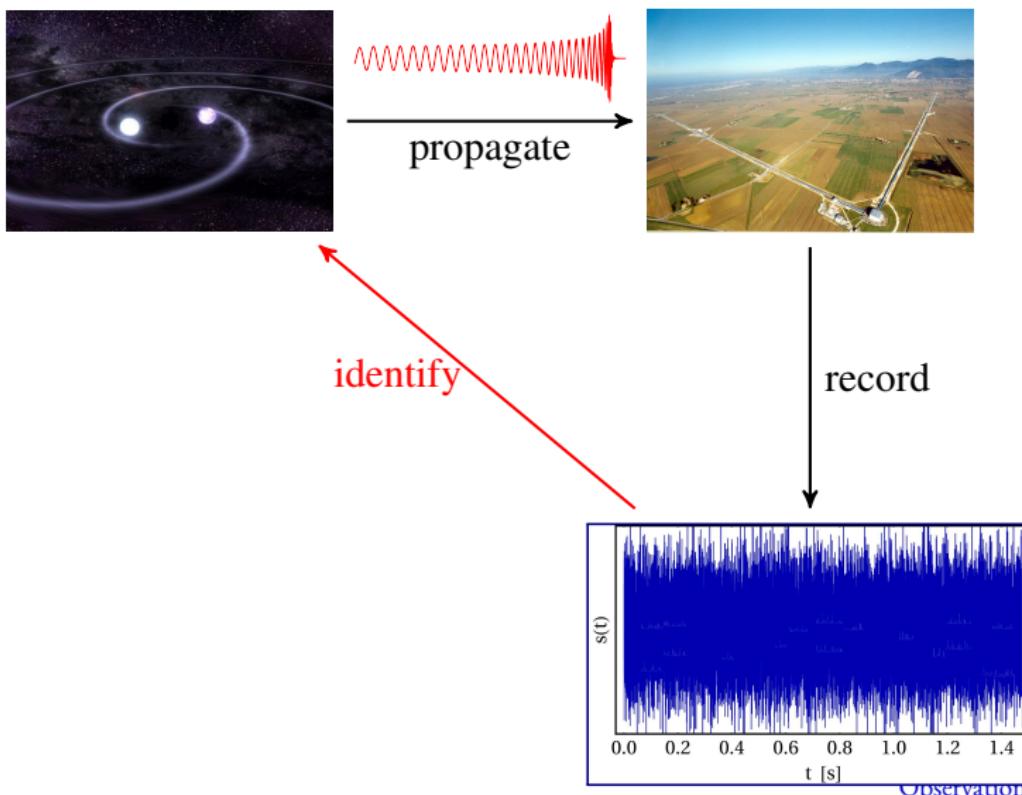
A Catalog of Compact Binary Mergers



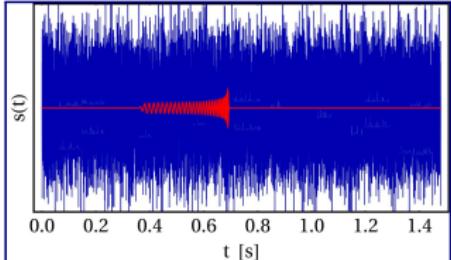
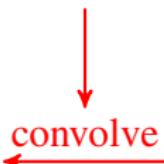
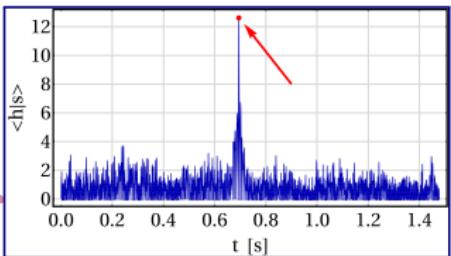
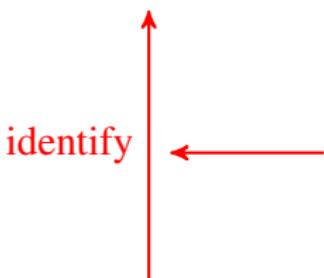
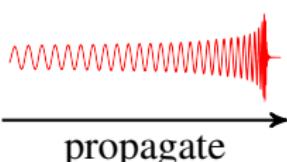
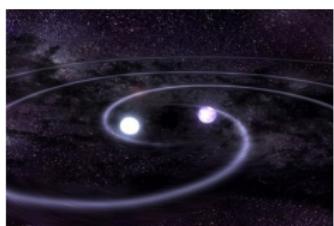
Data confronted with $> 10^7$ theoretically modeled signals



The Analysis Challenge



The Analysis Challenge

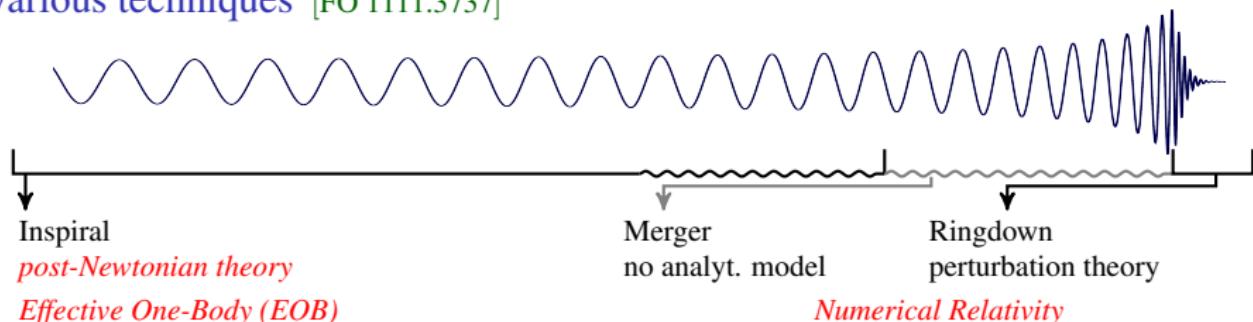


Observations &
Numerical Relativity

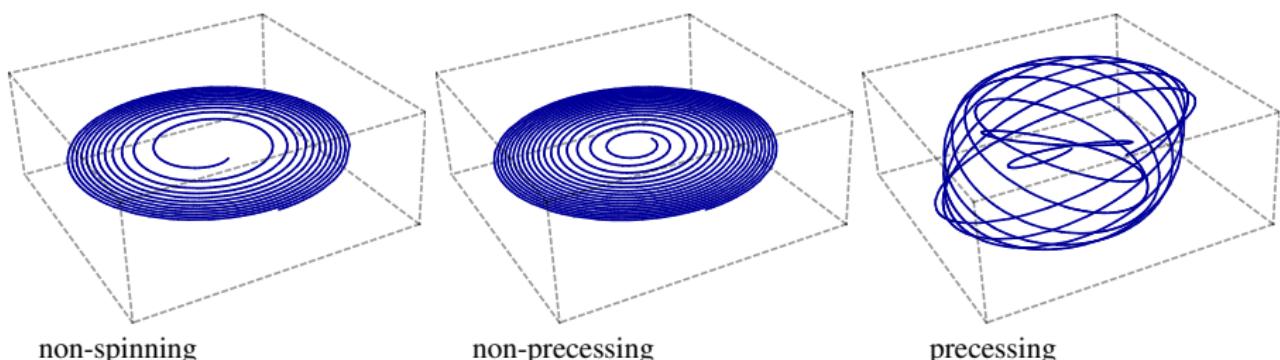


The challenge to produce complete waveform models

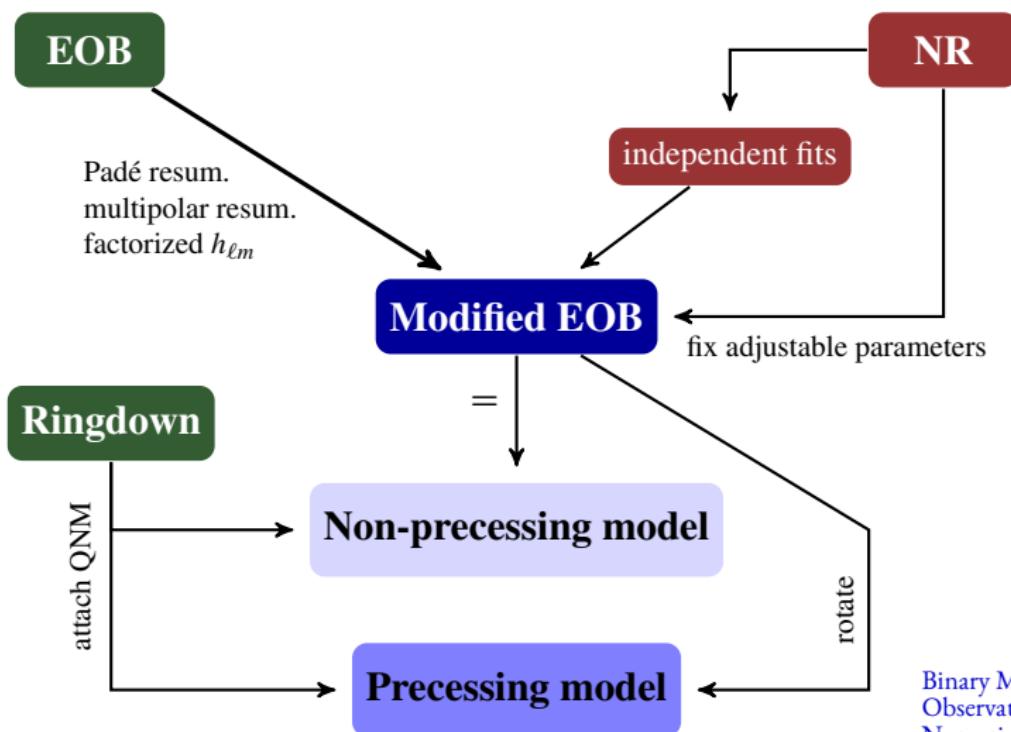
Various techniques [FO 1111.3737]



Various effects

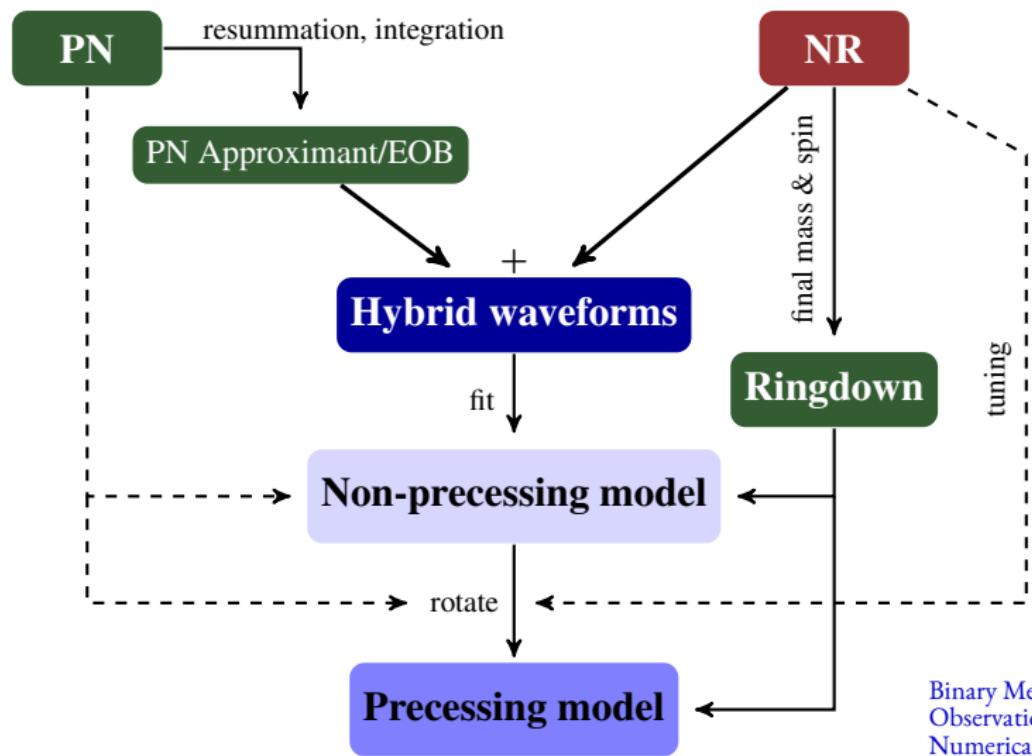


A (somewhat simplified) sketch of EOBNR



Phenomenological models

A (somewhat simplified) sketch of Phenomenological models



Binary Merger Observations & Numerical Relativity

Comparison

Complexity, Efficiency

Numerical Relativity

- Einstein's Equation
- Coupled PDEs
- Time integration

EOBNR

- Hamiltonian eq.
- ODEs
- Time integration

Phenomenological

- Fitting formulae
- Explicit closed form
- Frequency domain

complexity

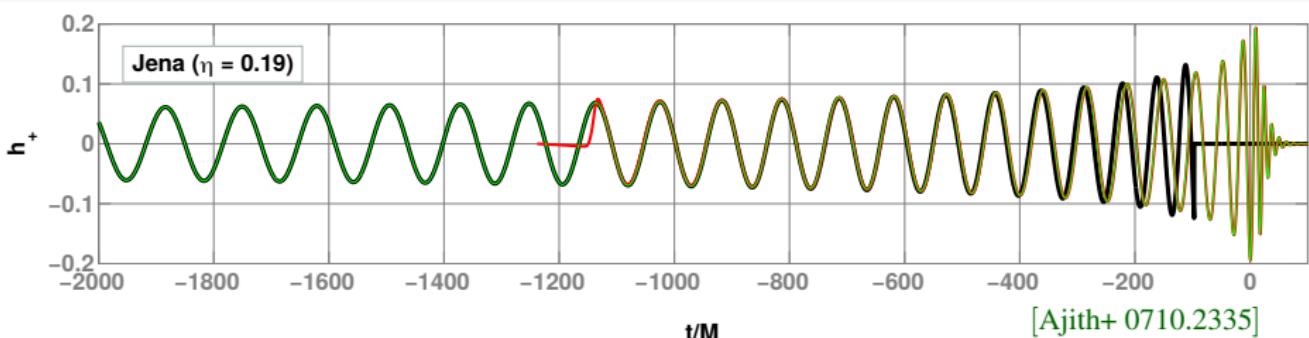


efficiency



PhenomA

The beginning: non-spinning signals



Analytical form

$$h(f) = \sum_{\ell,m}^{-2} Y_{\ell m} h_{\ell m} \approx {}^{-2} Y_{22} h_{22} = A(f) e^{i\Psi(f)}$$

$$\Psi(f) = 2\pi f t_0 + \varphi_0 + \sum_{k=0}^7 \psi_k f^{(k-5)/3}$$

$$A(f) = C \begin{cases} (f/f_m)^{-7/6} & \text{if } f < f_m \\ (f/f_m)^{-2/3} & \text{if } f_m \leq f \leq f_{RD} \\ \omega \mathcal{L}(f) & \text{if } f_{RD} \leq f < f_{cut} \end{cases}$$

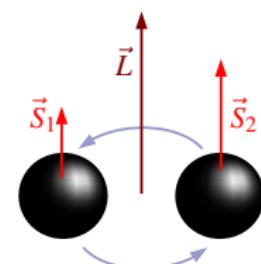
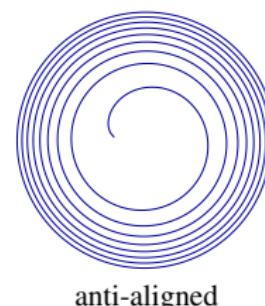
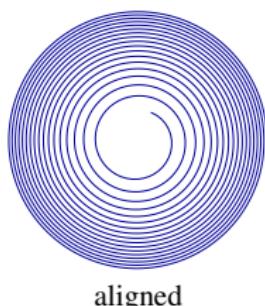
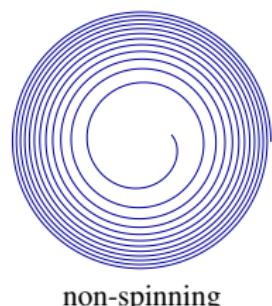
Restrictions

- no spins
- dominant harmonic
- no eccentricity
- 7 parameter space points
(mass ratio ≤ 4)

Numerical Relativity

PhenomB+C

Spinning, non-precessing signals



Additions

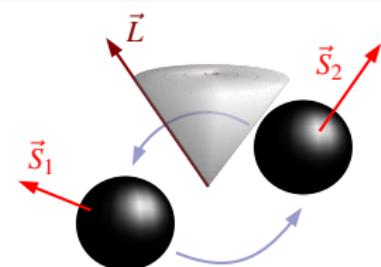
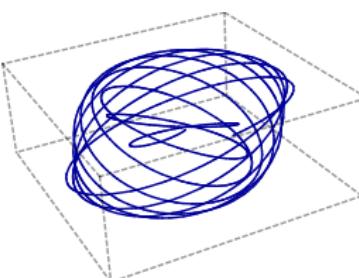
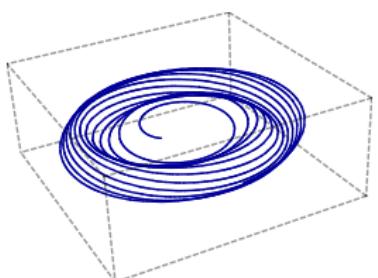
- Dominant spin effect: $\chi_{\text{eff}} = \frac{m_1 \chi_1 + m_2 \chi_2}{m_1 + m_2}$
- Extreme mass-ratio limit (PhenB)
[Ajith..FO+ 0909.2867]
- Fourier-domain hybridization (PhenC)
[Santamaría, FO+ 1005.3306]
- Smooth (tanh) transitions

Restrictions

- no precession
- dominant harmonic
- no eccentricity
- 24 NR signals
(mass ratio ≤ 4 ,
 $|\chi_{\text{eff}}| \leq 0.85$)

Precessing signals

[Hannam..FO+ 1308.3271]



Separation of complex dynamics

- Precession dominated by χ_p (larger in-plane spin component)

[Schmidt, FO, Hannam 1408.1810]

- Full signal = non-precessing \times rotation

$$h_{\ell m}^{\text{prec}} = e^{-im\alpha} \sum_{|m'| \leq \ell} e^{im'\epsilon} d_{m'm}^2(-\iota) h_{\ell m'}^{\text{np}}$$

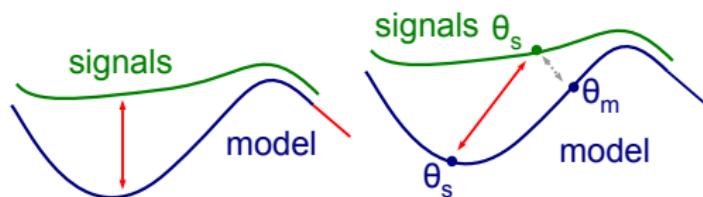
[Schmidt, Hannam, Husa 1207.3088]

Restrictions

- dominant spin effects
- dominant harmonic
- no eccentricity
- single-spin precession, not NR-tuned

Accuracy Concerns

Signal geometry



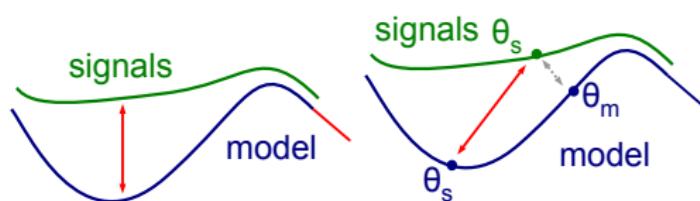
$$\langle h_1 | h_2 \rangle = 4 \operatorname{Re} \int \frac{h_1(f) h_2^*(f)}{S_n(f)} df$$

$$O = \frac{\langle h_1 | h_2 \rangle}{\sqrt{\langle h_1 | h_1 \rangle \langle h_2 | h_2 \rangle}}$$

- signals missed: $(\max O)^3$
- biased characterization at SNR 10:
 $1 - O > 0.5\%$

Accuracy Concerns

Signal geometry



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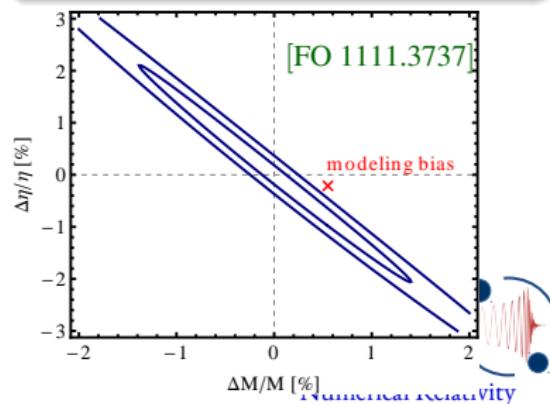
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Results

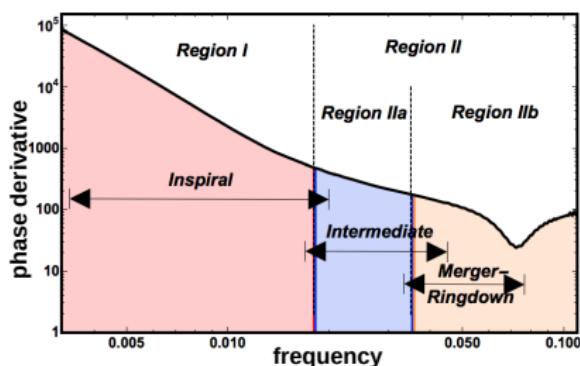
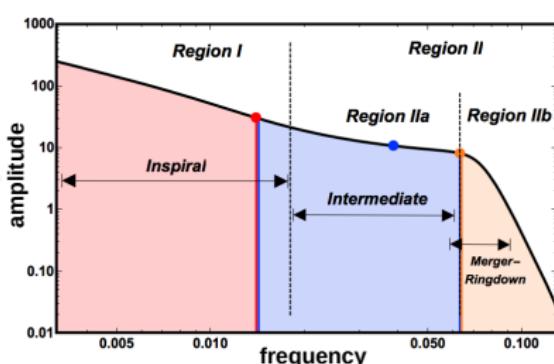
[FO+ 1107.0996]

origin	$1 - O$
Hybridization	< 0.02%
NR	< 0.1%
Interpolation	$\lesssim 3\%$
PN inspiral	$\sim \mathcal{O}(10\%)$



Overhaul of the fit

[Husa..FO+ 1508.07250, Khan..FO+ 1508.07253]



Novelties

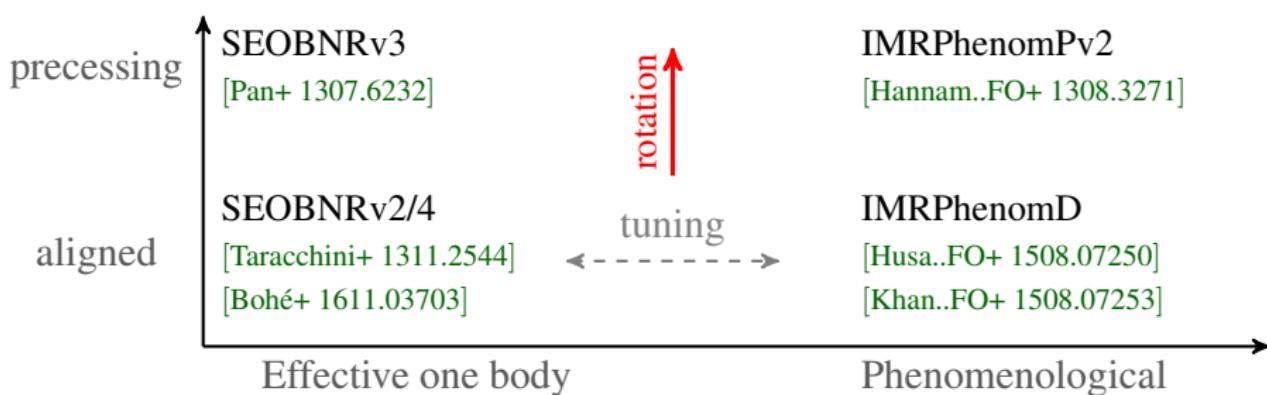
- EOB hybrids, PN-inspired fit
- two spins in inspiral and ringdown
- C^1 continuous, step transitions
- vastly improved fit with 17 independent phenomenological parameters

Restrictions

- precession as before
- dominant harmonic
- no eccentricity
- 19 NR signals
(mass ratio ≤ 18 ,
 $|\chi_{\text{eff}}| \leq 0.85/0.98$)

Current models

Main models used in most recent GW observations

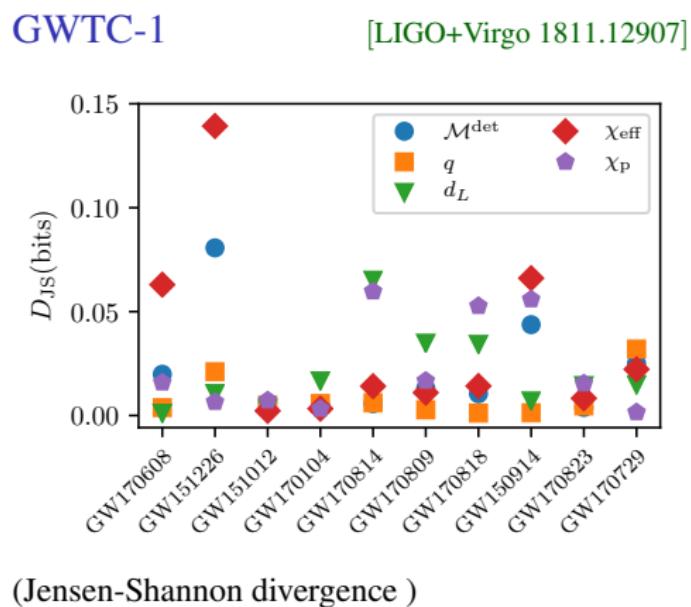
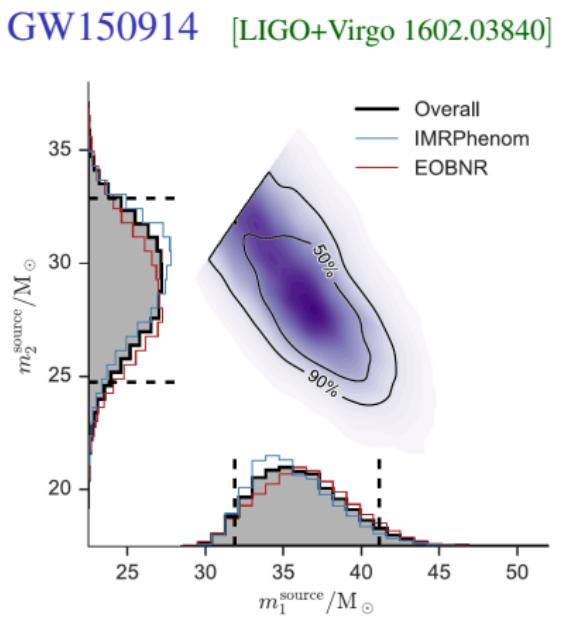


Notes (some details later in this talk)

- More flavors, including optimized versions, of those models are in use
- Alternatives exist that avoid construction of analytical model altogether

Impact of model differences

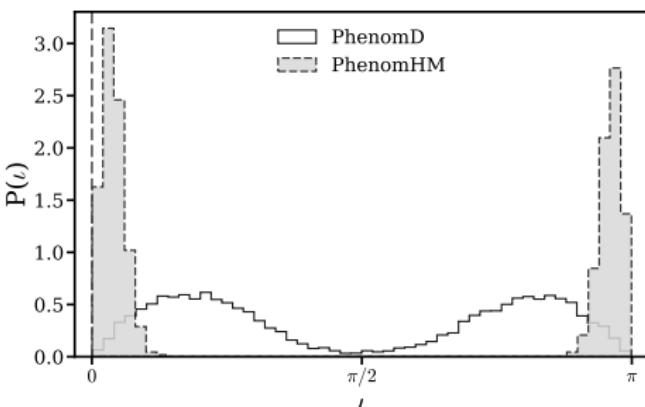
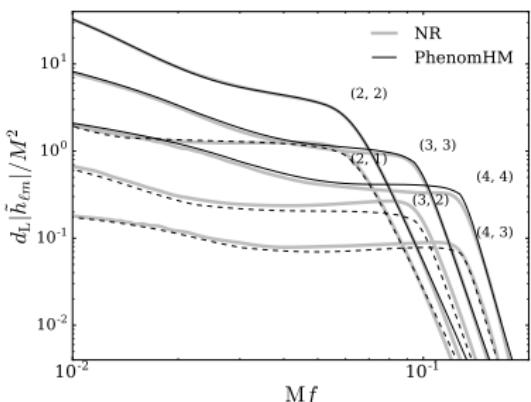
Observation + model \rightarrow posterior distribution



Note: Sources have been in the easiest-to-model part of the parameter space.

Adding higher multipoles

[London..FO+ 1708.00404]



A simple mapping

$$h(f) = \sum_{\ell m} -2 Y_{\ell m} h_{\ell m} = \sum_{\ell m} -2 Y_{\ell m} A_{\ell m} e^{i\Psi_{\ell m}}$$

$$A_{\ell m}(f) = \beta_{\ell m}(f) A_{22}(f_{22})$$

$$\Psi_{\ell m}(f) = \Psi_{22}(f_{22}) + \Delta_{\ell m}$$

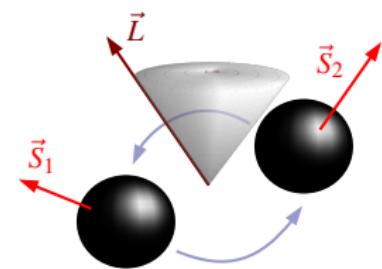
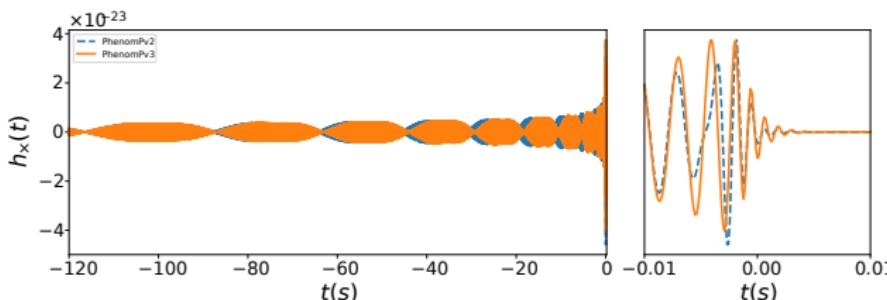
f_{22} : linear transition from $(2f/m)$ to $f_{\ell m}^{\text{RD}}$

Restrictions

- no eccentricity
- no additional NR tuning
- no mode mixing in ringdown

Improved precession

[Khan..FO+ 1809.10113]



Analytical solution for precession

- Analytical solution of orbit-averaged spin dynamics [Kesden+ 1411.0674]
- Multiple scale analysis using $T_{\text{prec}} \ll T_{rr}$
→ closed-form evolution of precession angles [Chatzioannou+ 1606.03117, 1703.03967]

Restrictions

- no eccentricity
- inspiral precession, no NR tuning

What's next?

Ongoing developments

Precessing higher modes

- Combine analytical precession with higher multipole mapping
[Khan+]
- Tune precession angles to NR
[Hamilton+]

Eccentric model

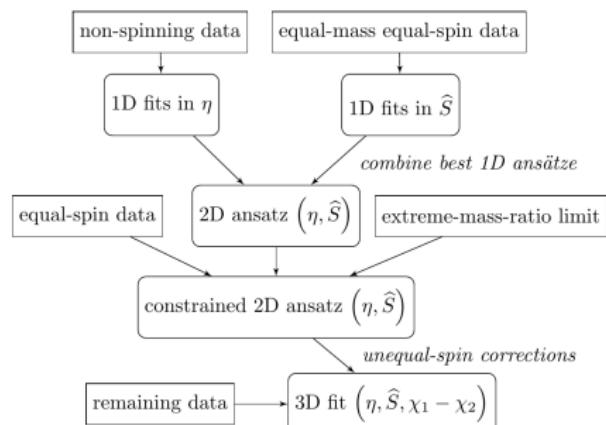
- 70 eccentric NR waveforms eccentricities ≤ 0.5 , mass ratio ≤ 4 , non-precessing
- PN+NR hybrids

[Ramos Buades, Husa, Haney]

PhenomX

[Pratten, Husa+]

- Automated NR processing
- Test particle Kerr dynamics
- Higher-multipole hybridization and tuning



[Jiménez-Forteza+, 1611.00332]

Complexity and Efficiency again

Numerical Relativity

Effective-One-Body

Phenomenological

complexity



Complexity and Efficiency again

Numerical Relativity

Effective-One-Body

Phenomenological

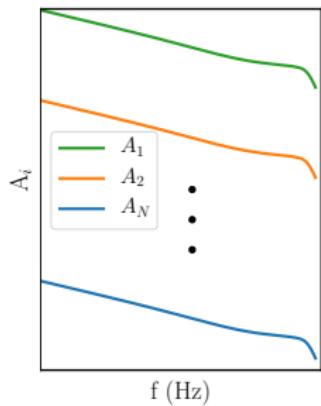
complexity

efficiency



Basic approach

Surrogate models



Basis representation

$$A_i \approx \sum_{k=1}^n c_{ik} \hat{A}_k \quad (n < N)$$

Interpolation

$$A(\boldsymbol{\theta}) \approx \sum_{k=1}^n \tilde{c}_k(\boldsymbol{\theta}) \hat{A}_k$$



Basis construction

Example: Singular Value Decomposition

$$A = U\Sigma V^T$$

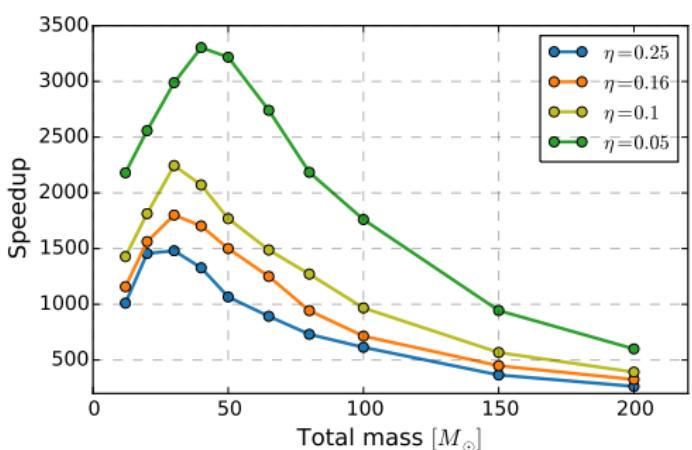
- A : gravitational-wave amplitudes or phases
- V : singular (basis) vectors
- Σ : diagonal matrix of singular values
- $U\Sigma$: projection coefficients

Application

Successfully implemented in combination with **tensor spline interpolation** in

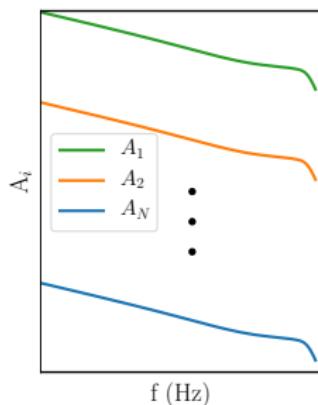
`SEOBNR_ROMv2/4`

[Pürrer 1512.02248]



Enhancing the waveforms

Enhancing models on the way

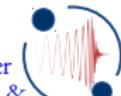


Basis representation

$$A_i \approx \sum_{k=1}^n c_{ik} \hat{A}_k \quad (n < N)$$

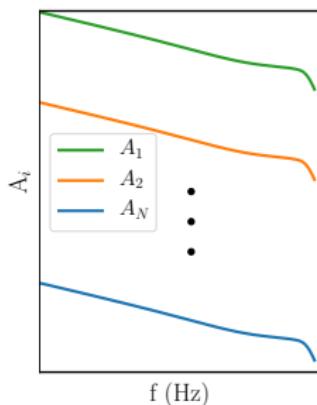
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Enhancing the waveforms

Enhancing models on the way



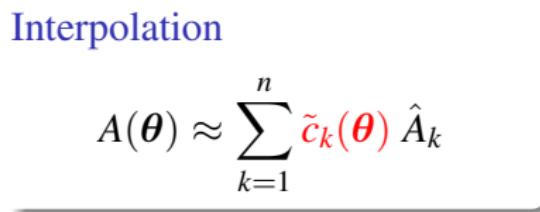
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New information

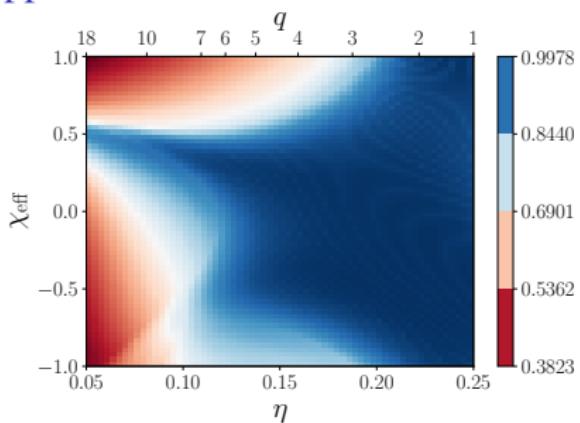


Enhancing the waveforms

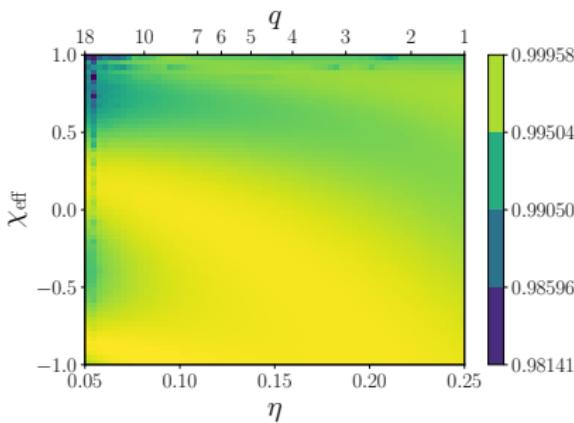
Proof of principle: PhenomB → PhenomD

[Setyawati, FO, Khan, 1810.07060]

Approximate basis



Enriched model



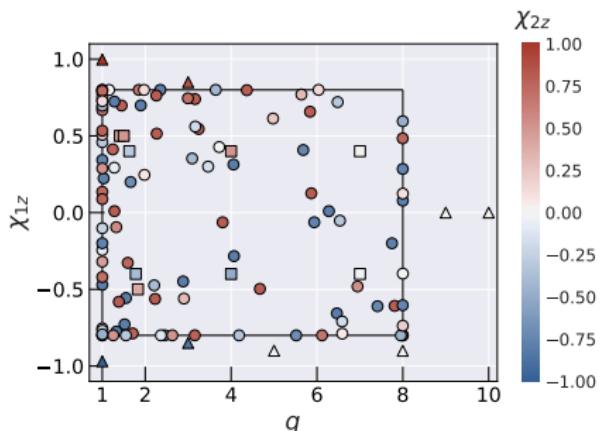
- Enriching of approximate model through few(er) accurate models
- No manual re-tuning
- Idea first sketched by [Cannon+ 1211.7095]

Greedy basis

Optimal placement: greedy basis

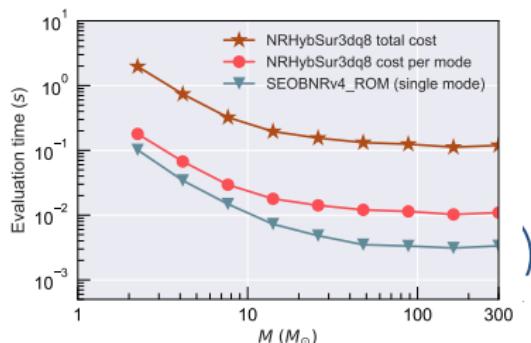
Greedy strategy

- ① Project test signals onto basis
 - ② Add signal with highest deviation from its basis projection to the basis
 - ③ Repeat until projection error sufficient
- [Field+ 1101.3765]



Variants for NR

- Use PN as proxy to find greedy points in parameter space
[Blackman+ 1701.00550, Varma+ 1812.07865]
- Use Gaussian Process Regression to estimate error
[Doctor+ 1706.05408]



Conclusion

- So far, waveform models for compact binaries lived up to the challenge of gravitational-wave astronomy thanks to a combinations of many techniques (**NR, analytical information, reduced-order interpolation**)
- More frequent observations → efficient models (simplified, hierarchical, optimized)
- Observations with higher signal-to-noise → accurate, complex models
- Need ways to efficiently incorporate new NR data
- **How can we gain confidence in a surprising measurement?**
(Keep independent, alternative modeling approaches.)
- Promising early developments regarding tidal effects, eccentricity, alternative theories

