

# LISA parameter estimation with time-domain response

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# Motivation for using time-domain response in LISA

- ▶ Parameter estimation and searches use *matched filtering* defined in Fourier domain

$$(d|h) = 4\text{Re} \int_0^\infty \frac{\tilde{d}(f)\tilde{h}^*(f)}{S_n(f)} df$$

- ▶ For ground-based detectors (LIGO), the response is constant in time

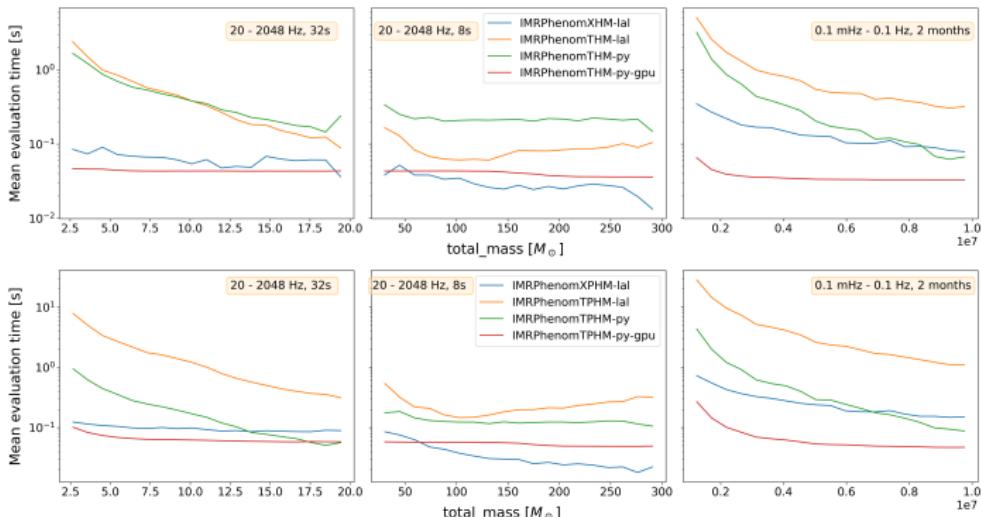
$$F_+ h_+(t) + F_\times h_\times(t)$$

- ▶ For space-based (LISA), response of each arm is time-dependent

$$y_{slr} = (x_+(t)(h_+(t_s) - h_+(t_r)) + x_\times(t)(h_\times(t_s) - h_\times(t_r)) / (1 - \mathbf{k} \cdot \mathbf{n})$$

- ▶ Time delay interferometry (TDI) reduces the laser frequency noise produced by deviations from the equilateral triangular shape
- ▶ In Fourier domain, LISA response requires the development of *transfer functions* both for amplitude and phase
  - ▶ Aligned-spin case EOBNRv2HM (S. Marsat+, 2021)
  - ▶ Precessing case IMRPhenomXPHM (G. Pratten+, 2023)
- ▶ More complicated effects like eccentricity or memory can be difficult to incorporate in the transfer function formalism
- ▶ In time domain, there is no theoretical challenge, but a computational challenge: need for a fine time sampling so the Fast Fourier transform algorithm works

- ▶ Analytical expressions for amplitude and phase as a function of time
- ▶ Include the effects of subdominant harmonics 22, 21, 33, 44, 55
- ▶ Precession prescriptions
  - ▶ Post-Newtonian Euler angles (NNLO, single-spin)
  - ▶ Multi-Scale Analysis (MSA, double-spin)
  - ▶ Numerical evolution of spins PN equations (SpinTaylor)
- ▶ Reimplemented in python (phenomxpy)
  - ▶ numpy vectorization
  - ▶ numba just-in-time compilation
  - ▶ cupy GPU support
  - ▶ Parallelization of time array, for one waveform
  - ▶ Efficient calculation of polarizations from rotating frames (M. Boyle+, 2014)
  - ▶ 20 mode (M. Roselló, 2024)
  - ▶ Extension to eccentricity in progress (M. de Lluc Planas, A. Ramos-Buades)



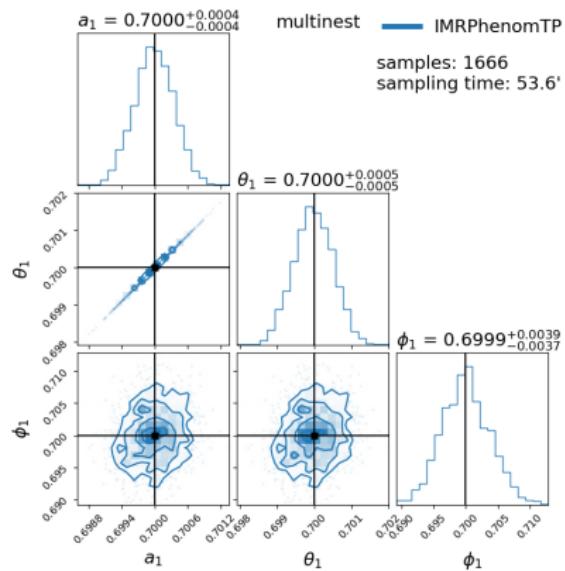
# LISA parameter estimation pipeline

## Parameter estimation pipeline

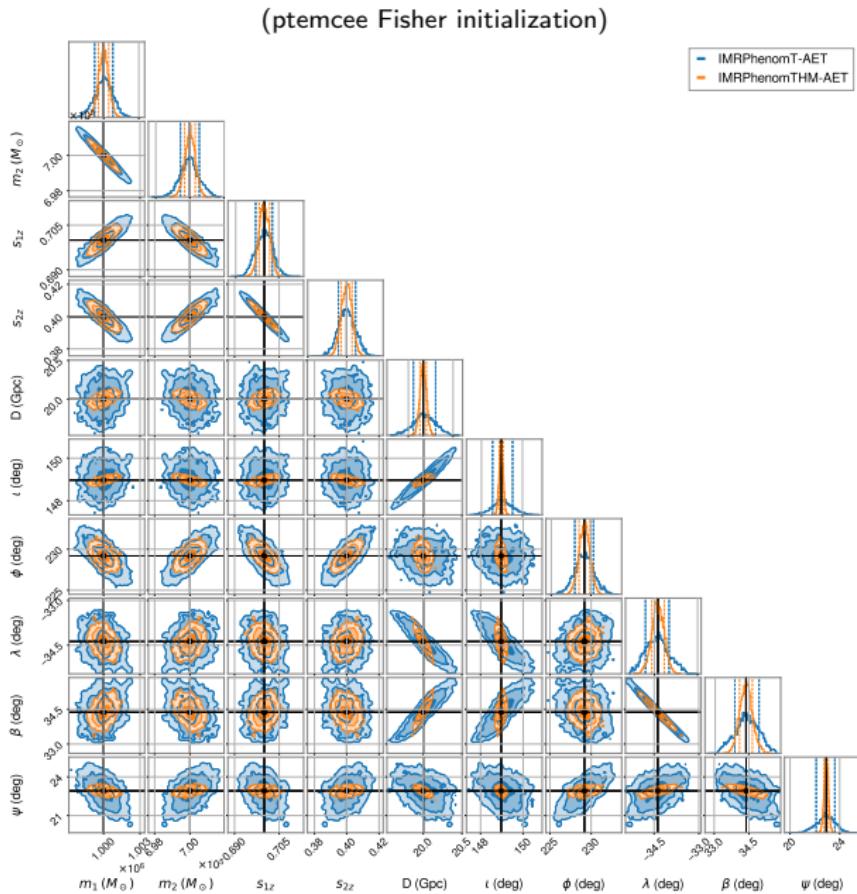
- ▶ IMRPhenomT  $h_{+, \times}(t)$  (GPU phenomxpy)
- ▶ Time domain LISA arm response & TDI at delayed times (GPU LISA Data Challenge tools)
- ▶ Fourier transform and likelihood calculation for each TDI channel (AET) (GPU phenomxpy)
- ▶ Sampling with Bilby (multinest, dynesty, ptemcee)

## Injection of a Massive Black Hole Binary

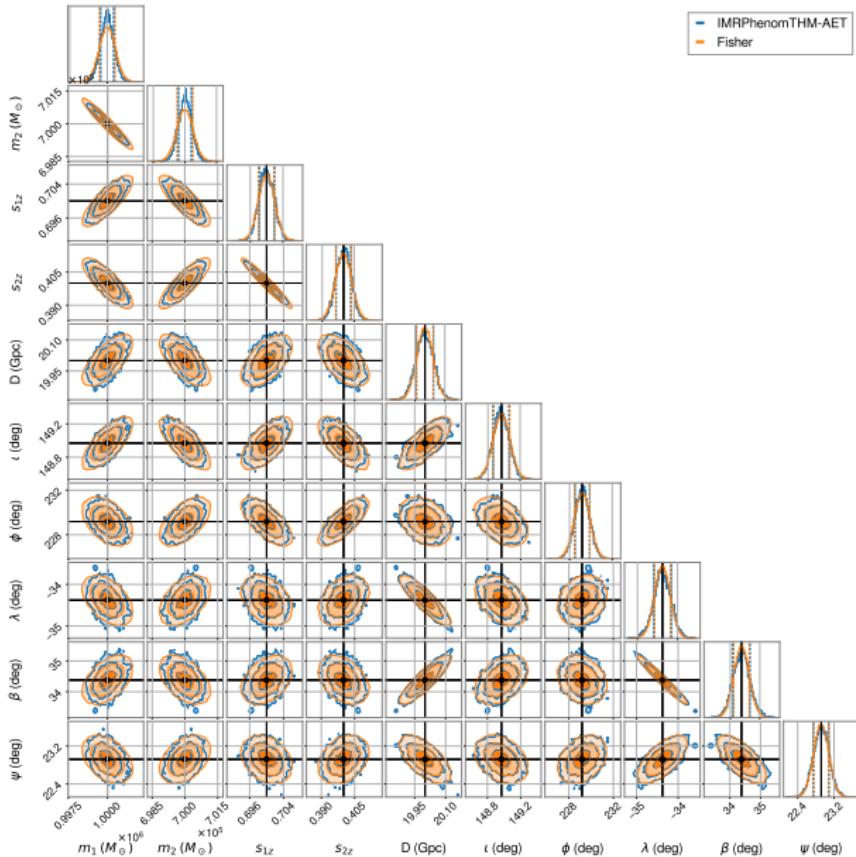
- ▶  $M_{\text{tot}} = 1.7 \cdot 10^6 M_{\odot}$
- ▶ Mass ratio = 1.4
- ▶ Spins  $a_{1,2} = 0.7, 0.4$
- ▶ Distance = 20 Gpc
- ▶  $f_{\text{low}} = 0.1 \text{ mHz}$
- ▶ Sampling rate = 0.2 Hz
- ▶ Duration = 61 days
- ▶ SNR  $\sim 3800$
- ▶ Uniform, wide priors
- ▶ Zero noise



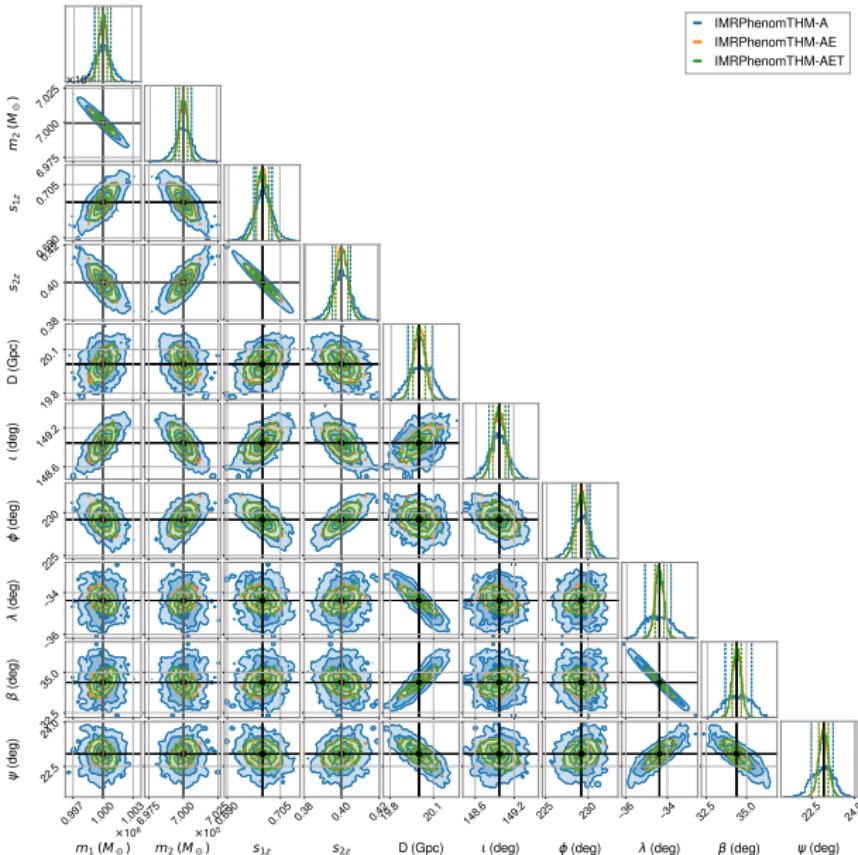
# Aligned spins: Effect of subdominant modes



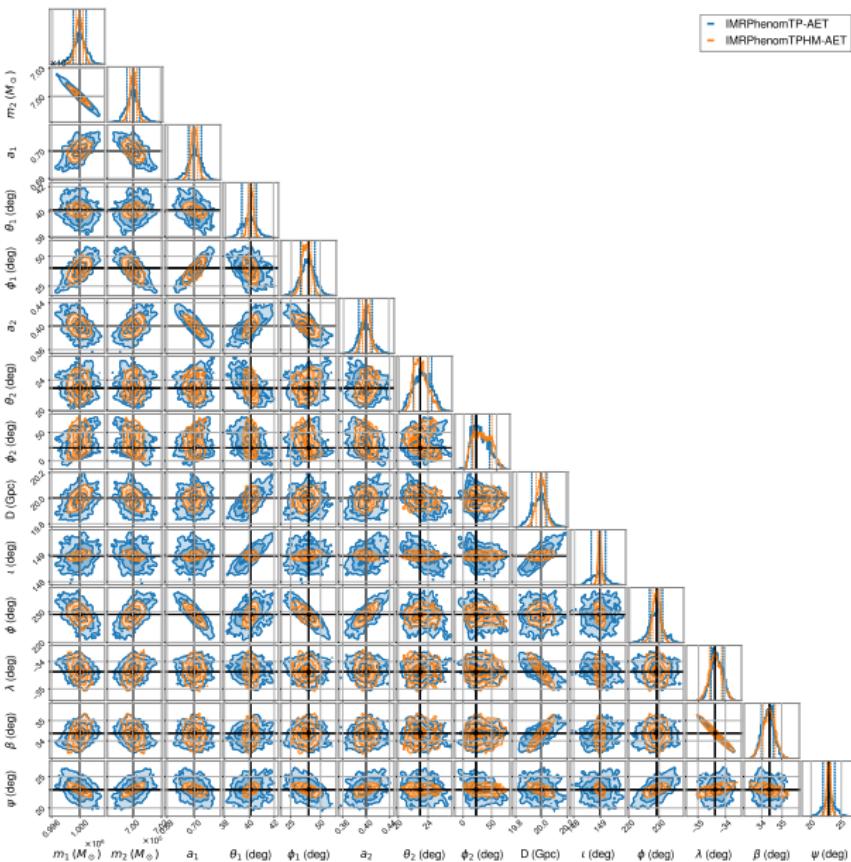
# Aligned spins: Comparison with Fisher analysis



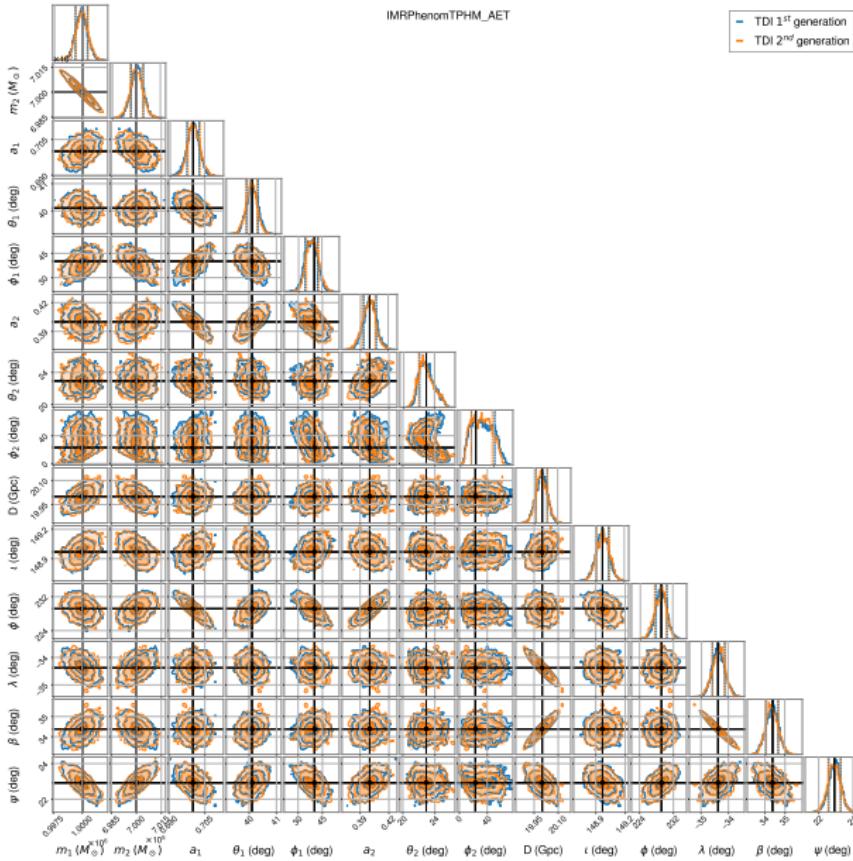
# Aligned spins: Comparison between TDI channels



# Precessing spins: Effect of subdominant modes



# Precessing spins: TDI generations 1<sup>st</sup> vs 2<sup>nd</sup>

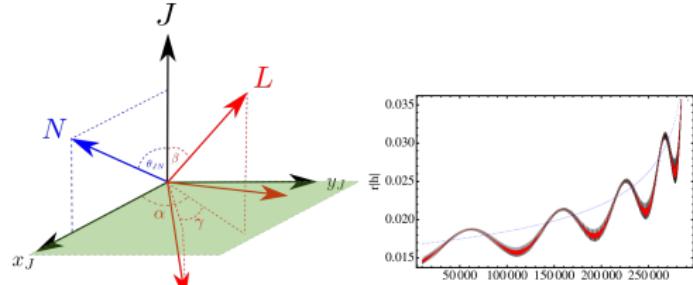


# Conclusions

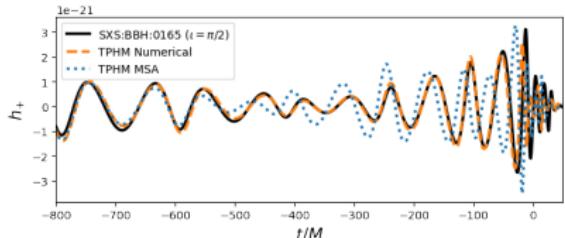
- ▶ Speed-up in waveform generation and LISA response → parameter estimation tractable
- ▶ First Bayesian analysis of MBBHs for LISA with a time-domain response including the effects of subdominant harmonics and precession
- ▶ Mass components recovered at 0.1% accuracy. Spins components at 1%
- ▶ Sampling all the parameters requires a smarter initialization of the sampler or more restrictive priors
- ▶ Smarter algorithms combining non-uniform time grids and interpolation (Multibanding, Relative binning, FANTA, etc.) will reduce more the computational cost

- ▶ Analytical expressions for amplitude and phase as a function of time
- ▶ Include the effects of subdominant harmonics 22, 21, 33, 44, 55
- ▶ Precession
  - ▶ Post-Newtonian Euler angles (NNLO, single-spin)
  - ▶ Multi-Scale Analysis (MSA, double-spin)
  - ▶ Numerical evolution of spins PN equations (SpinTaylor)
  - ▶ No mode asymmetries

- ▶ Reimplemented in python (phenomxpy)
  - ▶ numpy vectorization
  - ▶ numba just-in-time compilation
  - ▶ cupy GPU support
  - ▶ Parallelization of time array, for one waveform
  - ▶ Efficient calculation of polarizations from rotating frames (M. Boyle+, 2014)
  - ▶ 20 mode (M. Roselló, 2024)
  - ▶ Extension to eccentricity in progress (M. de Lluc Planas, A. Ramos-Buades)



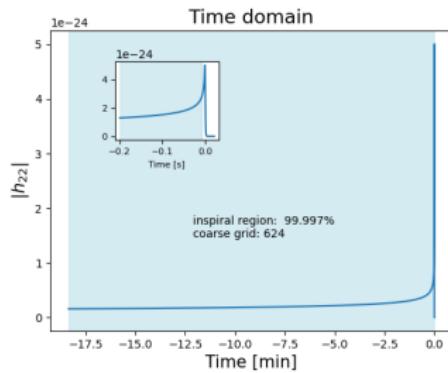
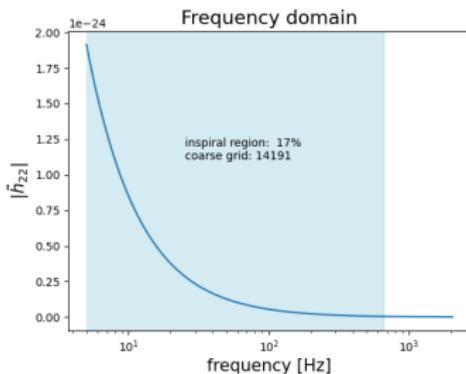
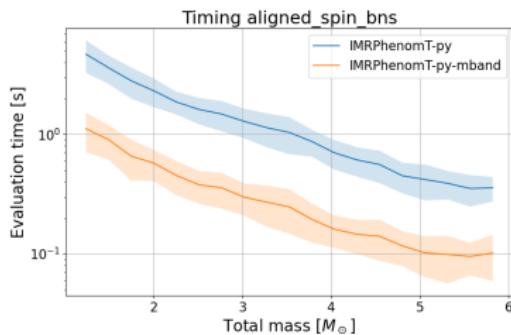
P. Schmidt+ 2012



H. Estellés+, 2021

# Multibanding for PhenomT (in progress)

- ▶ Evaluate model amplitude/phase in a coarse non-uniform grid
- ▶  $\Delta x = \sqrt{2R/\phi(x)''}$  (computed on-the-fly)
- ▶ Interpolate to final fine grid
- ▶ Build waveform
- ▶ Speed-up as long as interpolation is cheaper than model evaluation



# Efficient calculation of polarizations from rotating frames

$$CP \rightarrow J \rightarrow L_0$$

Traditional Phenom method

$$h_{\ell m}^J(t) = \sum_{m'=-\ell}^{\ell} \mathcal{D}_{mm'}^{\ell*}(\alpha(t), \beta(t), \gamma(t)) h_{\ell m'}^{CP}(t)$$

$$h_{\ell m}^{L0}(t) = \sum_{m'=-\ell}^{\ell} \mathcal{D}_{mm'}^{\ell*}(\alpha_0, \beta_0, \gamma_0) h_{\ell m'}^J(t)$$

$$h_+ - ih_\times = \sum_{l=2}^{\infty} \sum_{m=-l}^l h_{lm}^{L0}(t) {}_{-2}Y_{lm}(\iota, \phi)$$

Wigner-D calculations

$$\text{PhenomT: } 74 + 155 + 32 = 250$$

$$\text{Up to } l=5: 155 + 155 + 32 = 321$$

$$\text{Up to } l=8: 959 + 959 + 77 = 1939$$

Quaternion method (M. Boyle+, 2014)

$${}_{-2}Y_{lm}(\iota, \phi) = (-1)^s \sqrt{\frac{2l+1}{4\pi}} \mathcal{D}_{m,-s}^l(R_{\iota,\phi})$$

$$h_+ - ih_\times = e^{2i\hat{\delta}(t)} \sum_{l,m} h_{lm}^{CP}(t) {}_{-2}Y_{lm}(\hat{\iota}(t), \hat{\phi}(t))$$

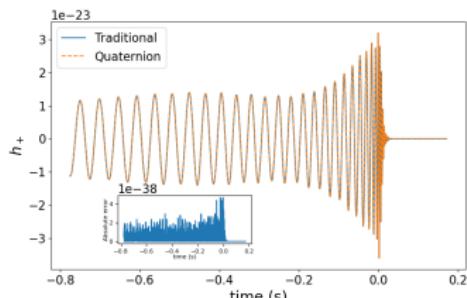
$$h_+ - ih_\times = \sum_{l,m} h_{lm}^{CP}(t) \sqrt{\frac{2l+1}{4\pi}} \mathcal{D}_{m,2}^l(R_{CP-J} R_{J-L_0} R_{\iota,\phi})$$

Wigner-D calculations

$$\text{PhenomT: } 10$$

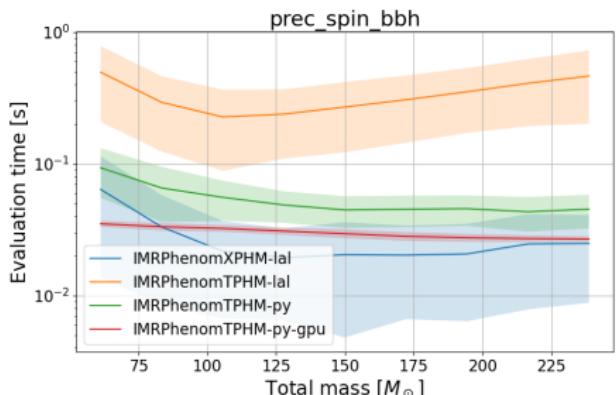
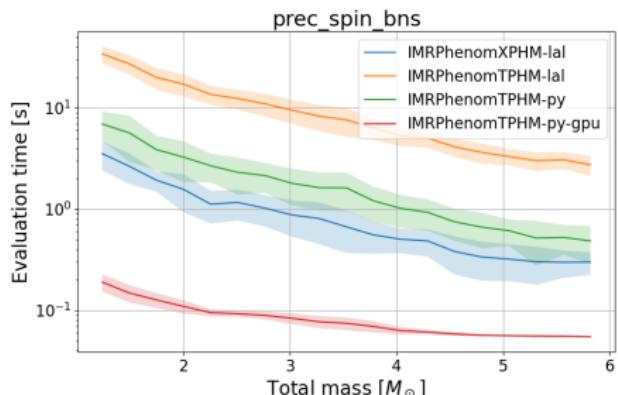
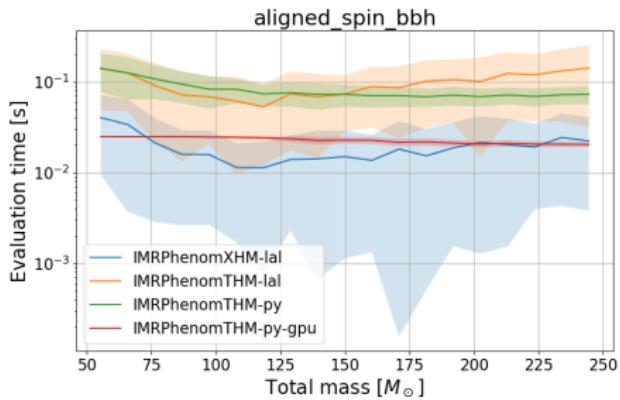
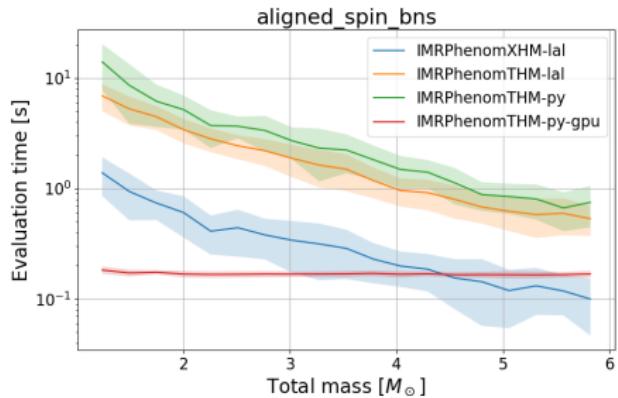
$$\text{Up to } l=5: 32$$

$$\text{Up to } l=8: 77$$



# Timing waveform generation (LIGO-like sources)

Catalogs of 1000 BBH, BNS,  $f_{low} \sim 20$  Hz



# Waveform families

## Effective-One-Body (time-domain)

- ▶ Solve Hamiltonian dynamics (ODEs)
- ▶ Very accurate inspiral (better than PN)
- ▶ Free Hamiltonian coefficients calibrated to Numerical Relativity (NR)
- ▶ Slow to evaluate → Reduced Order Models (ROMs)

## Surrogate (time-domain)

- ▶ Interpolate NR waveforms
- ▶ Most faithful to NR
- ▶ Needs high density of NR waveforms
- ▶ Reduced parameter space
- ▶ Reduced waveform length

## Other approximants

- ▶ Taylors (PN analytical)
- ▶ SpinTaylor (PN numerical evolution)
- ▶ NRTidal (matter effects)

## Phenom (Fourier or time domain)

- ▶ Closed-form expressions for amplitude and phase as functions of frequency and intrinsic parameters
- ▶ "Phenomenological" ansatzes calibrated to NR
- ▶ Fast to evaluate
- ▶ Large parameter space coverage
- ▶ Arbitrary low frequency
- ▶ Easily parallelizable
- ▶ Easier to implement non-uniform frequency grids speed-up algorithms (multibanding, relative binning, etc.)
- ▶ IMRPhenomX (Fourier), IMRPhenomT (time)

# Only masses

