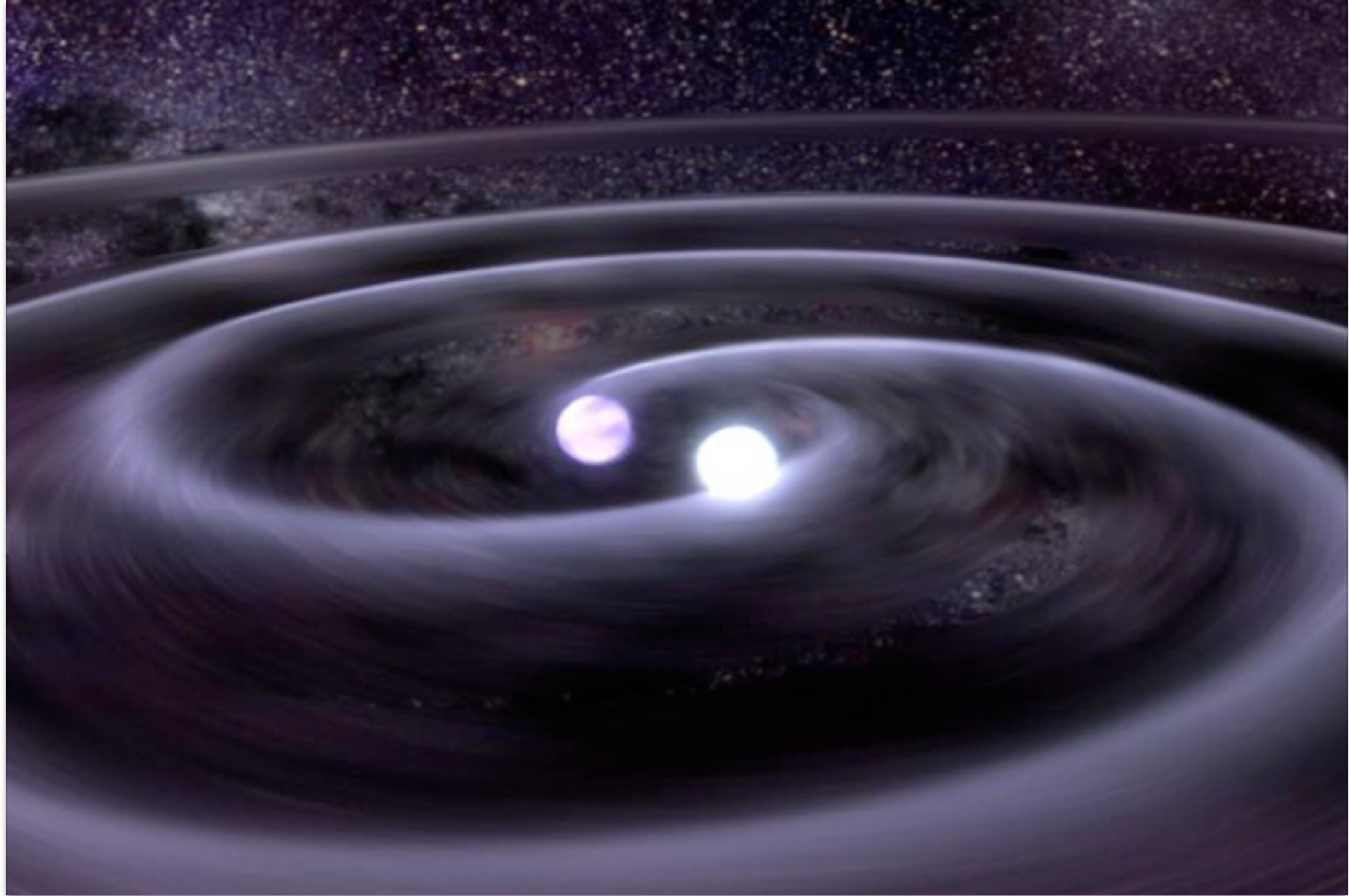


# Measuring the neutron-star equation of state with gravitational-wave observations



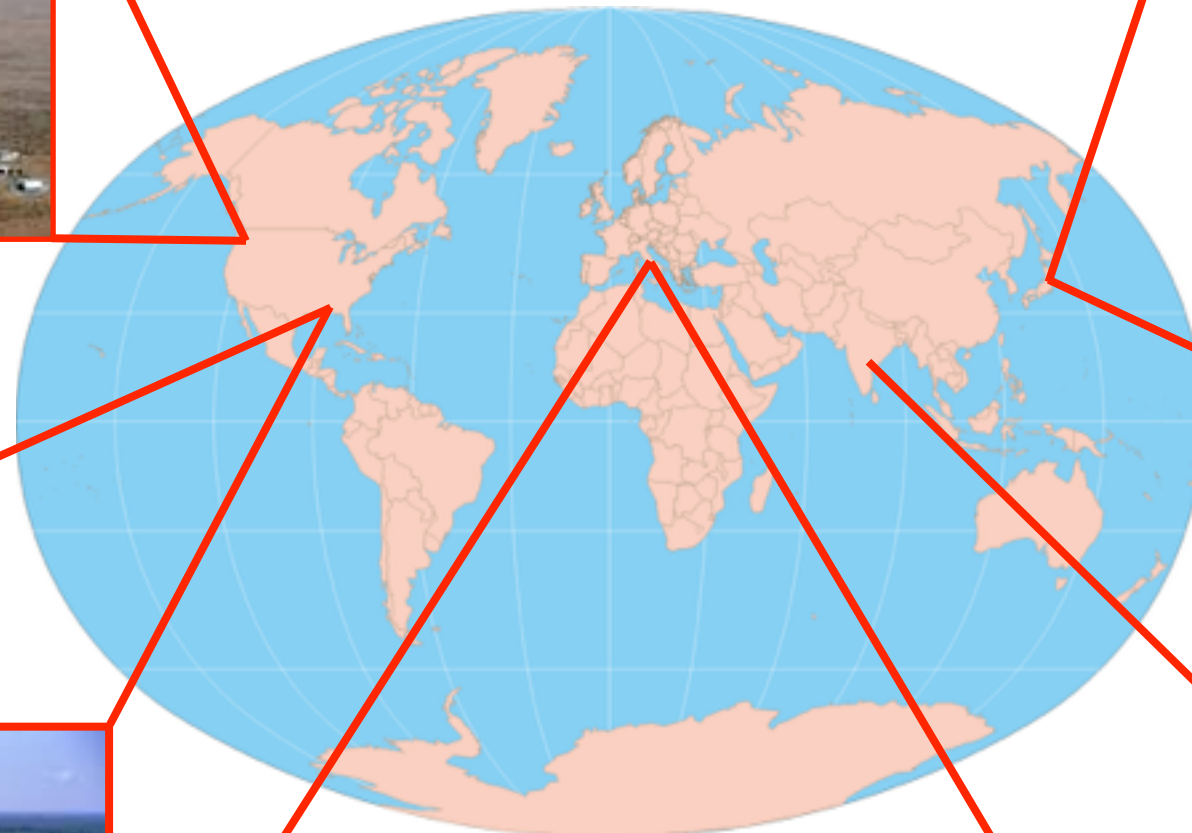
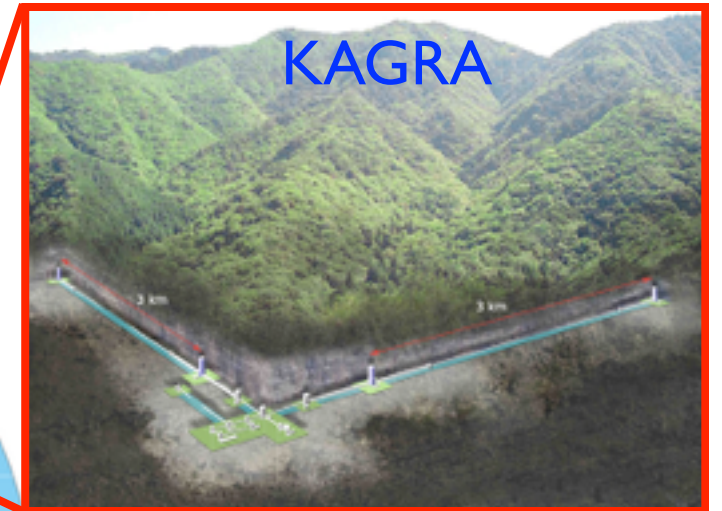
Ben Lackey

Syracuse University

INT, Seattle, Washington, 30 June 2016

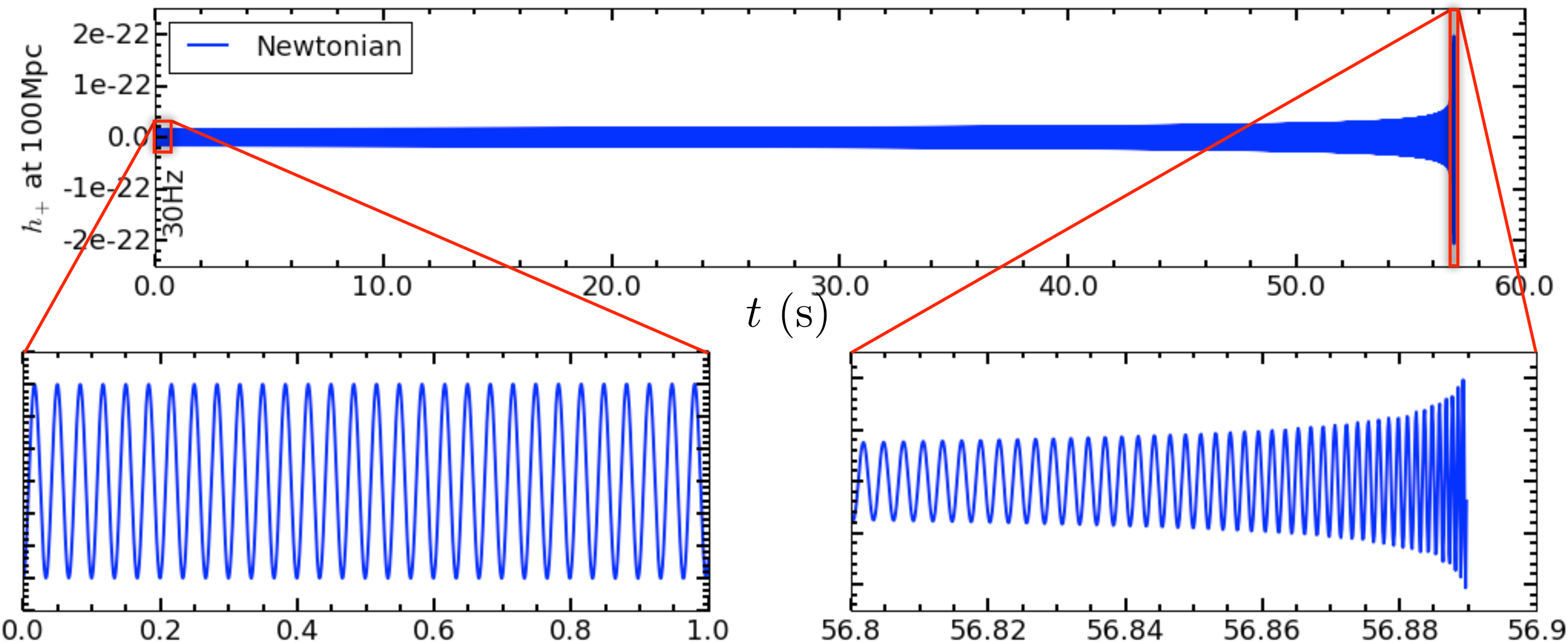
# Second generation gravitational-wave detectors

- Will reach design sensitivity in the next few years
- Sensitive to gravitational-waves between  $\sim 10\text{Hz}$  and a few kHz



LIGO-India

# Post-Newtonian waveform without matter effects



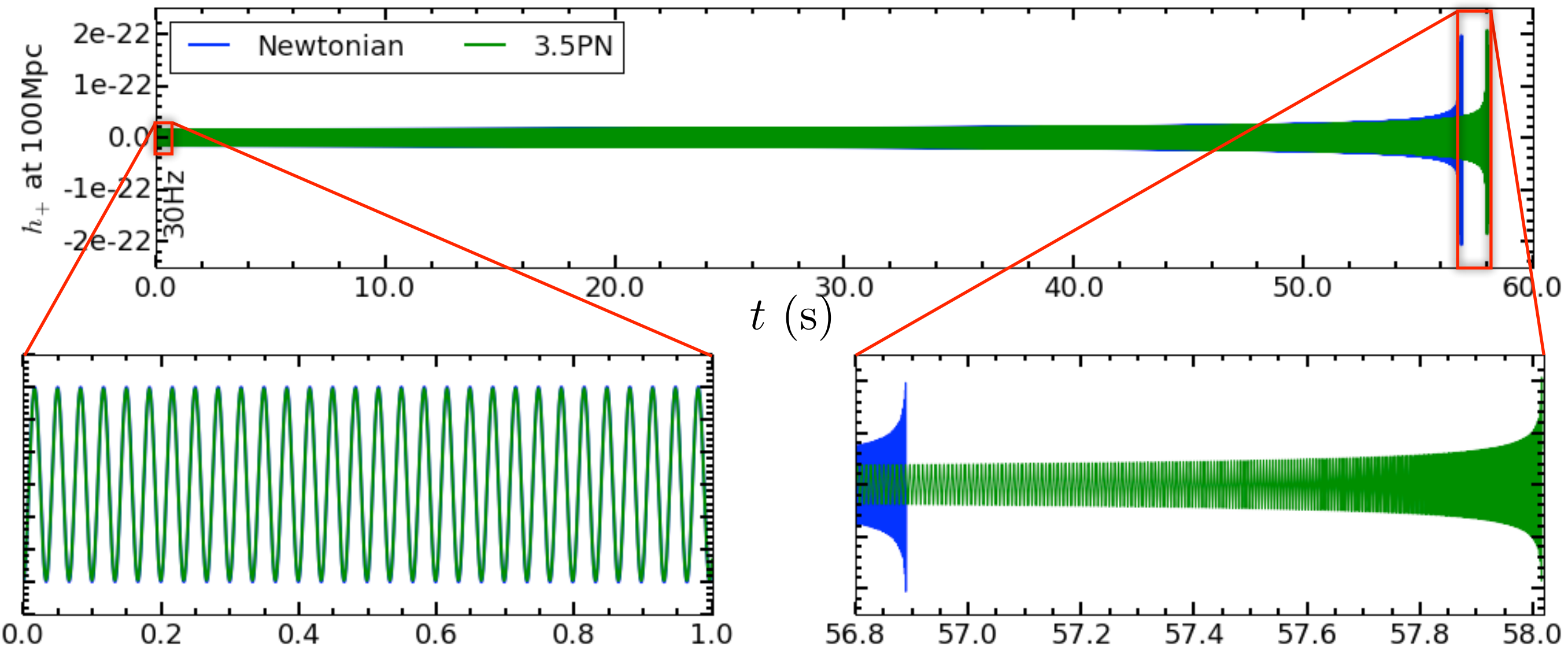
$$\text{Phase}(t) = 0\text{PN}(t; \mathcal{M})$$

Chirp mass:

$$\mathcal{M} = \frac{(m_1 m_2)^{3/5}}{(m_1 + m_2)^{1/5}}$$



# Post-Newtonian waveform without matter effects



$$(v/c)^2$$

$$(v/c)^7$$

$$\text{Phase}(t) = 0\text{PN}(t; \mathcal{M}) [1 + 1\text{PN}(t; \eta) + \dots + 3.5\text{PN}(t; \eta)]$$

Chirp mass:

$$\mathcal{M} = \frac{(m_1 m_2)^{3/5}}{(m_1 + m_2)^{1/5}}$$

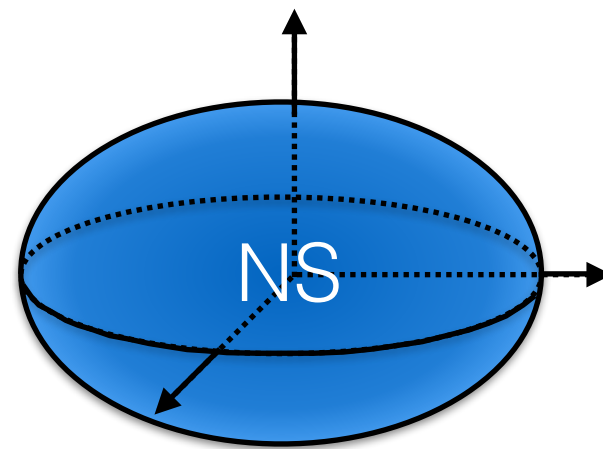
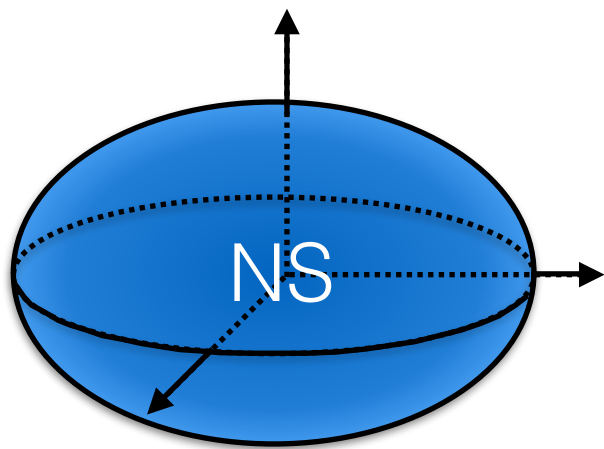
Symmetric mass ratio:

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

# Matter effects

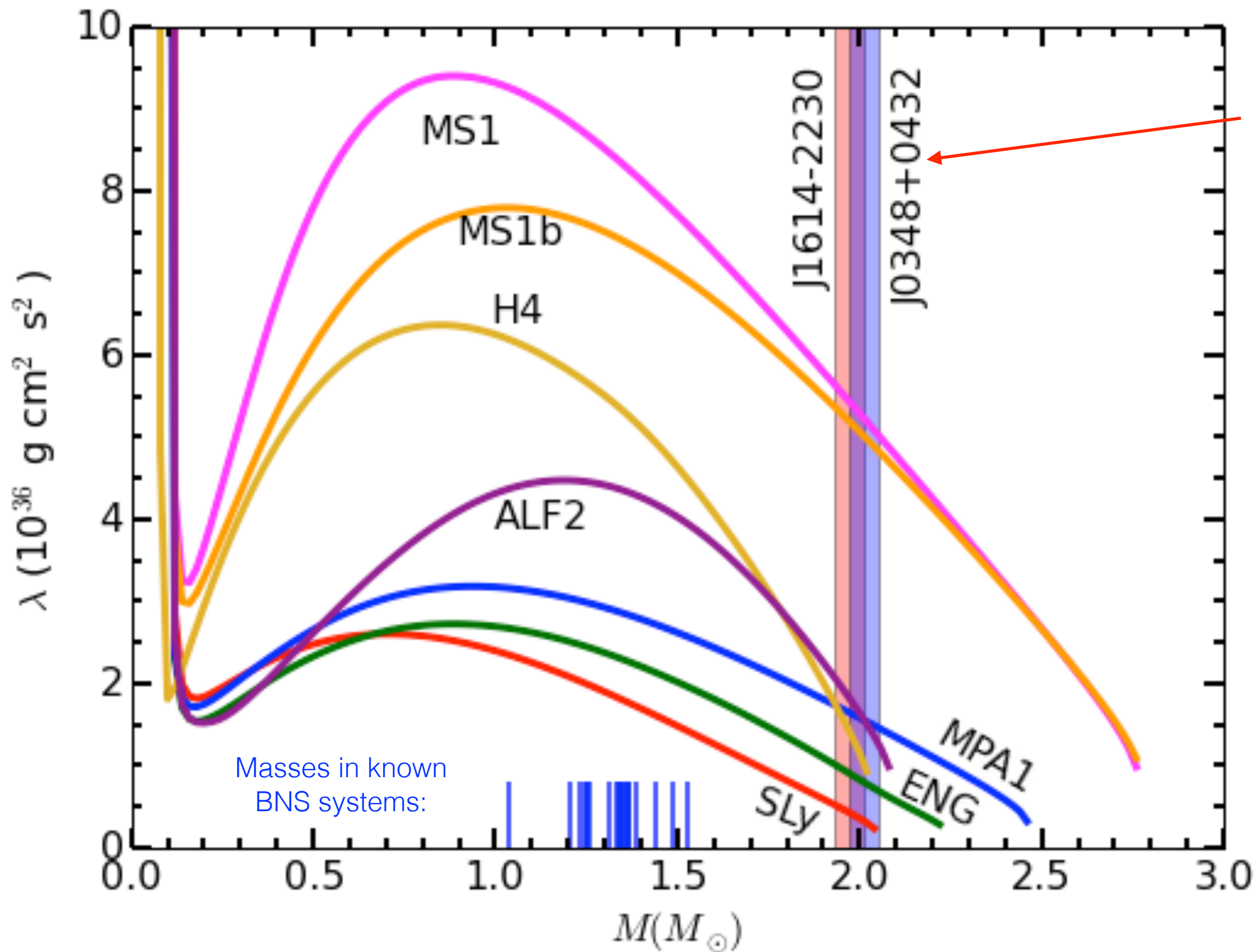
- Tidal field created by companion induces a quadrupole moment in the NS
  - Tidal field:  $\mathcal{E}_{ij}$
  - Quadrupole moment:  $Q_{ij}$
- Amount of deformation depends on stiffness of EOS via the tidal deformability  $\lambda$ :

$$Q_{ij} = -\lambda(\text{EOS}, M)\mathcal{E}_{ij}$$





# Matter effects

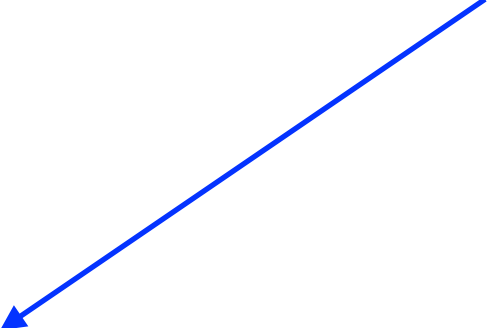


Highest known  
NS masses

# Matter effects

- Tidal effects first appear at same order as 5PN point-particle terms
- Leading term  $\tilde{\Lambda}$  is a linear combination of the tidal deformabilities of each object

$$\text{Phase}(t) = 0\text{PN}(t; \mathcal{M}) \left[ 1 + \overset{(v/c)^2}{1\text{PN}(t; \eta)} + \cdots + \overset{(v/c)^7}{3.5\text{PN}(t; \eta)} + \overset{(v/c)^{10}}{5\text{PN}(t; \tilde{\Lambda})} \right]$$

$$\tilde{\Lambda} = \frac{8}{13} \left[ (1 + 7\eta - 31\eta^2)(\Lambda_1 + \Lambda_2) + \sqrt{1 - 4\eta}(1 + 9\eta - 11\eta^2)(\Lambda_1 - \Lambda_2) \right]$$




# Matter effects

- Tidal effects first appear at same order as 5PN point-particle terms
- Leading term  $\tilde{\Lambda}$  is a linear combination of the tidal deformabilities of each object
- Effect of remainder term  $\delta\tilde{\Lambda}$  is  $\sim 10$ - $100$  times smaller

$$\text{Phase}(t) = 0\text{PN}(t; \mathcal{M}) \left[ 1 + 1\text{PN}(t; \eta) + \cdots + 3.5\text{PN}(t; \eta) + 5\text{PN}(t; \tilde{\Lambda}) + 6\text{PN}(t; \tilde{\Lambda}, \delta\tilde{\Lambda}) \right]$$

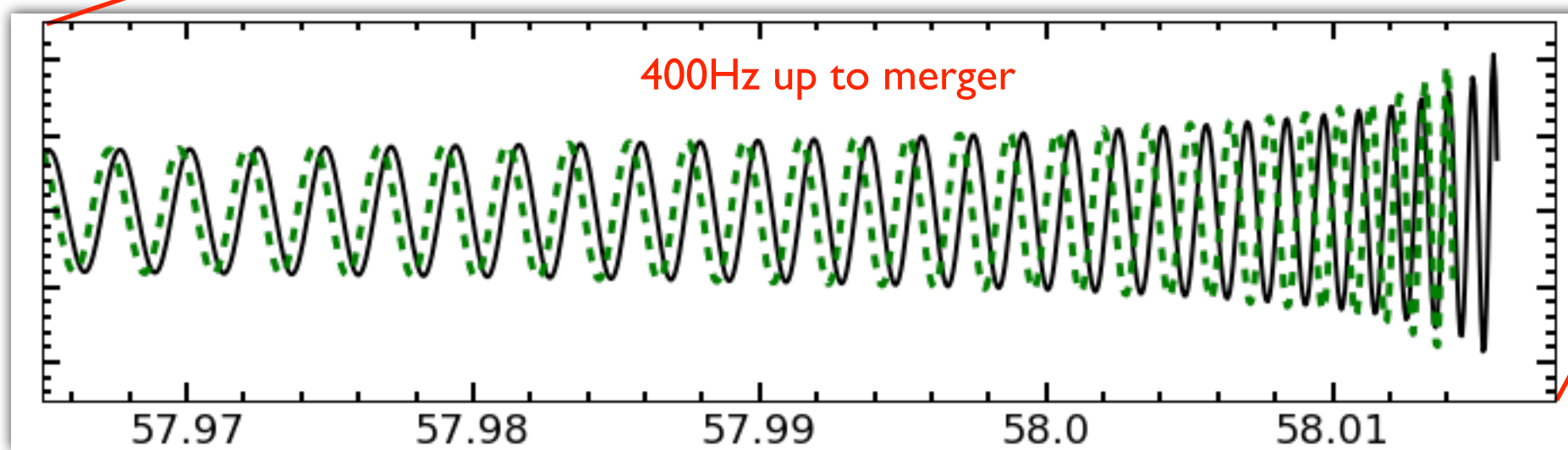
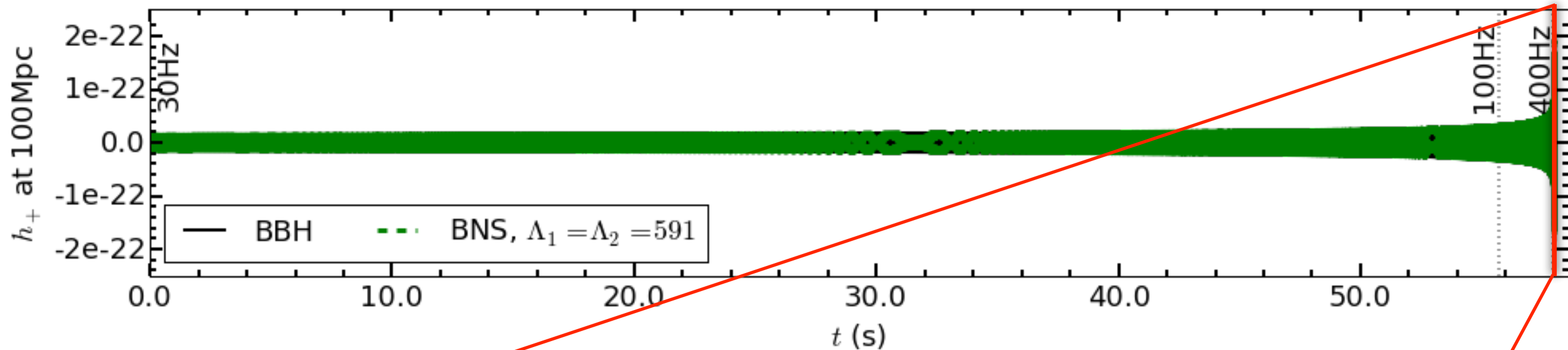
$(v/c)^2$ 
 $(v/c)^7$ 
 $(v/c)^{10}$ 
 $(v/c)^{12}$

$$\tilde{\Lambda} = \frac{8}{13} \left[ (1 + 7\eta - 31\eta^2)(\Lambda_1 + \Lambda_2) + \sqrt{1 - 4\eta}(1 + 9\eta - 11\eta^2)(\Lambda_1 - \Lambda_2) \right]$$

$$\delta\tilde{\Lambda} = \frac{1}{2} \left[ \sqrt{1 - 4\eta} \left( 1 - \frac{13272}{1319}\eta + \frac{8944}{1319}\eta^2 \right) (\Lambda_1 + \Lambda_2) + \left( 1 - \frac{15910}{1319}\eta + \frac{32850}{1319}\eta^2 + \frac{3380}{1319}\eta^3 \right) (\Lambda_1 - \Lambda_2) \right]$$

# Matter effects

- Both NSs contribute to tidal effect
- Leads to phase shift of 5–15 radians



# Parameter estimation

- Can estimate the parameters  $\vec{\theta}$  of each inspiral from the data  $\mathbf{d}$  with Bayes' theorem:

$$p(\vec{\theta}|\mathbf{d}) = \frac{\text{Prior Likelihood}}{\text{Evidence}} = \frac{p(\vec{\theta})p(\mathbf{d}|\vec{\theta})}{p(\mathbf{d})}$$

- Time series of stationary, Gaussian noise  $\mathbf{n}$  has the distribution

$$p_n[\mathbf{n}(t)] \propto e^{-(\mathbf{n},\mathbf{n})/2} \quad (a, b) = 4\text{Re} \int_0^\infty \frac{\tilde{a}(f)\tilde{b}(f)}{S_n(f)} df$$

- (data from detector  $\mathbf{d}$ ) = (noise  $\mathbf{n}$ ) + (model of GW signal  $m(\vec{\theta})$ )

$$p(\mathbf{d}|\vec{\theta}) \propto e^{-(\mathbf{d}-m, \mathbf{d}-m)/2}$$

# Parameter estimation

- Can estimate the parameters  $\vec{\theta}$  of each inspiral from the data  $\mathbf{d}$  with Bayes' theorem:

$$\text{Posterior } p(\vec{\theta}|\mathbf{d}) = \frac{\text{Prior } p(\vec{\theta}) \text{ Likelihood } p(\mathbf{d}|\vec{\theta})}{\text{Evidence } p(\mathbf{d})}$$

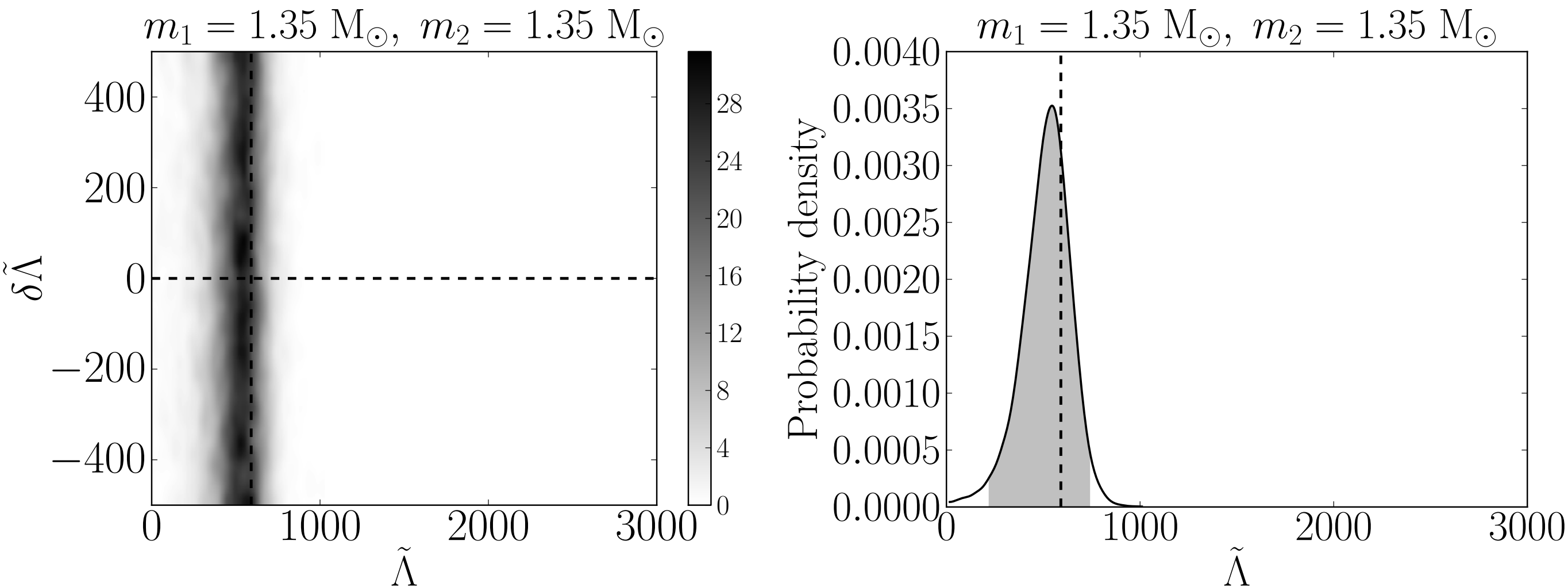
- Can sample the posterior with Markov chain Monte Carlo (MCMC), then marginalize over nuisance parameters

$$p(\mathcal{M}, \eta, \tilde{\Lambda}|d_n) = \int p(\vec{\theta}|d_n) d\vec{\theta}_{\text{nuisance}}$$

We only care about masses  
and tidal parameters here

# Parameter estimation

- Result of MCMC simulation for system with SNR=30
- Bayesian parameter estimation for aLIGO-aVirgo network:



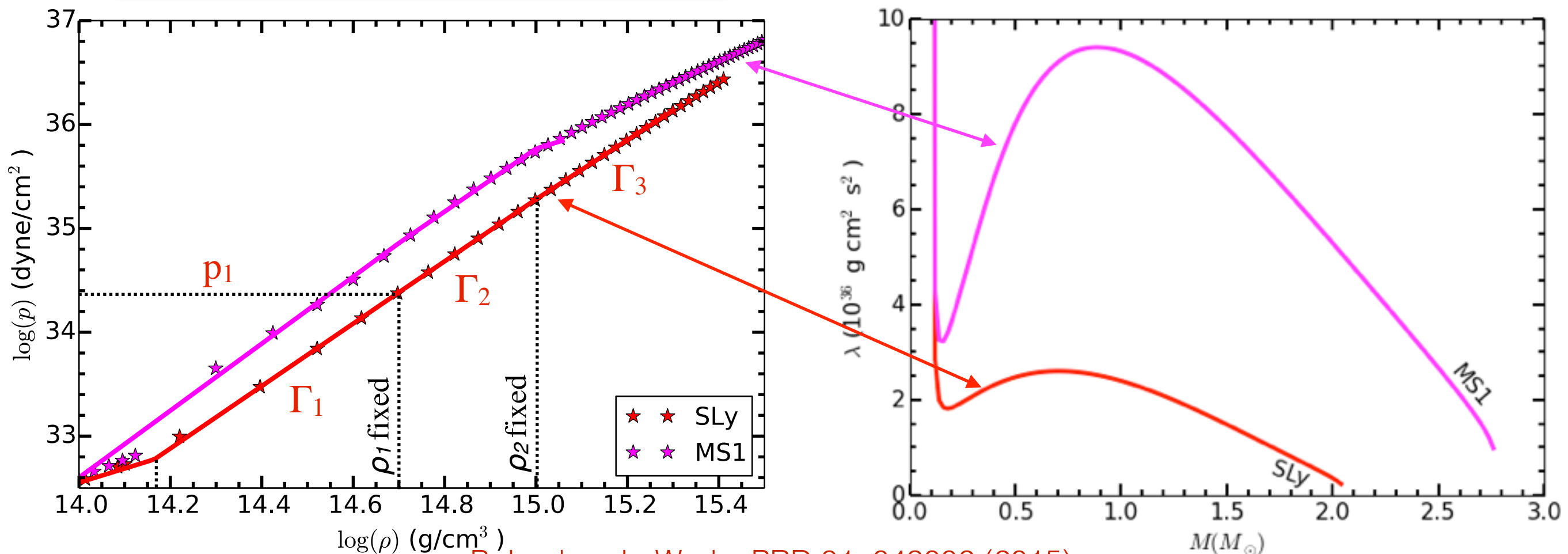


# Measuring the EOS directly

- The tidal deformability is calculated from the EOS
- This can be inverted to find EOS parameters from observations of the tidal parameters and masses

$$p(\rho) = \begin{cases} K_1 \rho^{\Gamma_1}, & \rho_0 < \rho < \rho_1 \\ K_2 \rho^{\Gamma_2}, & \rho_1 < \rho < \rho_2 \\ K_3 \rho^{\Gamma_3}, & \rho > \rho_2 \end{cases}$$

$$\begin{aligned} \lambda_1 &= \lambda_1[p(\rho), m_1] \\ \lambda_2 &= \lambda_2[p(\rho), m_2] \end{aligned}$$

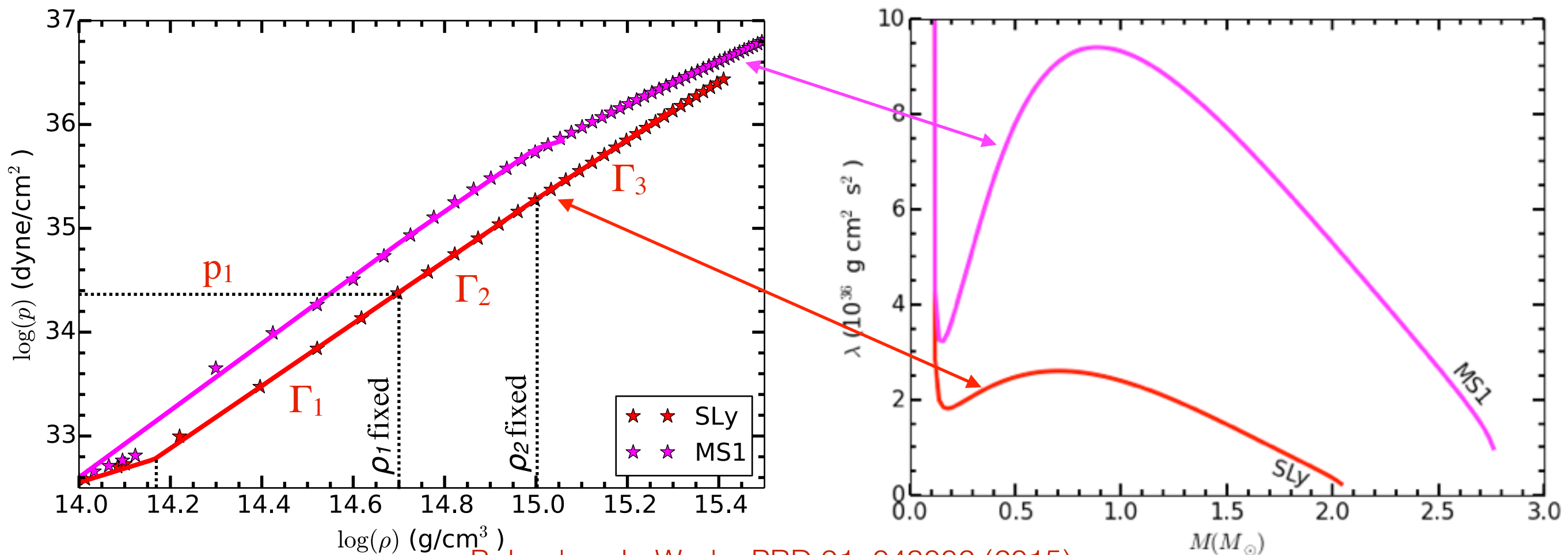


# Measuring the EOS directly

- Use Bayes' theorem again to estimate masses and EOS parameters:

$$\text{Posterior } p(\vec{x}|d_1 \dots d_N) = \frac{\text{Prior Likelihood } p(\vec{x})p(d_1 \dots d_N|\vec{x})}{\text{Evidence } p(d_1 \dots d_N)}$$

$$\vec{x} = \{\log(p_1), \Gamma_1, \Gamma_2, \Gamma_3, \mathcal{M}_1, \eta_1, \dots, \mathcal{M}_N, \eta_N\}$$



# Measuring the EOS directly

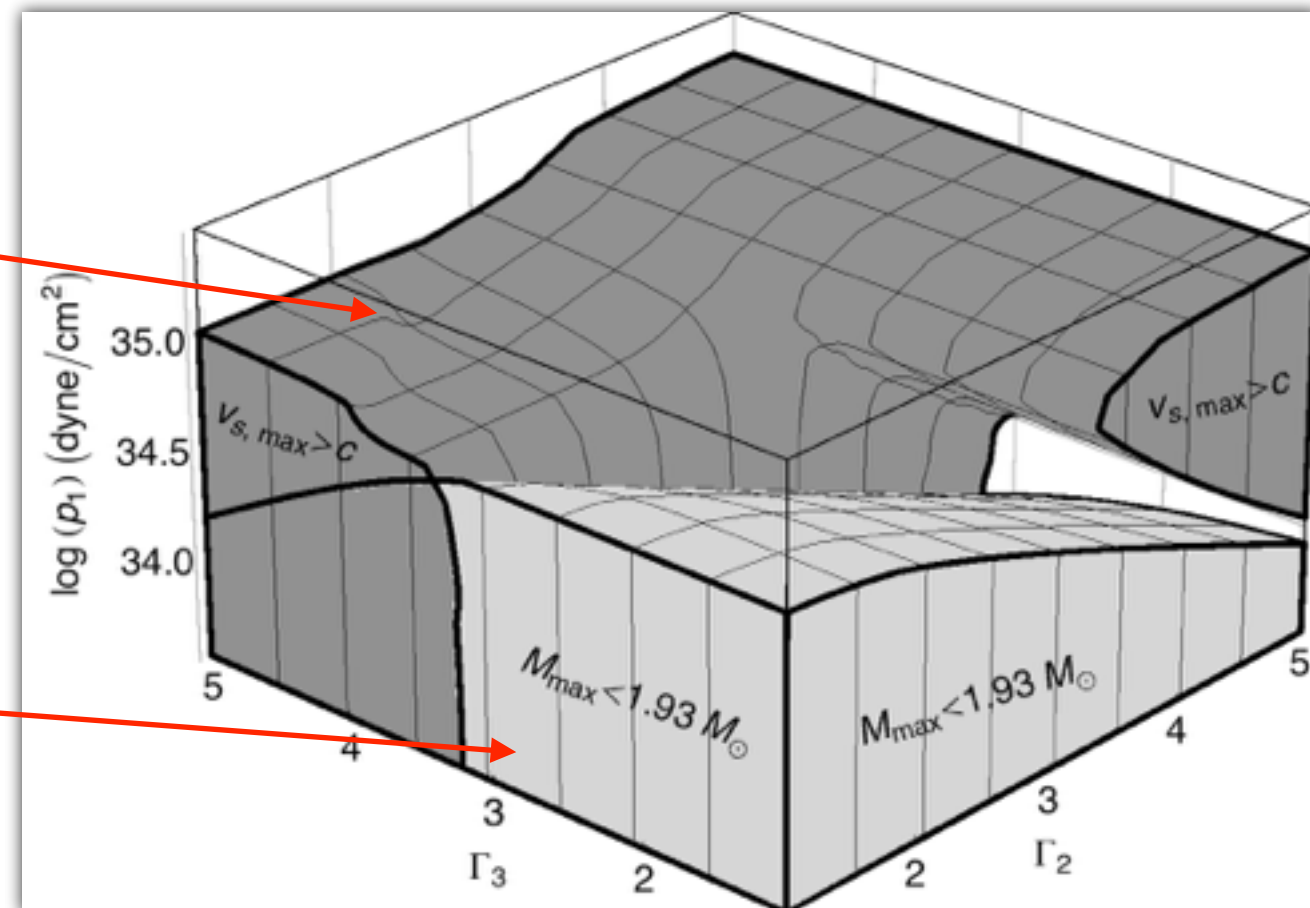
- Use Bayes' theorem again to estimate masses and EOS parameters:

$$\text{Posterior } p(\vec{x}|d_1 \dots d_N) = \frac{\text{Prior } p(\vec{x}) \text{ Likelihood } p(d_1 \dots d_N|\vec{x})}{\text{Evidence } p(d_1 \dots d_N)}$$

- Causality:** The speed of sound must be less than the speed of light


$$v_s = \sqrt{dp/d\epsilon} < c$$

- Maximum mass:** The EOS must allow for masses that are greater than observed NSs (at least  $1.93M_\odot$ )



# Measuring the EOS directly

- Use Bayes' theorem again to estimate masses and EOS parameters:

$$p(\vec{x}|d_1 \dots d_N) = \frac{\text{Prior} \quad \text{Likelihood}}{\text{Evidence}} \frac{p(\vec{x})p(d_1 \dots d_N|\vec{x})}{p(d_1 \dots d_N)}$$


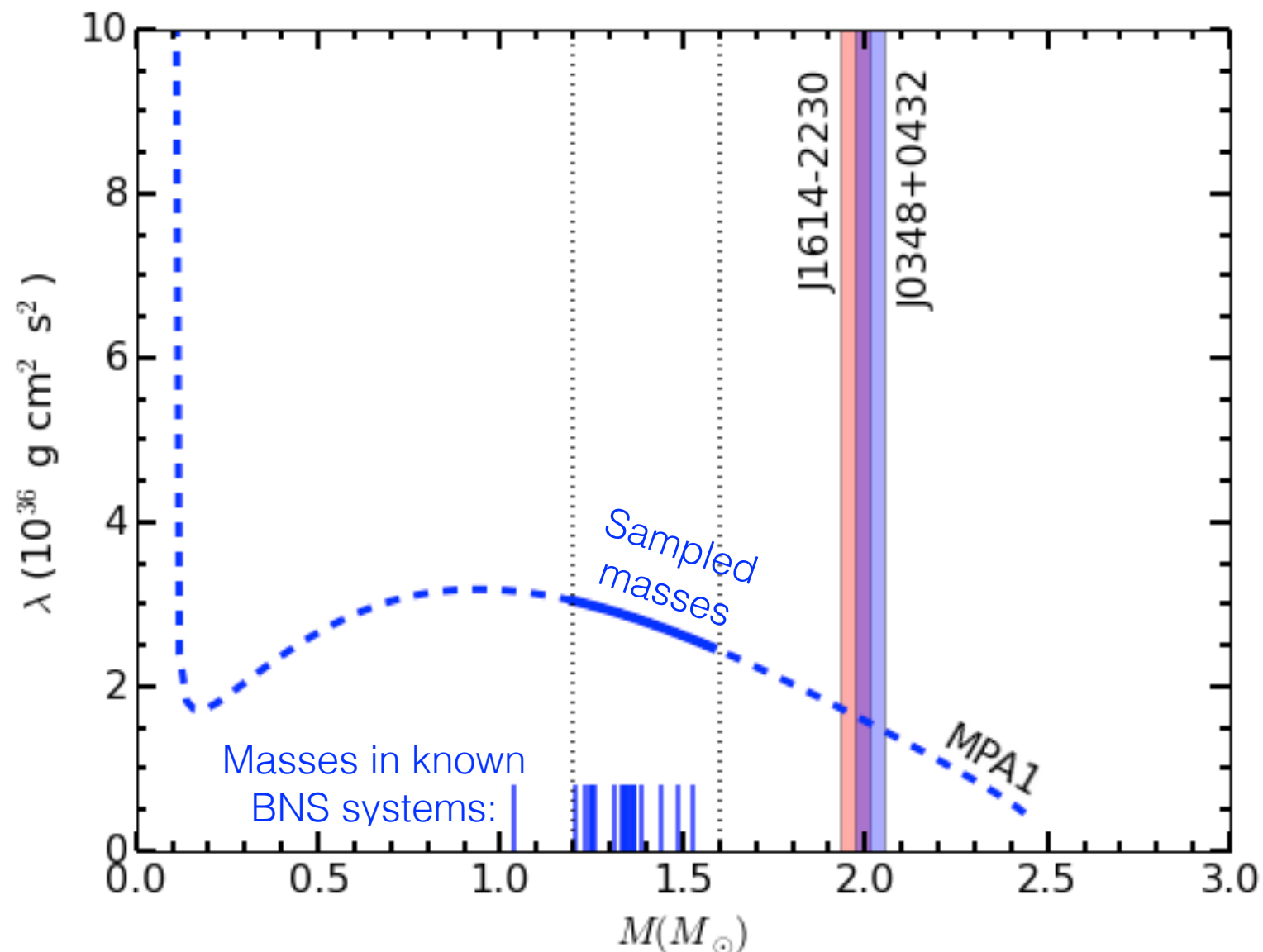
- Likelihood is the product of the marginalized distributions for each event

$$p(d_1, \dots, d_N|\vec{x}) = \prod_{n=1}^N \text{Marginalized posterior for single event} \quad p(\mathcal{M}_n, \eta_n, \tilde{\Lambda}_n|d_n)|_{\tilde{\Lambda}_n = \tilde{\Lambda}(\mathcal{M}_n, \eta_n, \text{EOS})}$$

- The 4 EOS parameters are found from an MCMC simulation of the 4+2N parameters by marginalizing over the 2N mass parameters

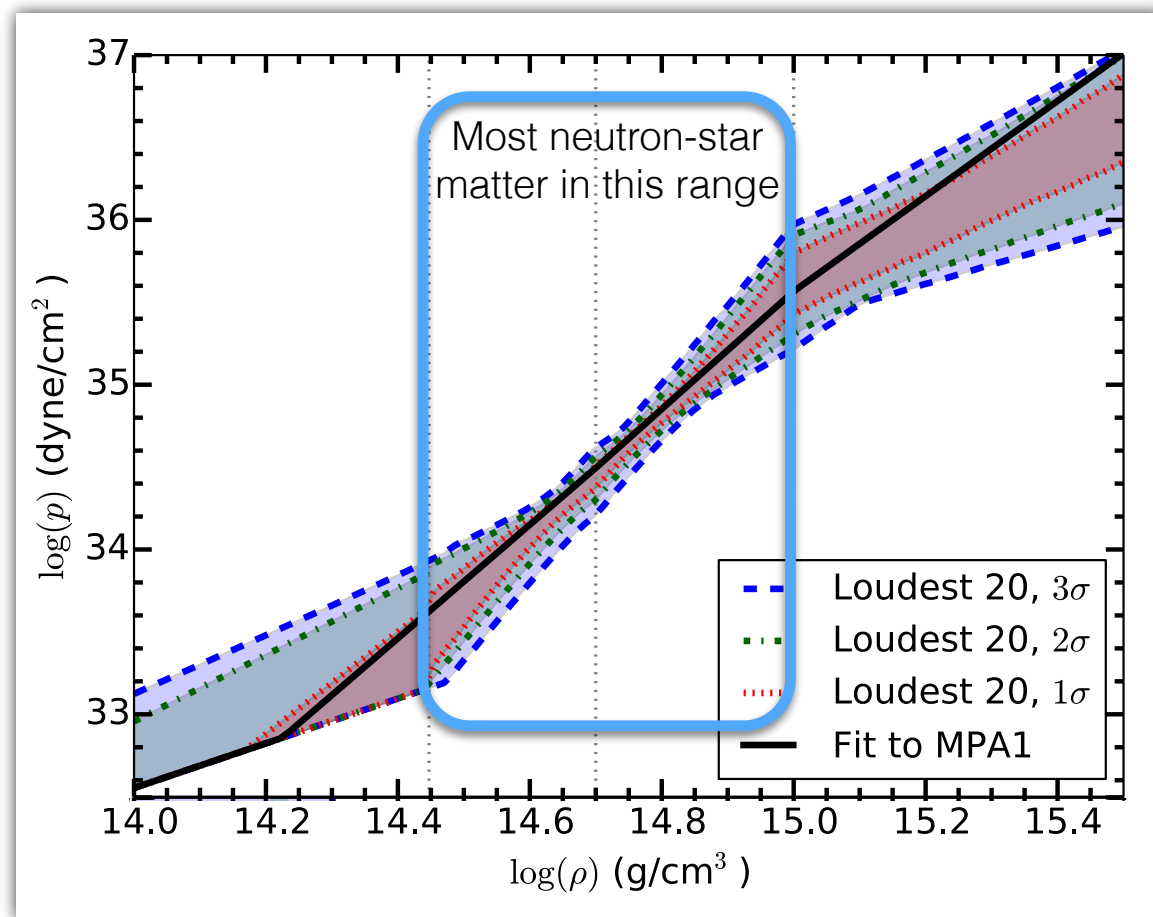
# Measuring the EOS directly

- We simulated a year of events for the aLIGO-aVirgo network
  - Each detector had  $\sim 40$  BNS events/year with  $\text{SNR} > 8$
  - Sampled the masses uniformly in the range  $1.2M_{\odot} - 1.6M_{\odot}$
  - Assumed MPA1 as the true EOS



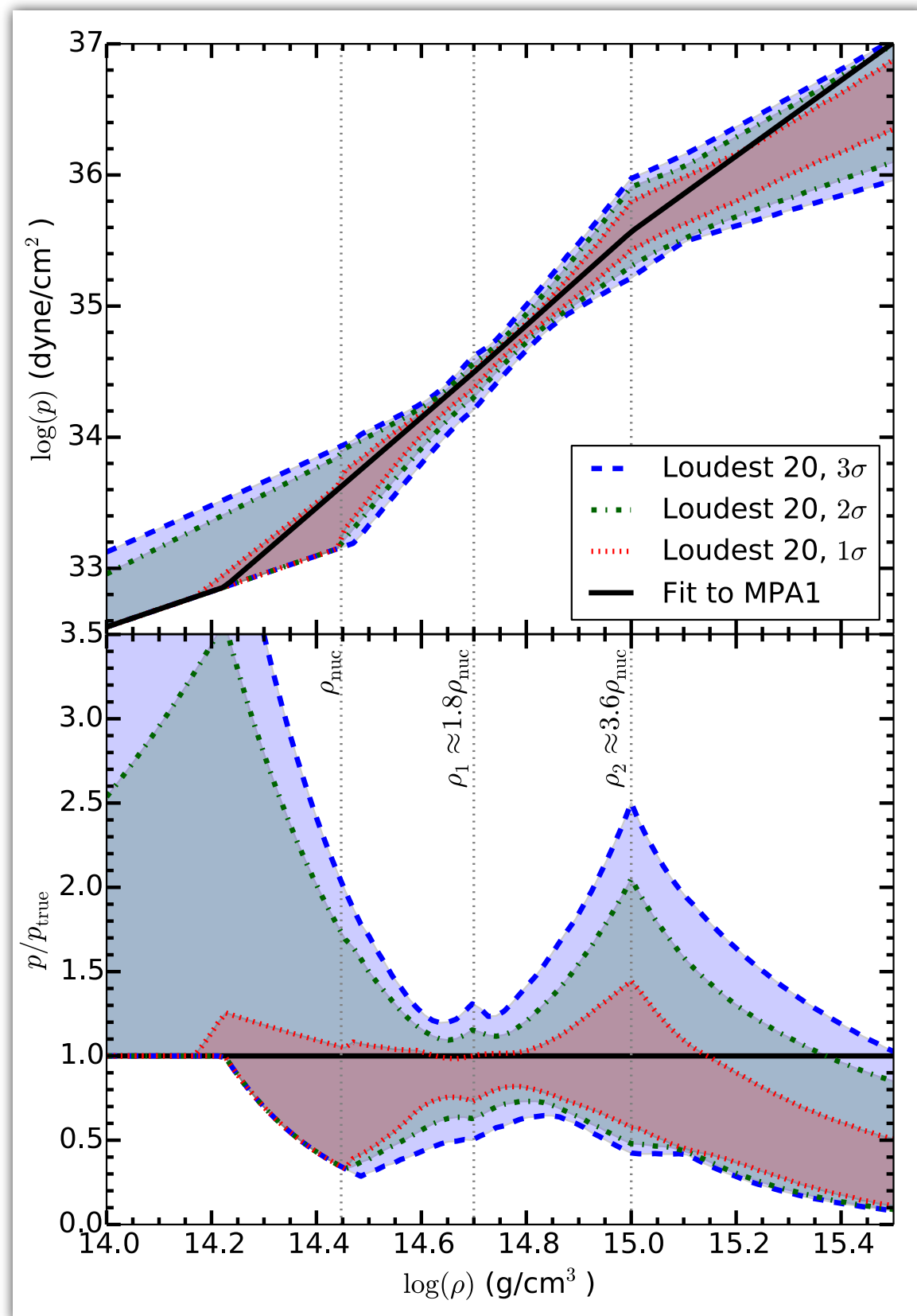


# Measuring the EOS directly

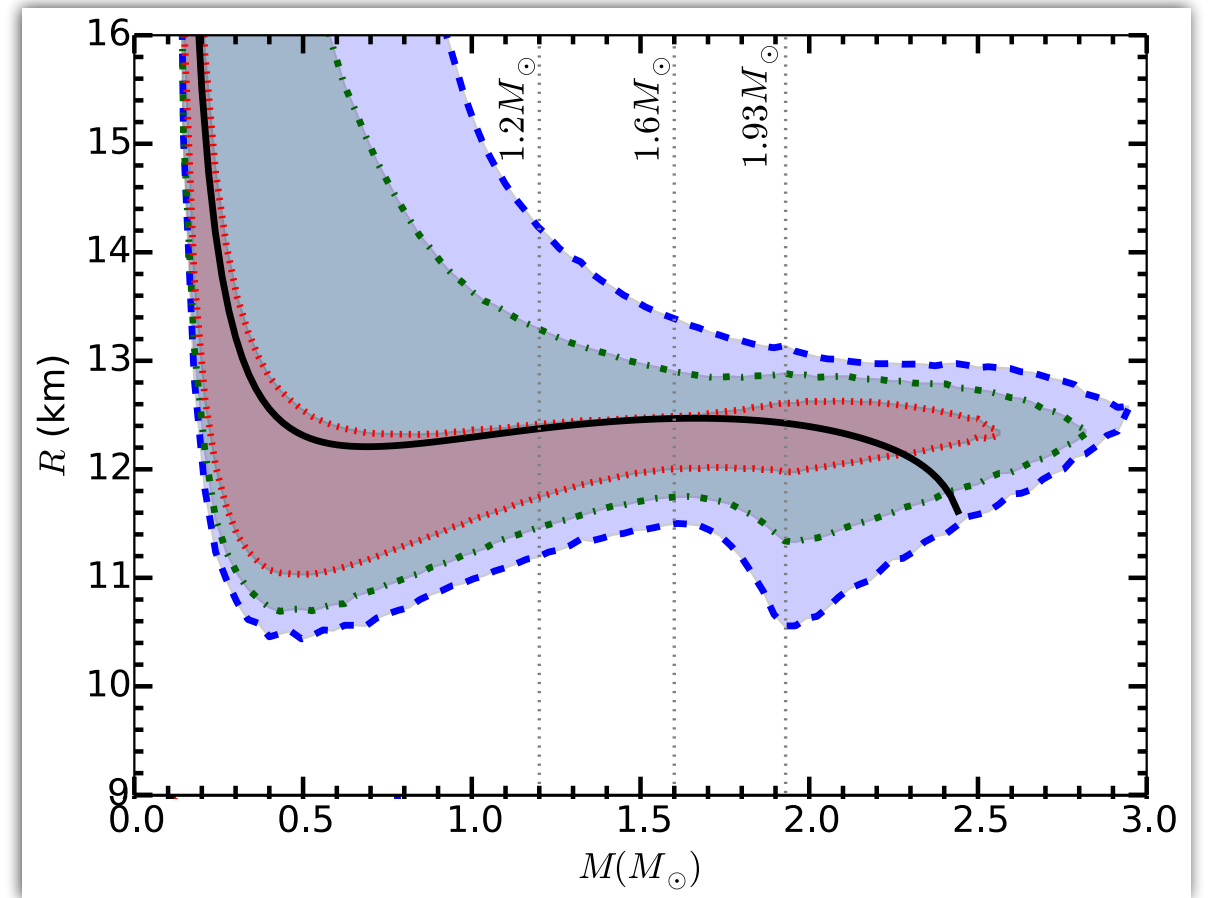
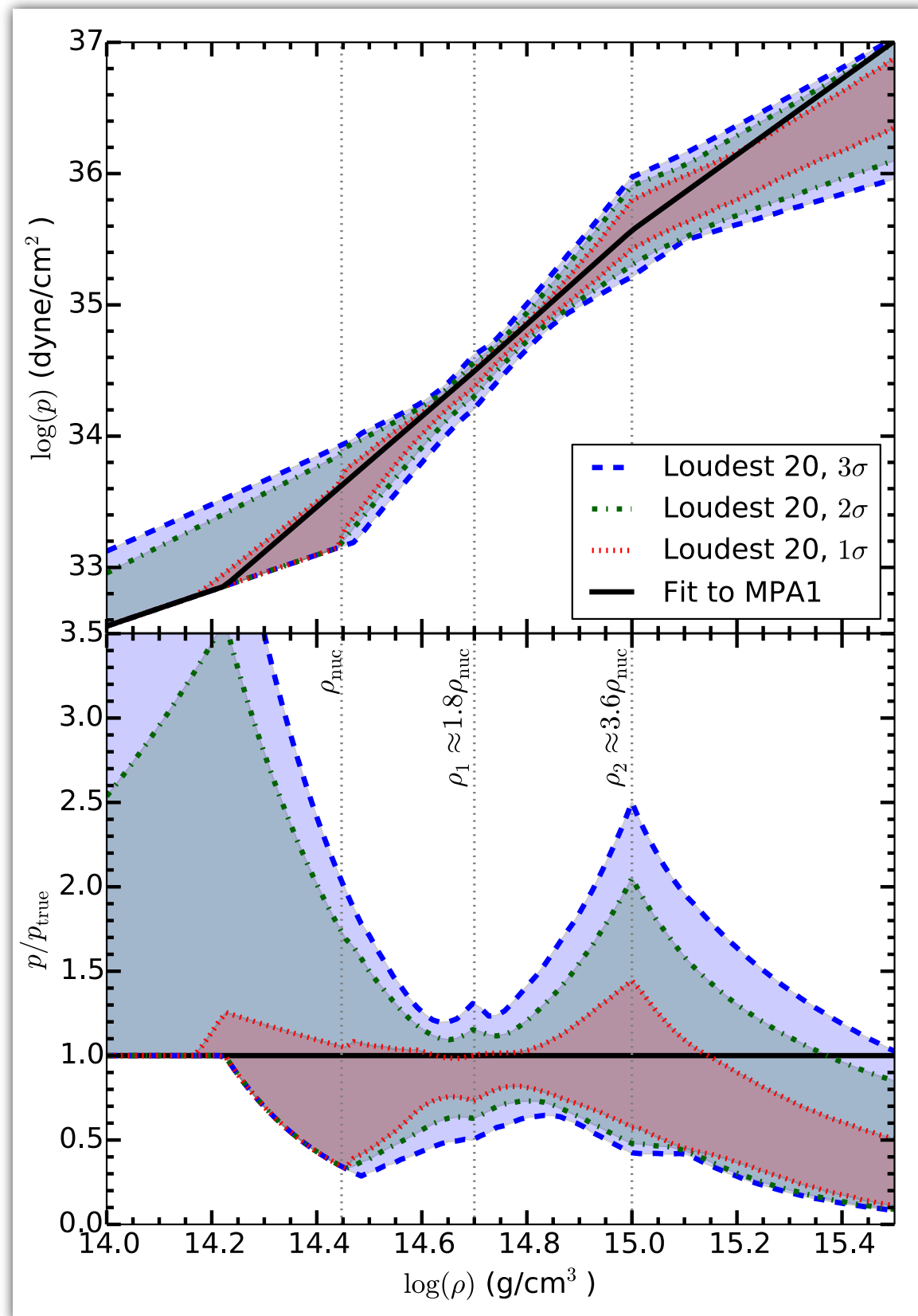


- 68% credible region
- 95% credible region
- 99.7% credible region

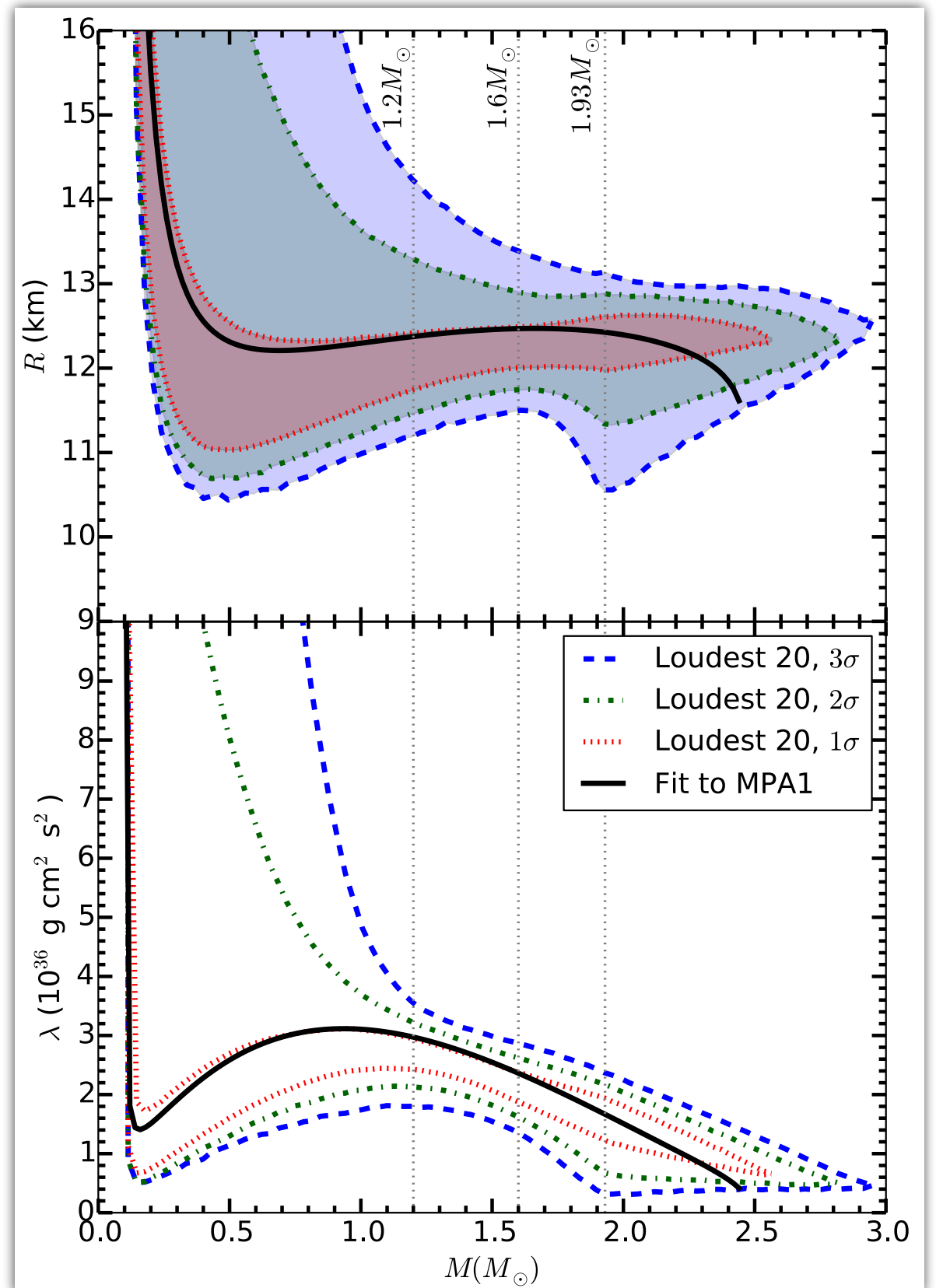
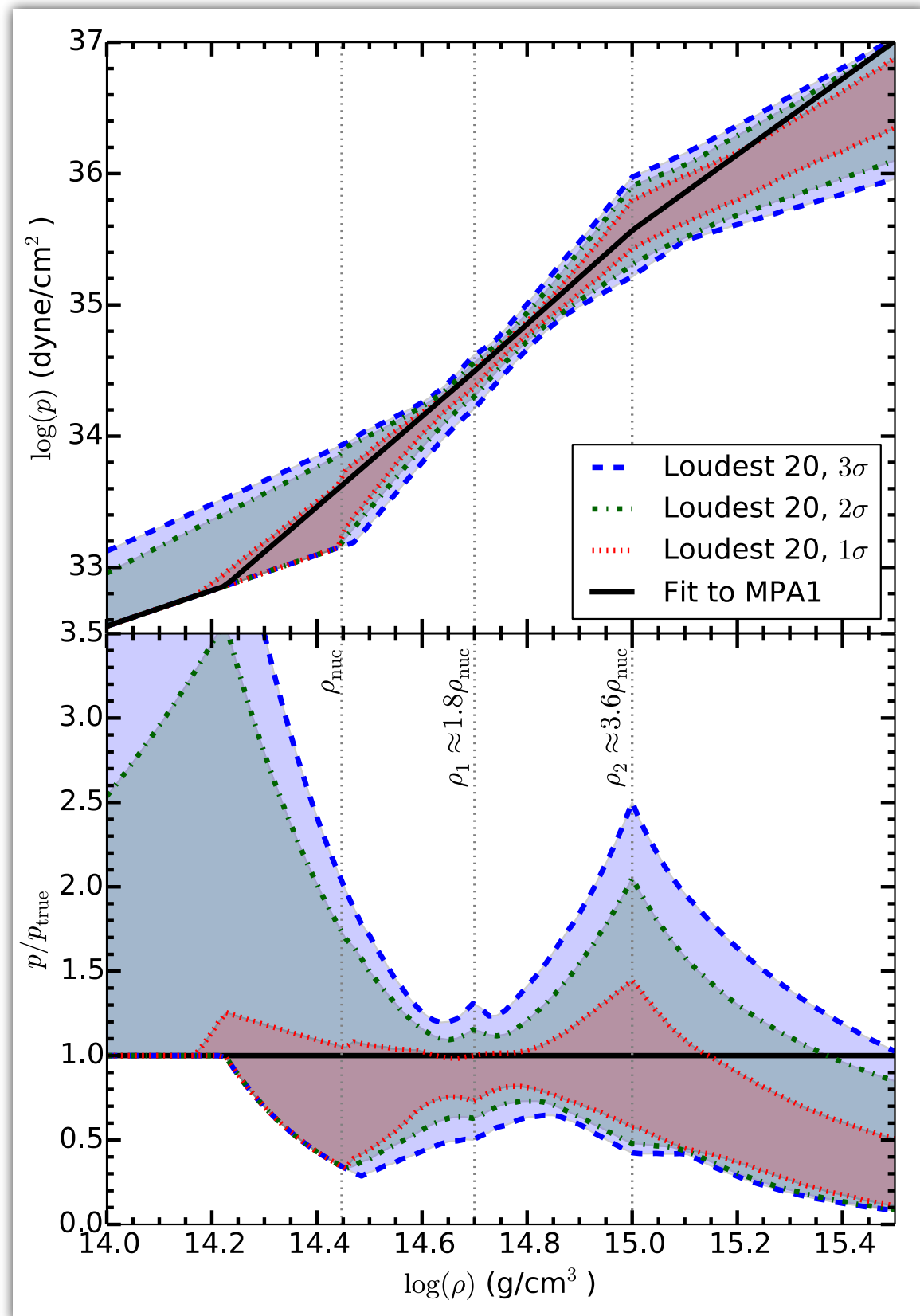
# Measuring the EOS directly



# Measuring the EOS directly

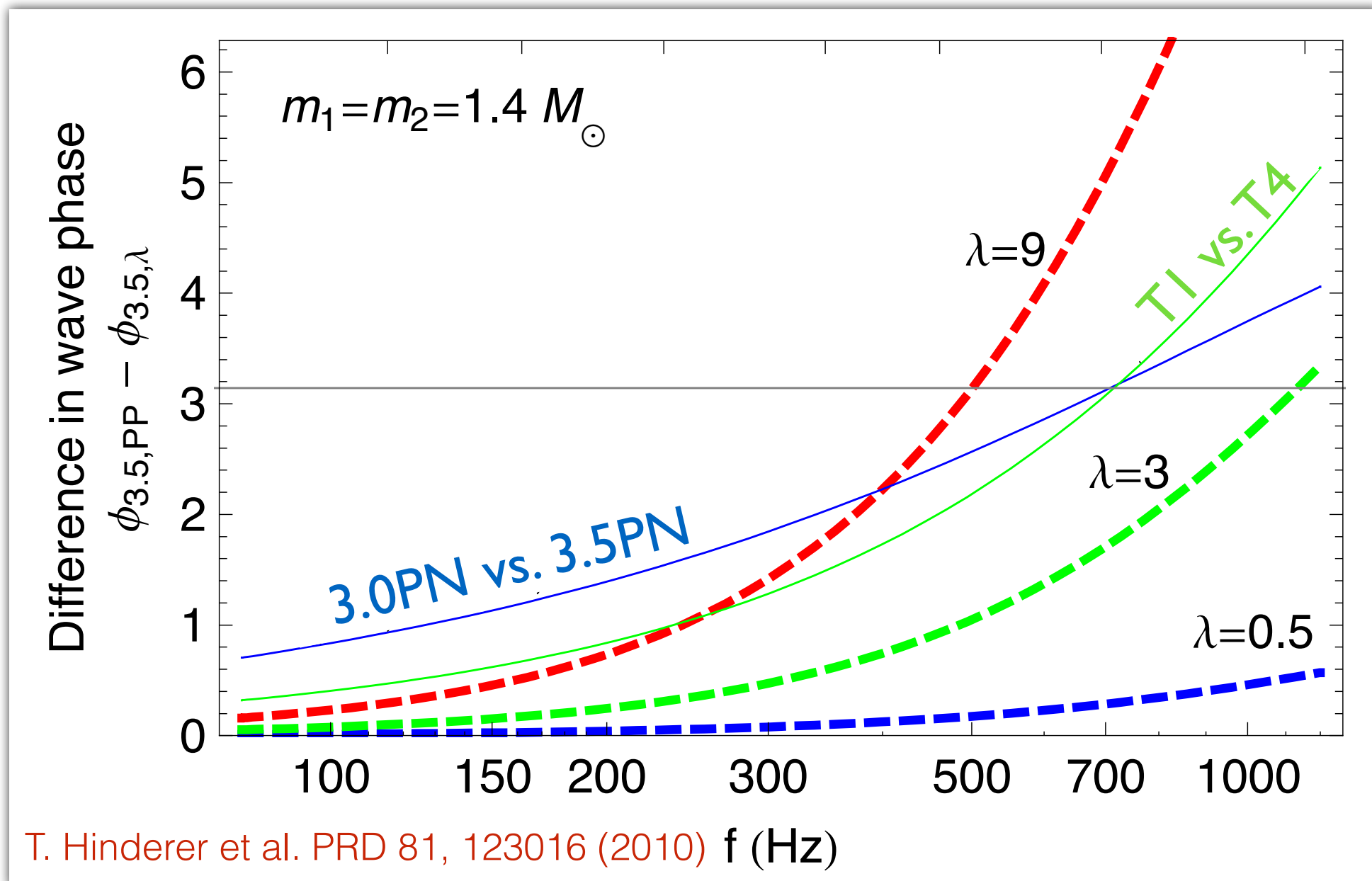


# Measuring the EOS directly



# Is the post-Newtonian waveform accurate enough?

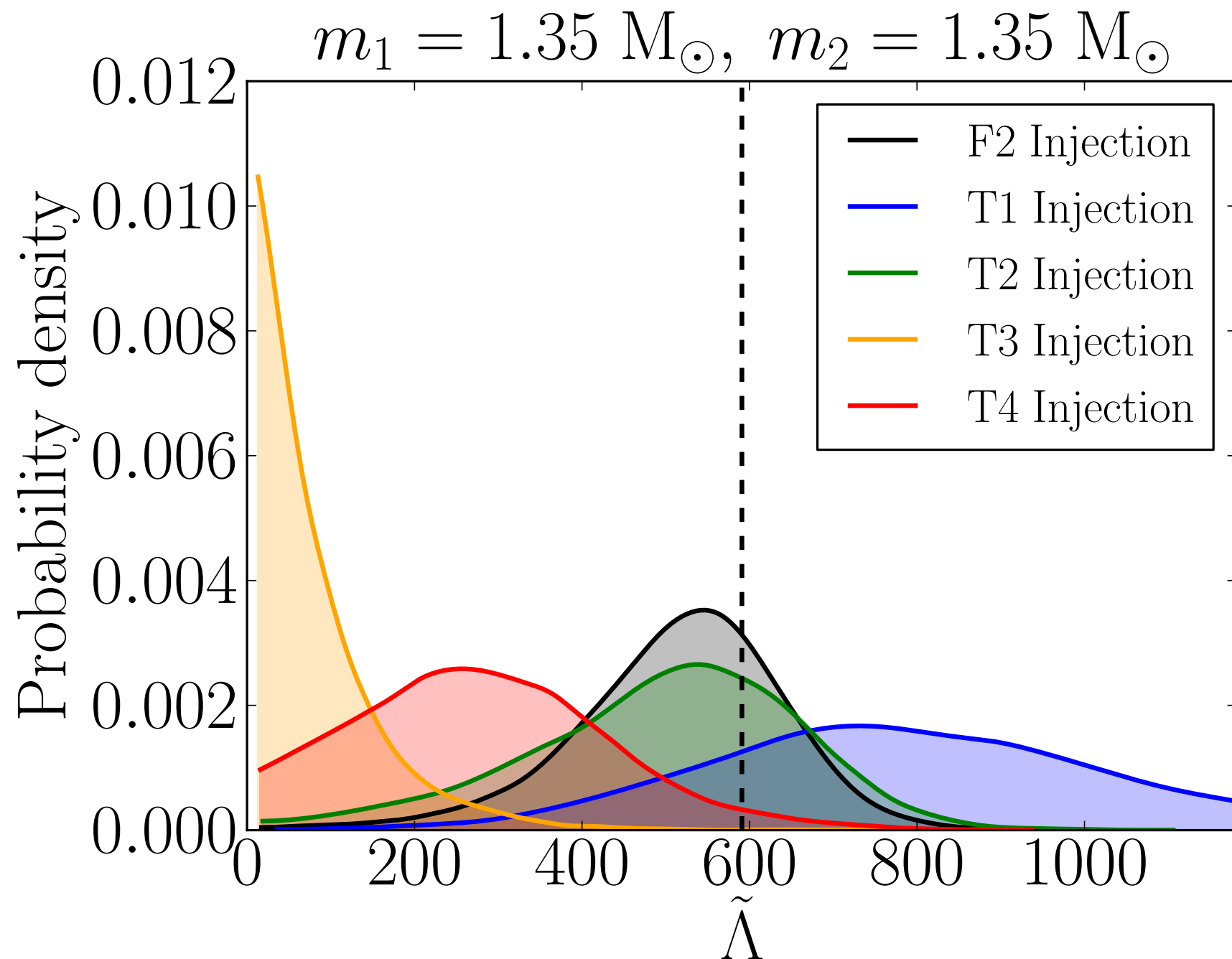
- Phase difference between 3PN and 3.5PN:  $\sim 1$  GW cycle
- At a given post-Newtonian order, there are many ways to calculate the phase
  - Approximants: Taylor **T1**, ..., Taylor **T4**, Taylor **F2**
  - Phase difference between approximants:  $\sim 1$  GW cycle



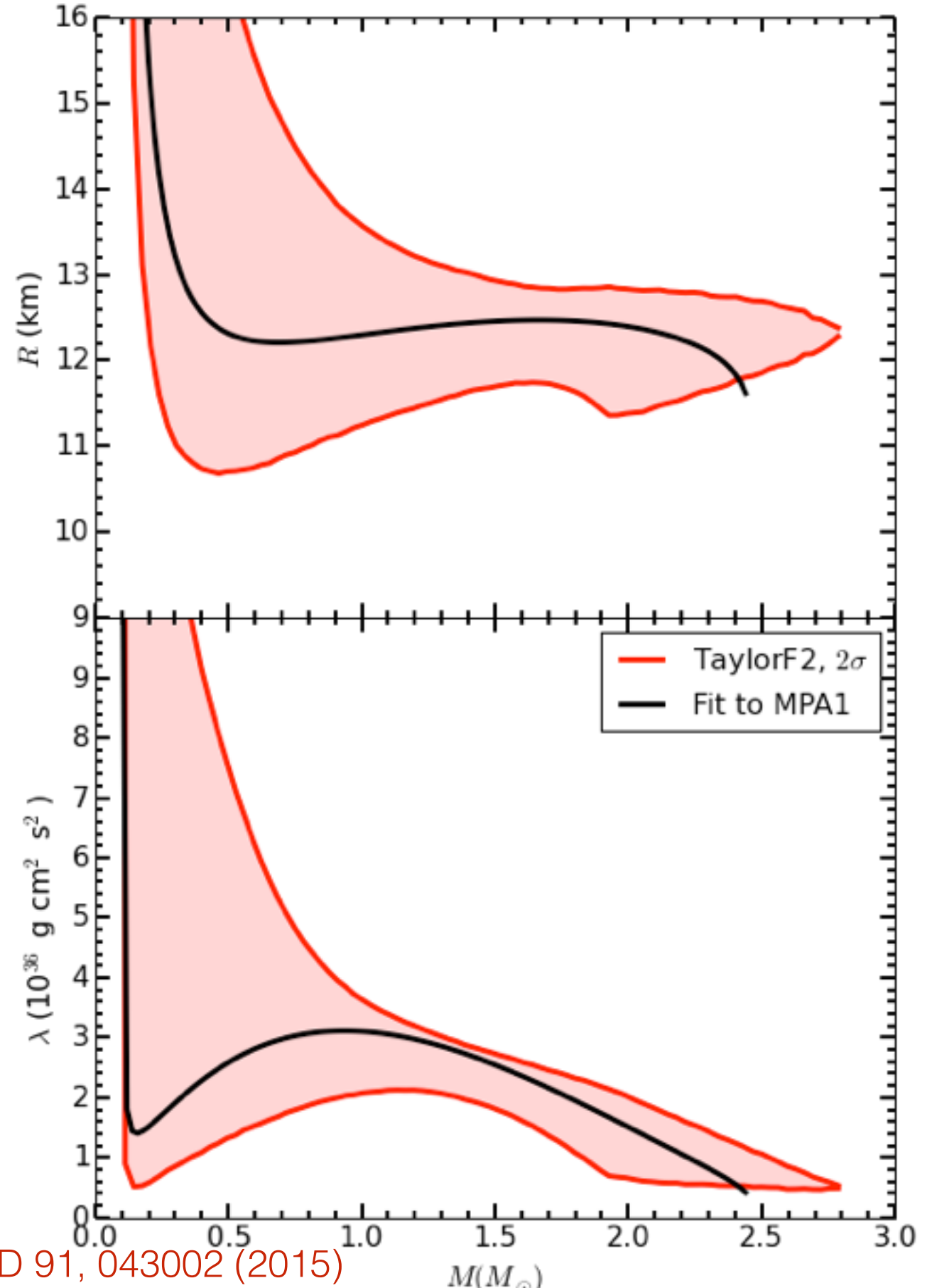
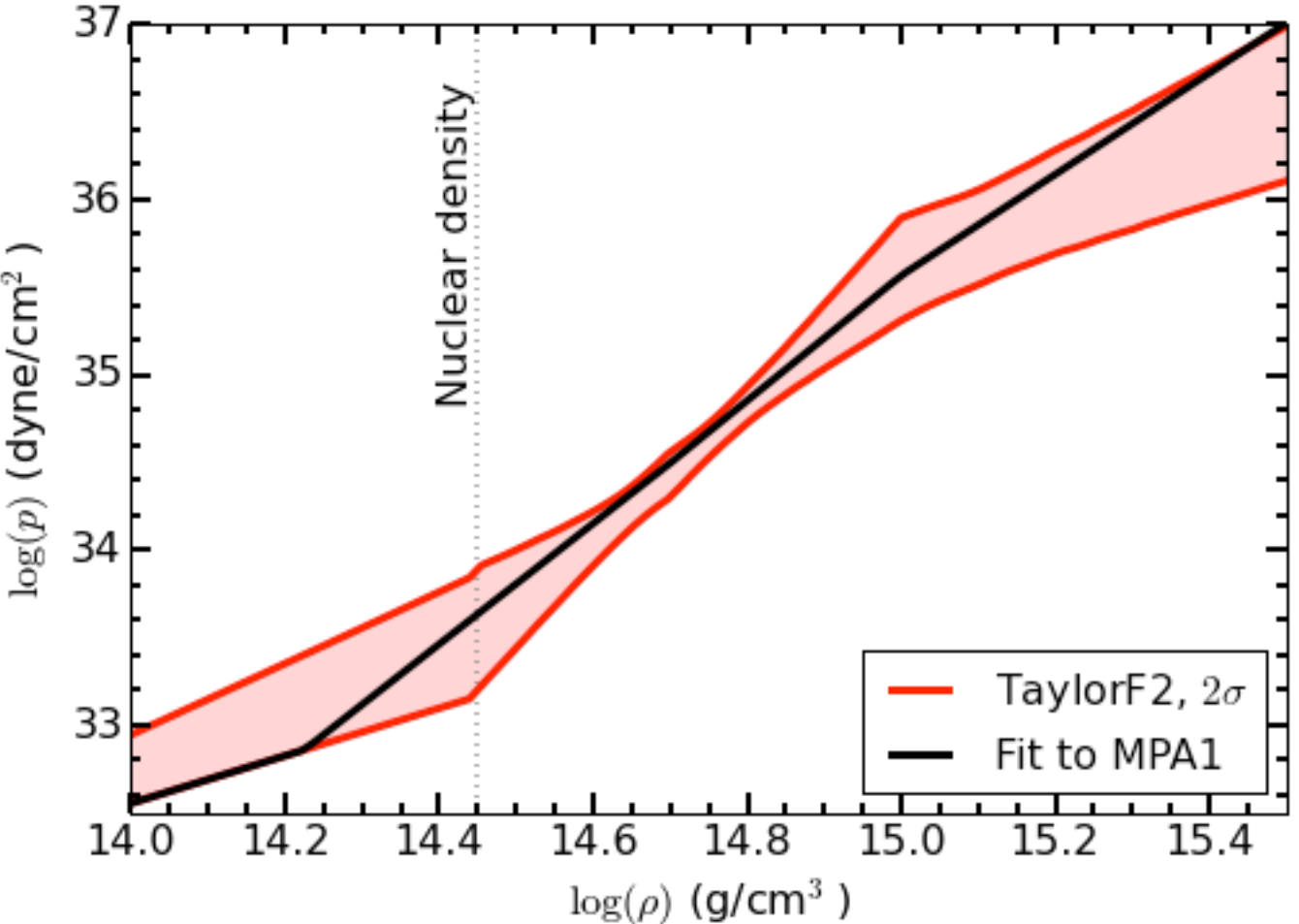


# Is the post-Newtonian waveform accurate enough? **No**

- Model used for constructing waveform template has dramatic effect on recovered tidal parameter
- Used TaylorF2 as waveform template

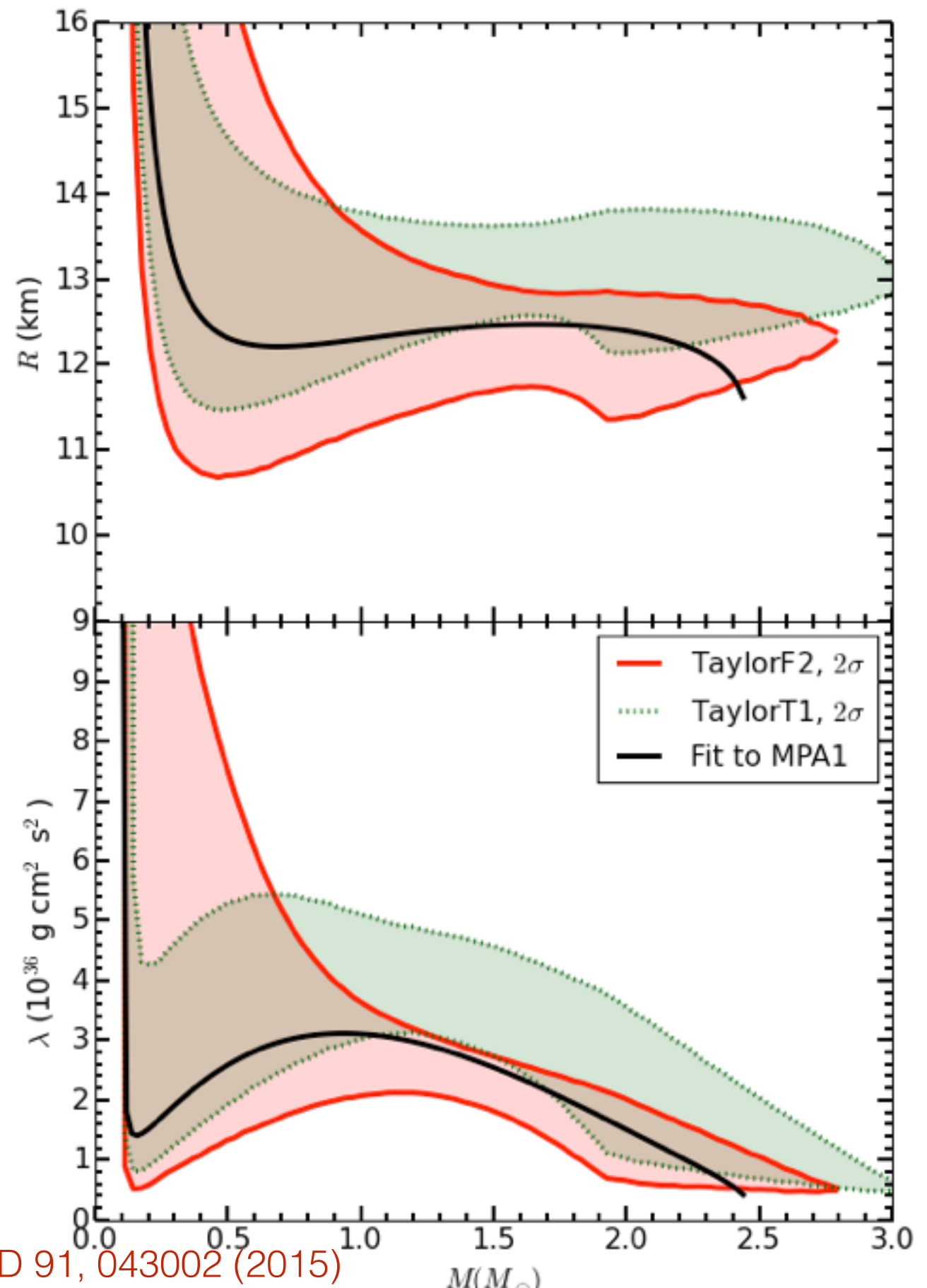
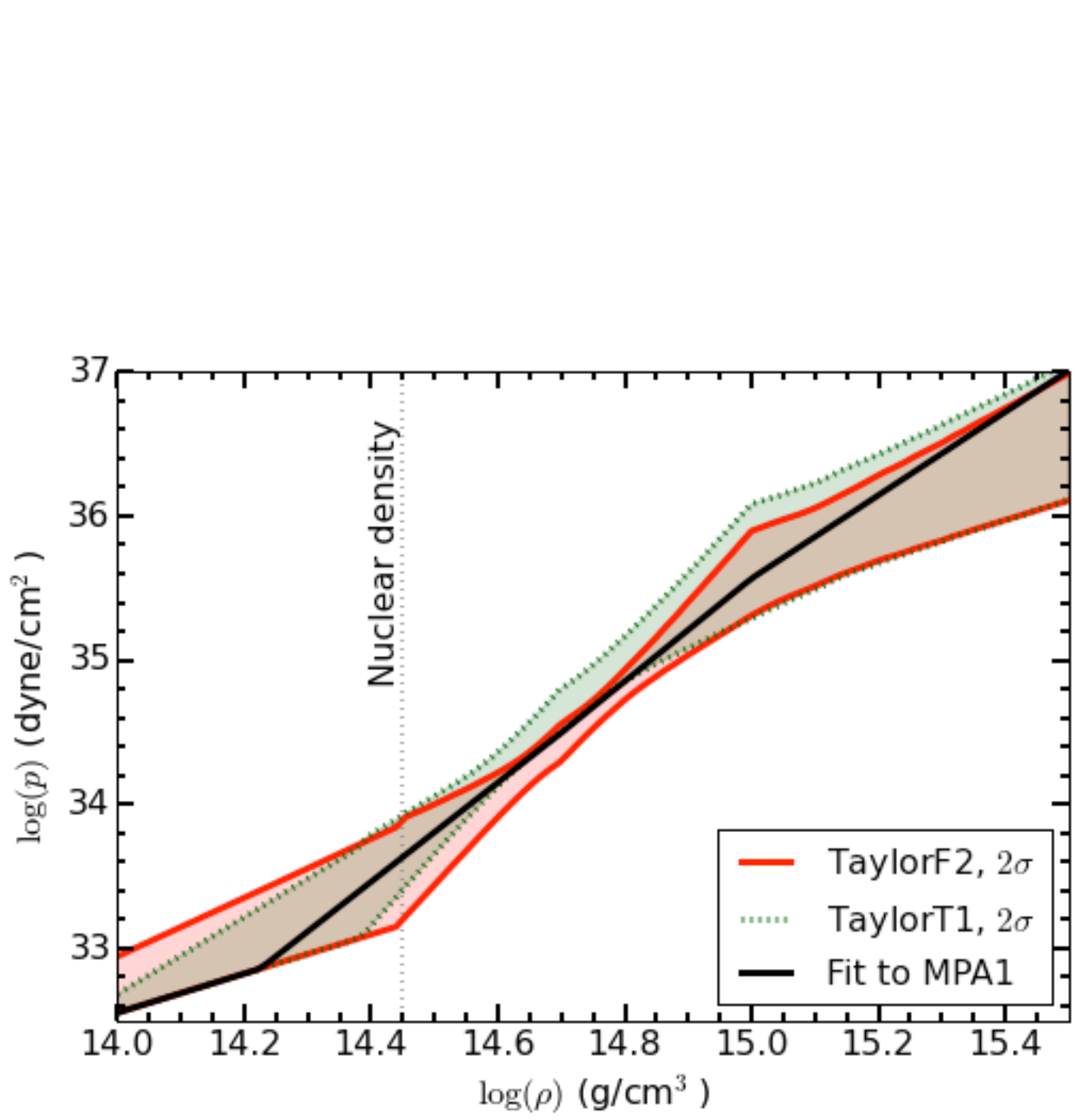


# Is the post-Newtonian waveform accurate enough? **No**



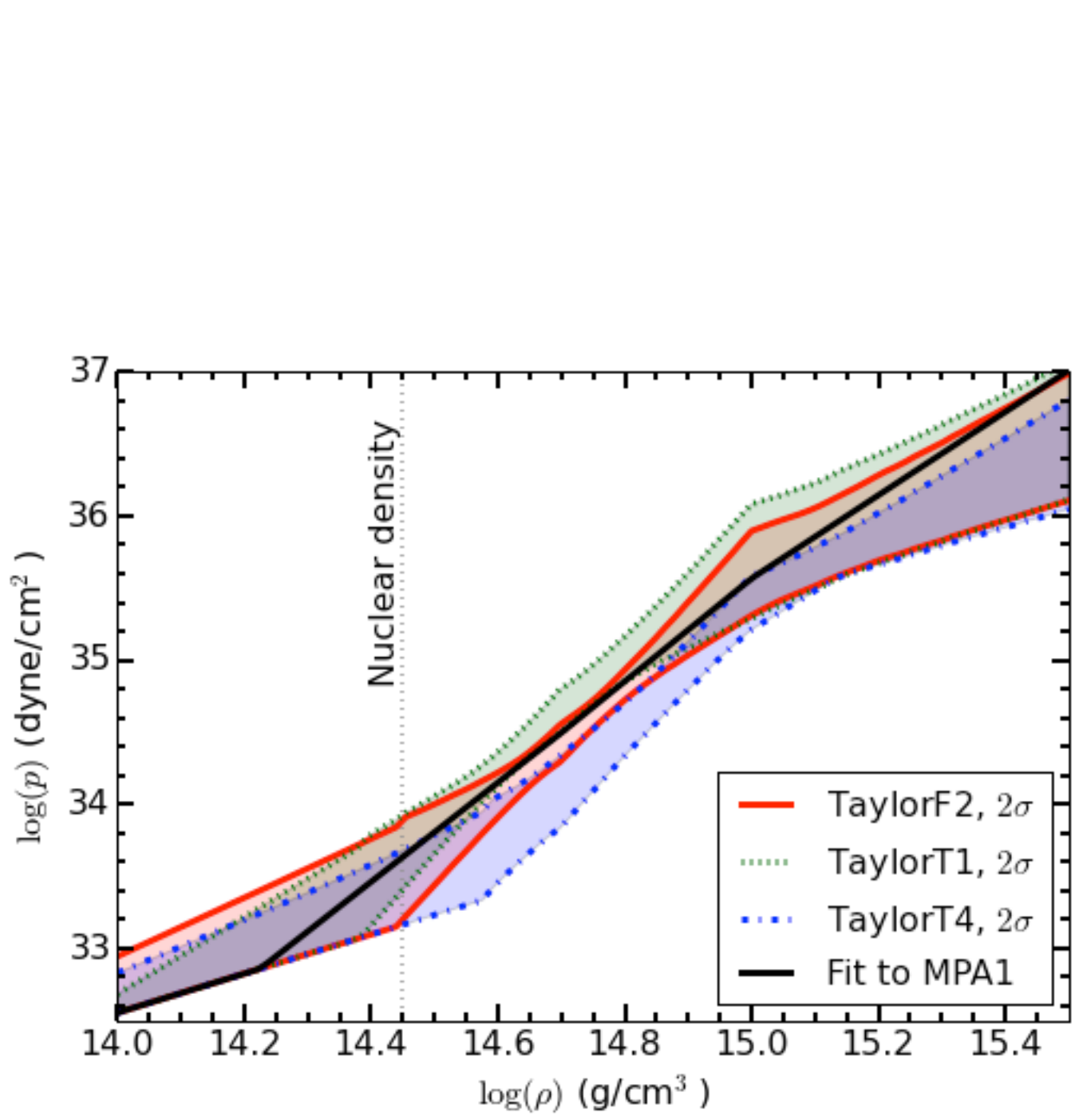
- 95% credible regions

# Is the post-Newtonian waveform accurate enough? **No**

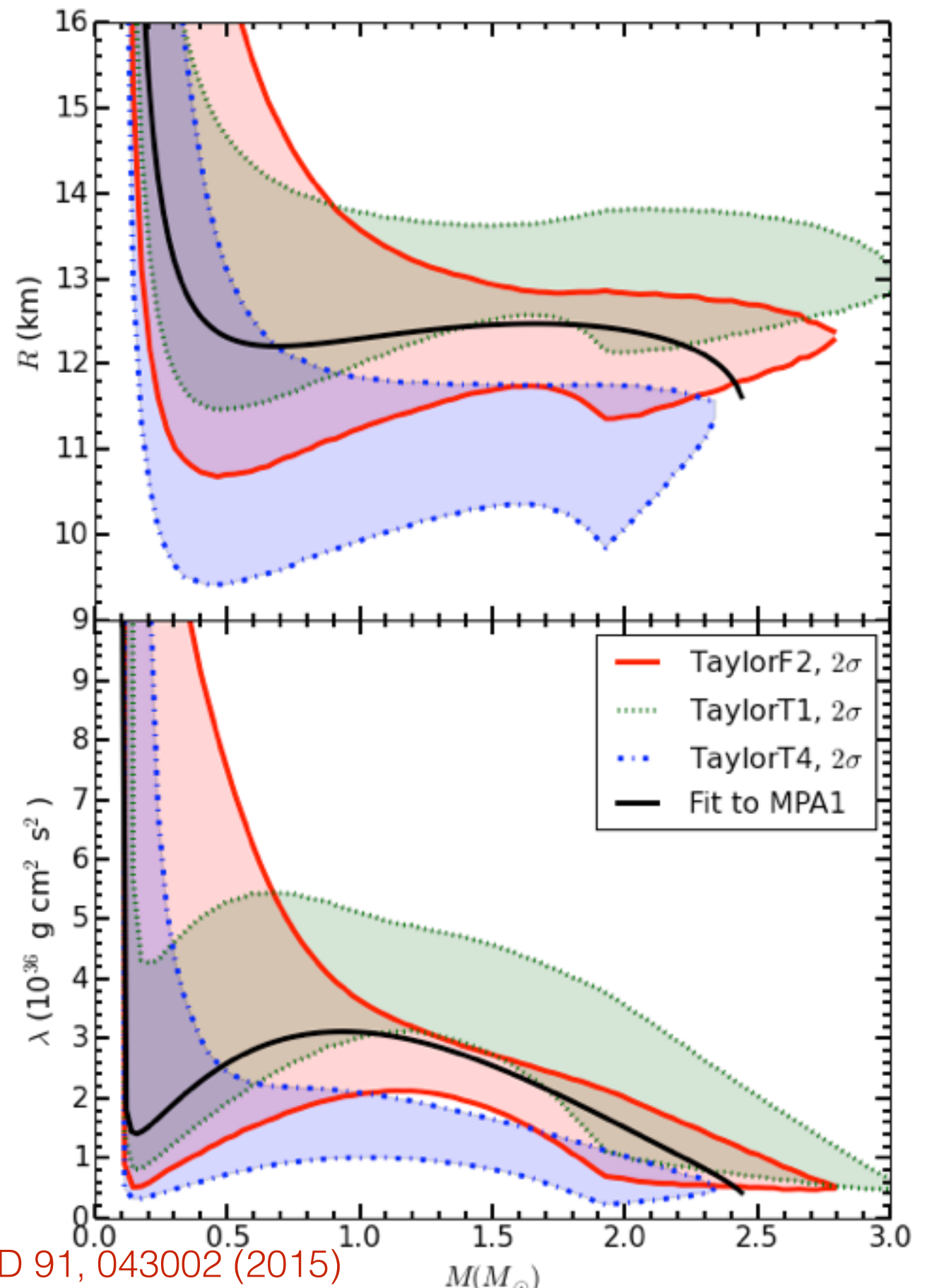


- 95% credible regions

# Is the post-Newtonian waveform accurate enough? **No**



- 95% credible regions

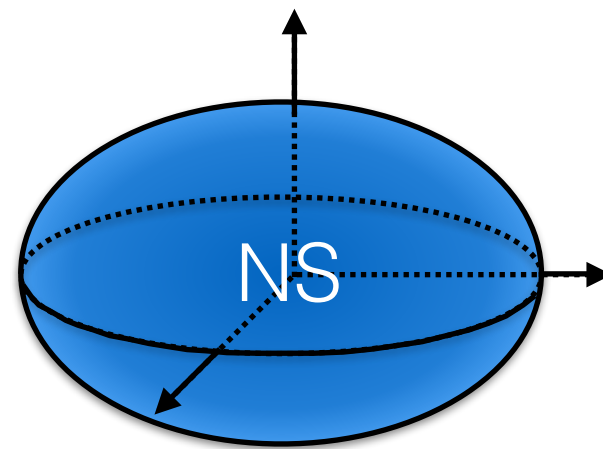
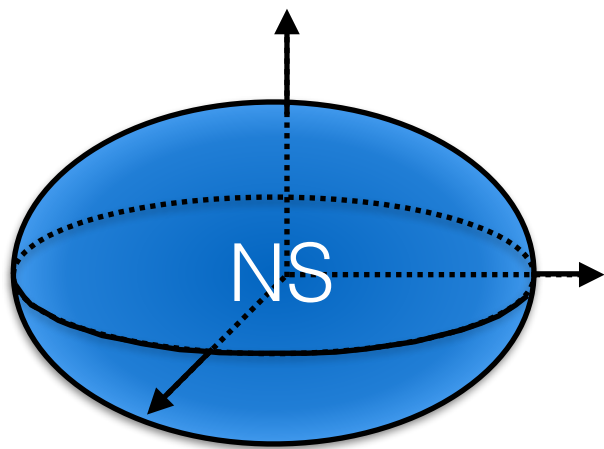


Constructing a more accurate  
waveform model



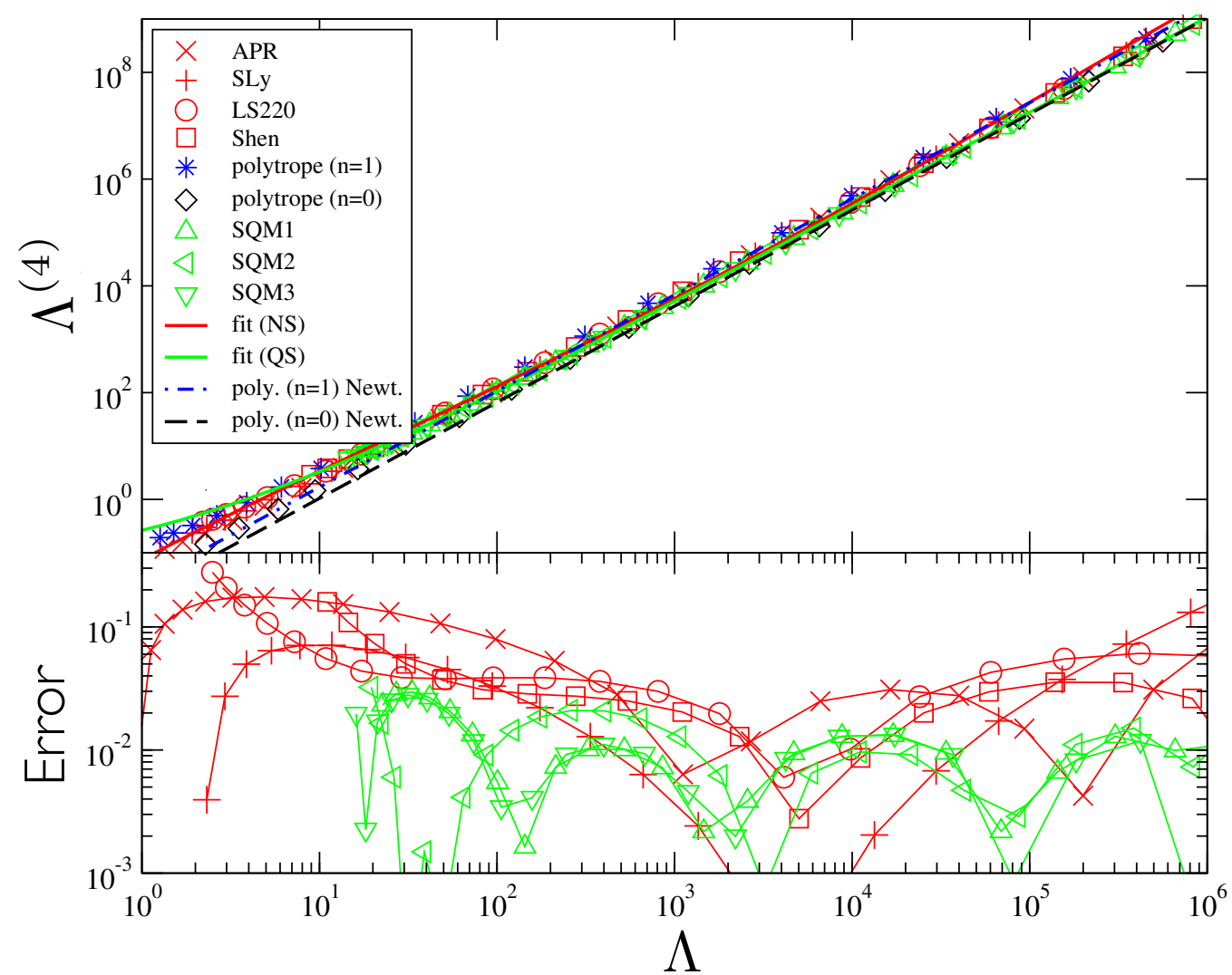
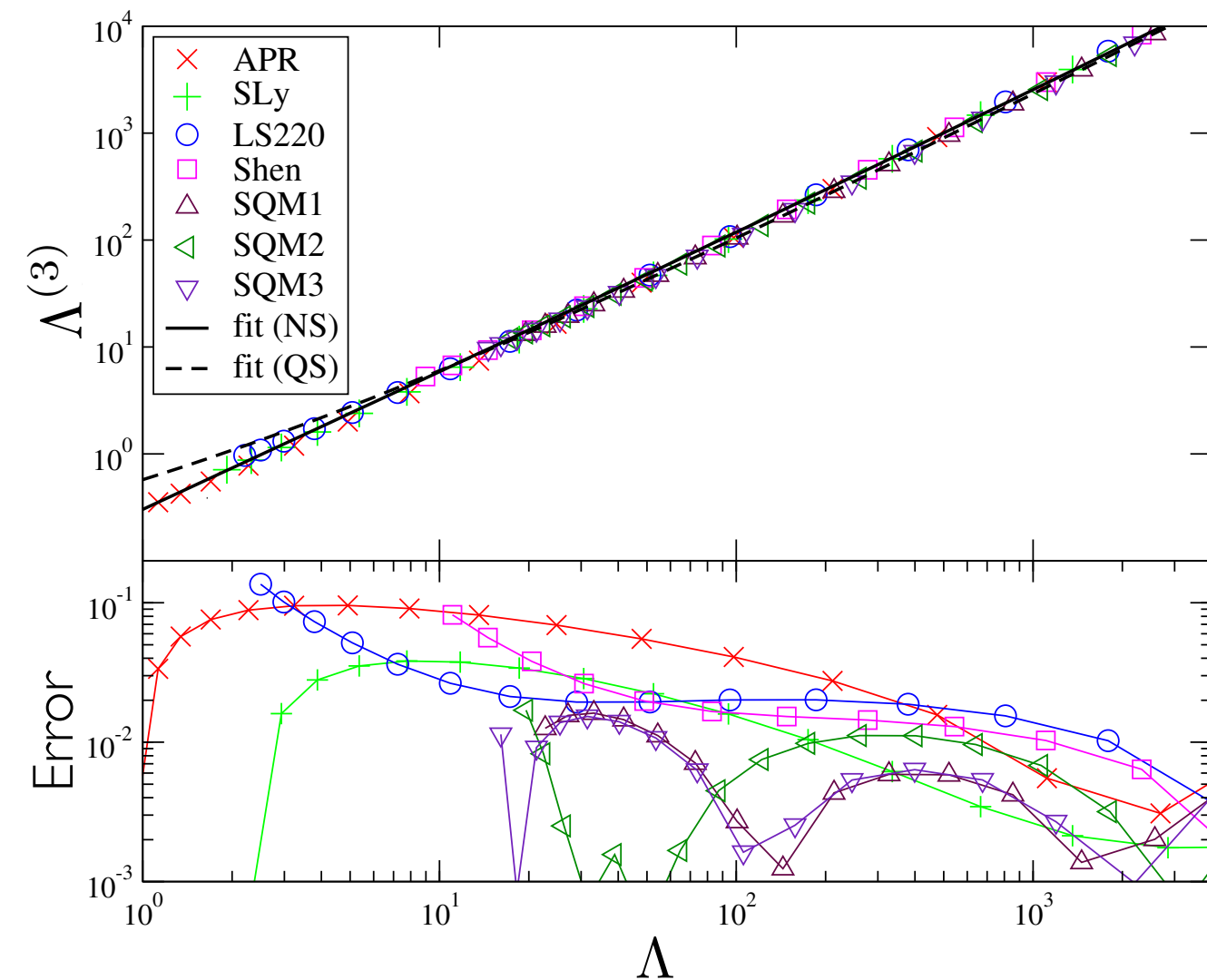
# Additional matter effects

- Tidal fields  $\mathcal{E}_{ij}\dots$  from companion star induce quadrupole  $Q_{ij}$  and higher order multipoles
  - Quadrupole:  $Q_{ij} = -\Lambda(\text{EOS}, m)m^5 \mathcal{E}_{ij}$  5PN
  - Octopole:  $O_{ijk} = -\Lambda^{(3)}(\text{EOS}, m)m^7 \mathcal{E}_{ijk}$  7PN
  - Hexadecapole:  $H_{ijkl} = -\Lambda^{(4)}(\text{EOS}, m)m^9 \mathcal{E}_{ijkl}$  9PN



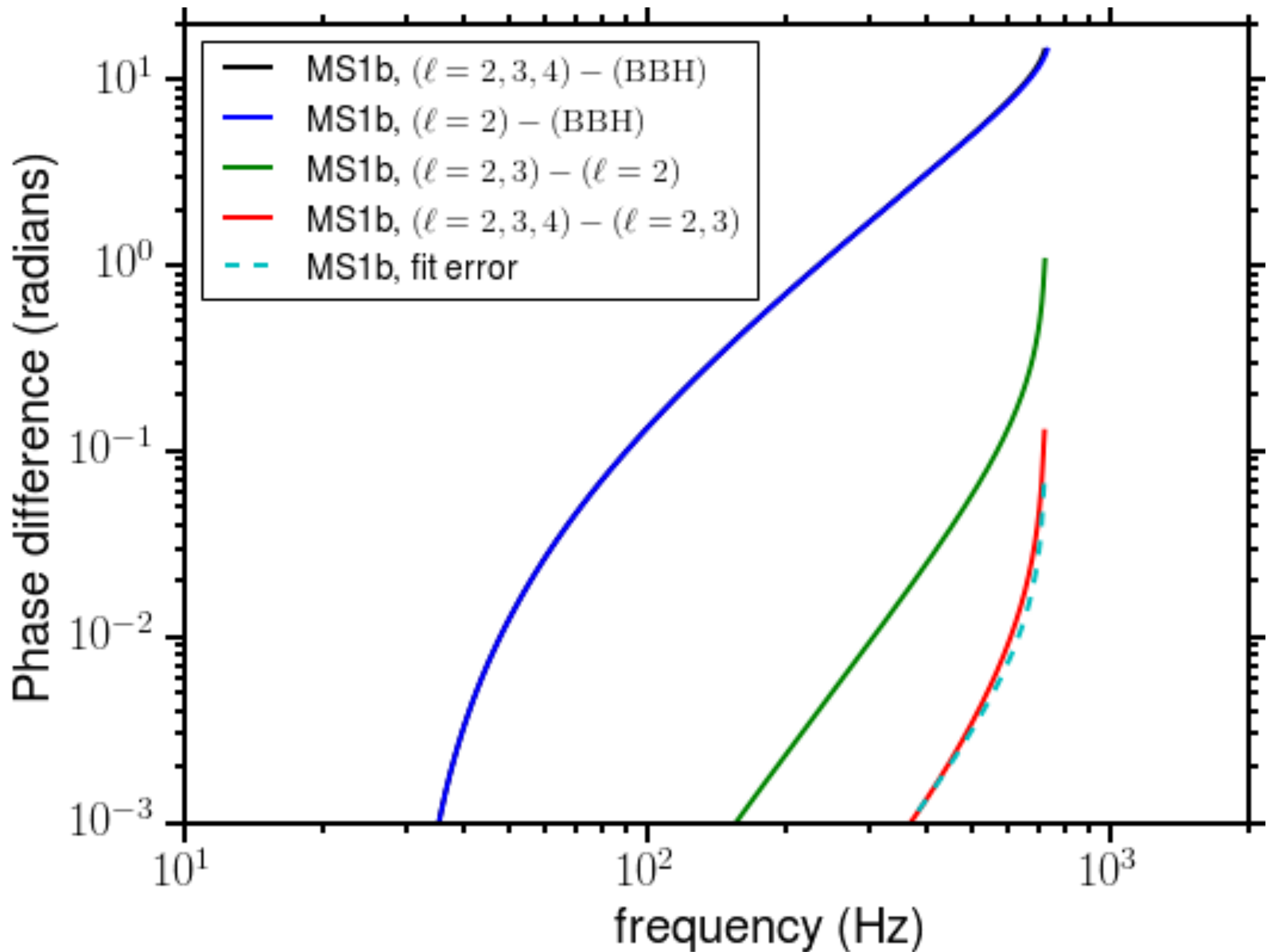
# Additional matter effects

- Most NS properties are correlated in a nearly EOS-independent way
- There is effectively only one EOS-dependent parameter during inspiral



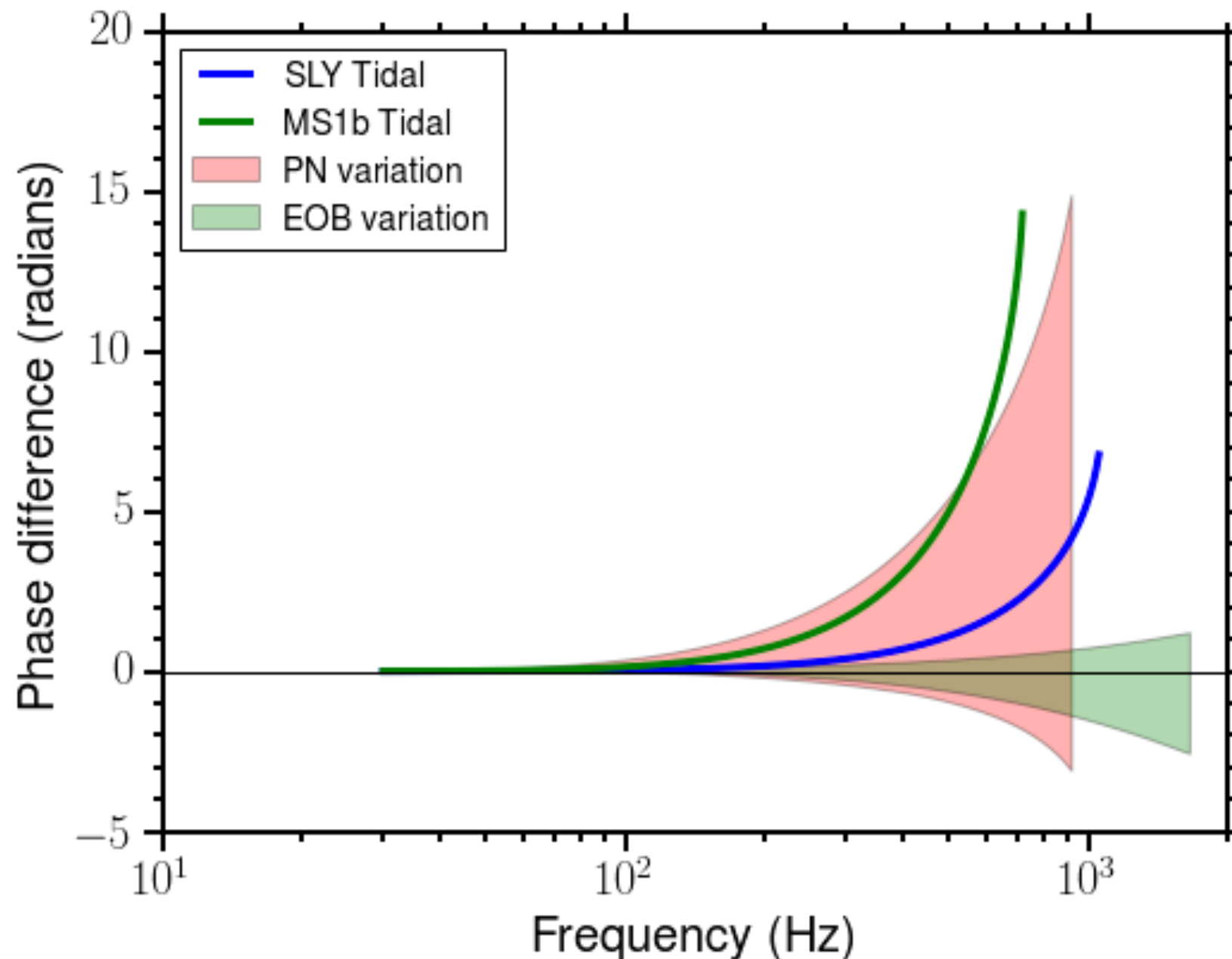
Kent Yagi. PRD 89, 043011 (2014)

# Additional matter effects



# More accurate BNS waveform

- Effective one body (EOB) waveforms have significantly smaller uncertainties
- Take  $\sim 10$  minutes to evaluate starting at 10Hz
  - Too slow for parameter estimation



4 different PN  
waveform models

6 different EOB  
waveform models

# Reduced order model

- Construct a set of orthonormal basis functions that approximate any waveform  $h(t; \vec{\theta}) = A(t; \vec{\theta})e^{i\Phi(t; \vec{\theta})}$  in the parameter space

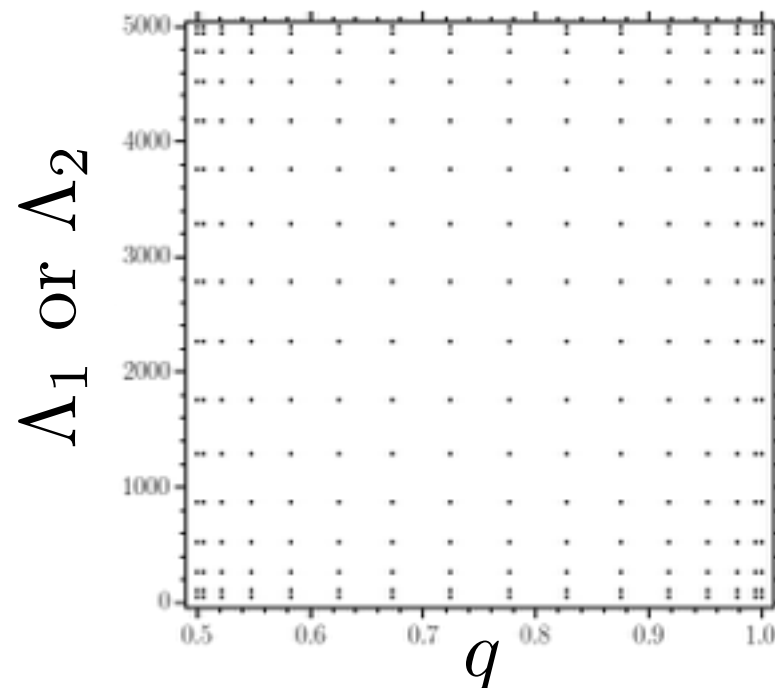
$$A(t; \vec{\theta}) \approx \sum_{i=1}^{N_A} c_i(\vec{\theta}) \hat{e}_i(t) \quad \Phi(t; \vec{\theta}) \approx \sum_{i=1}^{N_\Phi} c_i(\vec{\theta}) \hat{e}_i(t)$$

# Reduced order model

- Construct a set of orthonormal basis functions that approximate any waveform  $h(t; \vec{\theta}) = A(t; \vec{\theta})e^{i\Phi(t; \vec{\theta})}$  in the parameter space

$$A(t; \vec{\theta}) \approx \sum_{i=1}^{N_A} c_i(\vec{\theta}) \hat{e}_i(t) \quad \Phi(t; \vec{\theta}) \approx \sum_{i=1}^{N_\Phi} c_i(\vec{\theta}) \hat{e}_i(t)$$

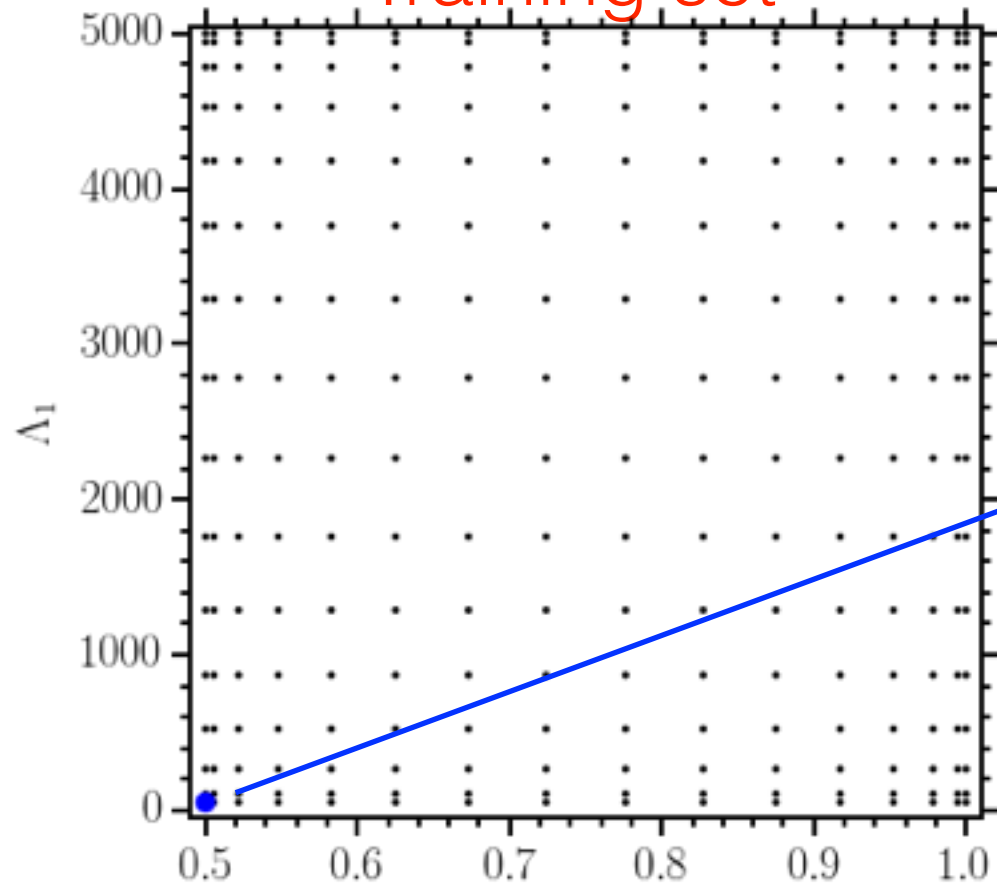
- 8-dimensional parameter space (1 mass and 3 tidal parameters for each NS)
  - Can use mass ratio  $q$  and rescale waveform with total mass  $M$
  - Use 10% accurate fits for  $l=3,4$  tidal parameter in terms of  $l=2$  tidal parameter
  - Results in 3 dimensional parameter space  $\vec{\theta} = \{q, \Lambda_1, \Lambda_2\}$



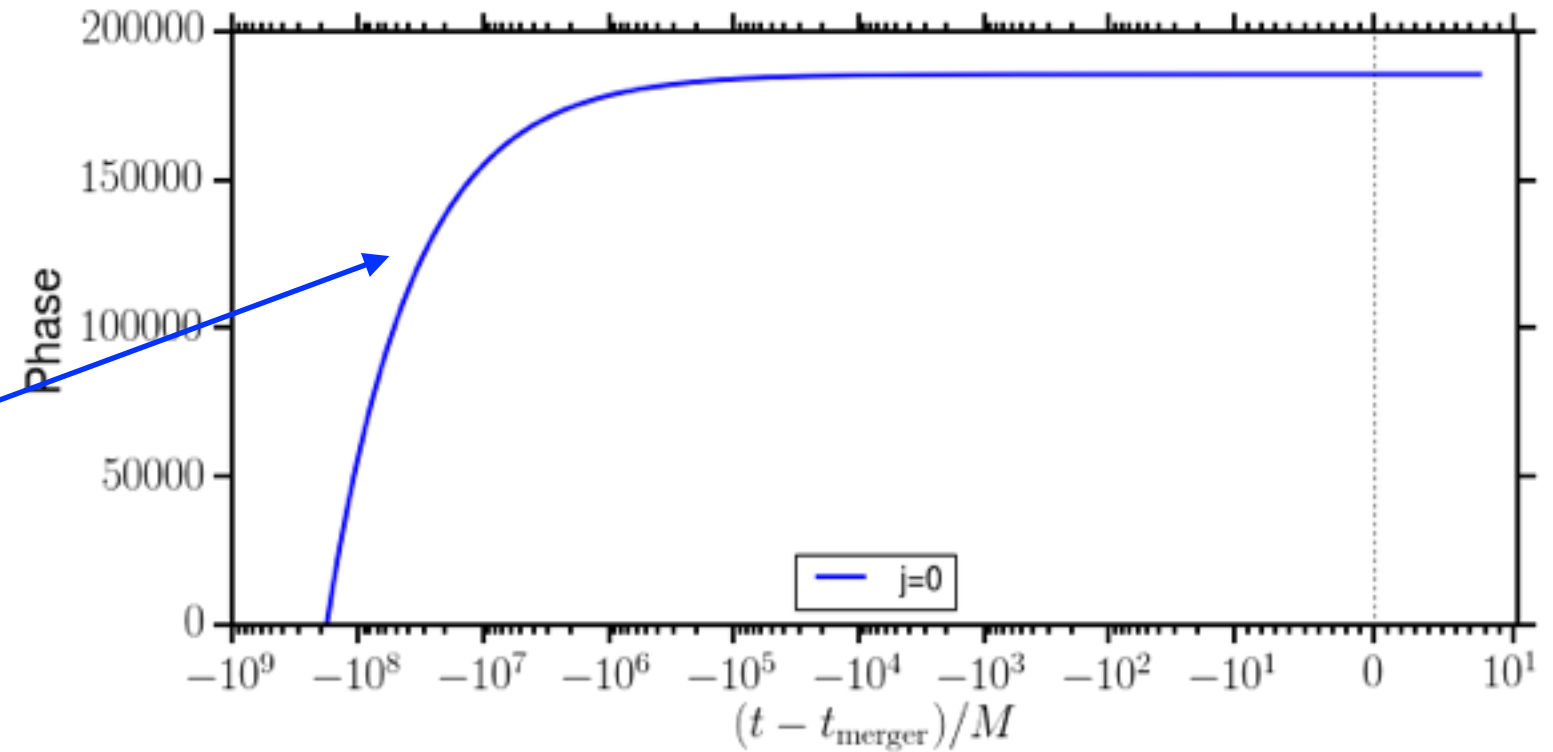


# Greedy algorithm

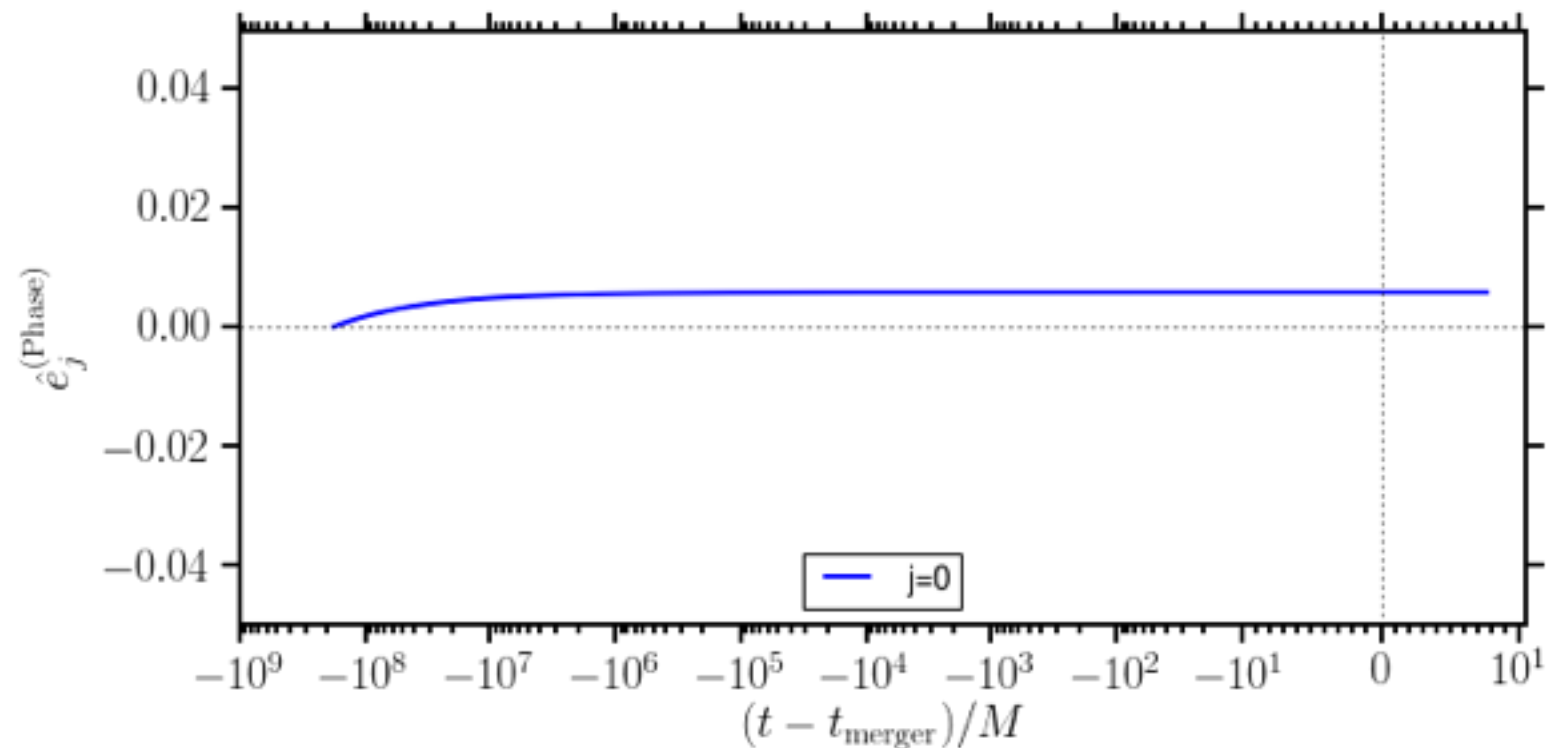
Training set



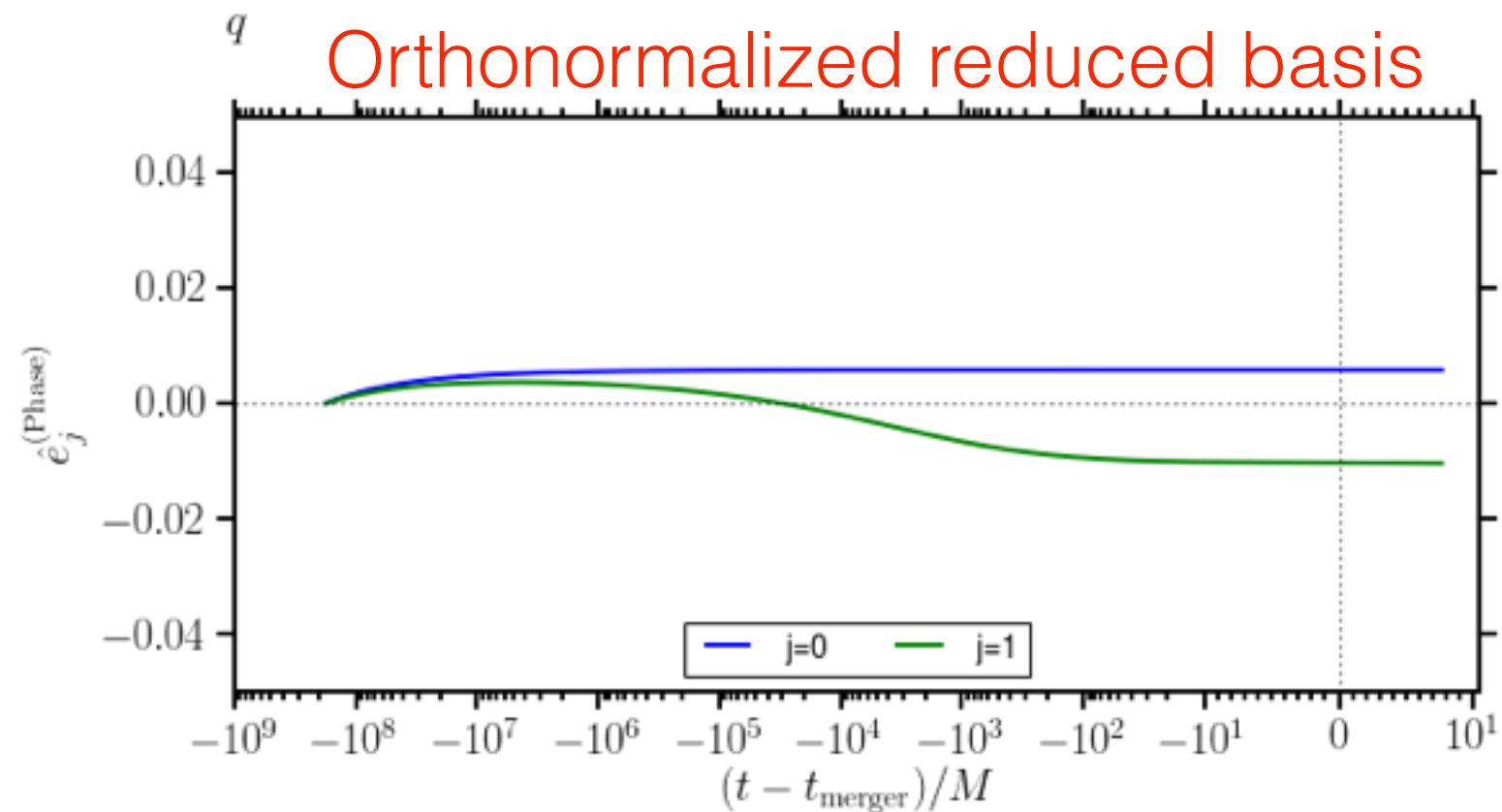
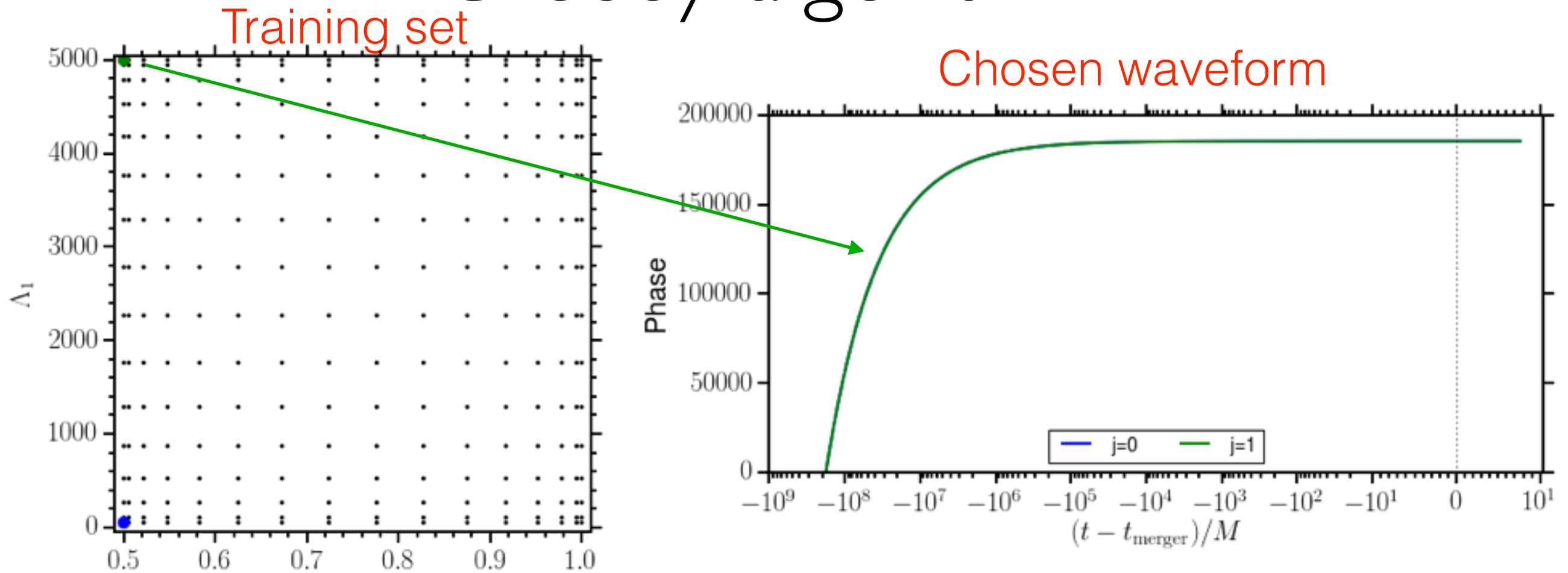
Chosen waveform



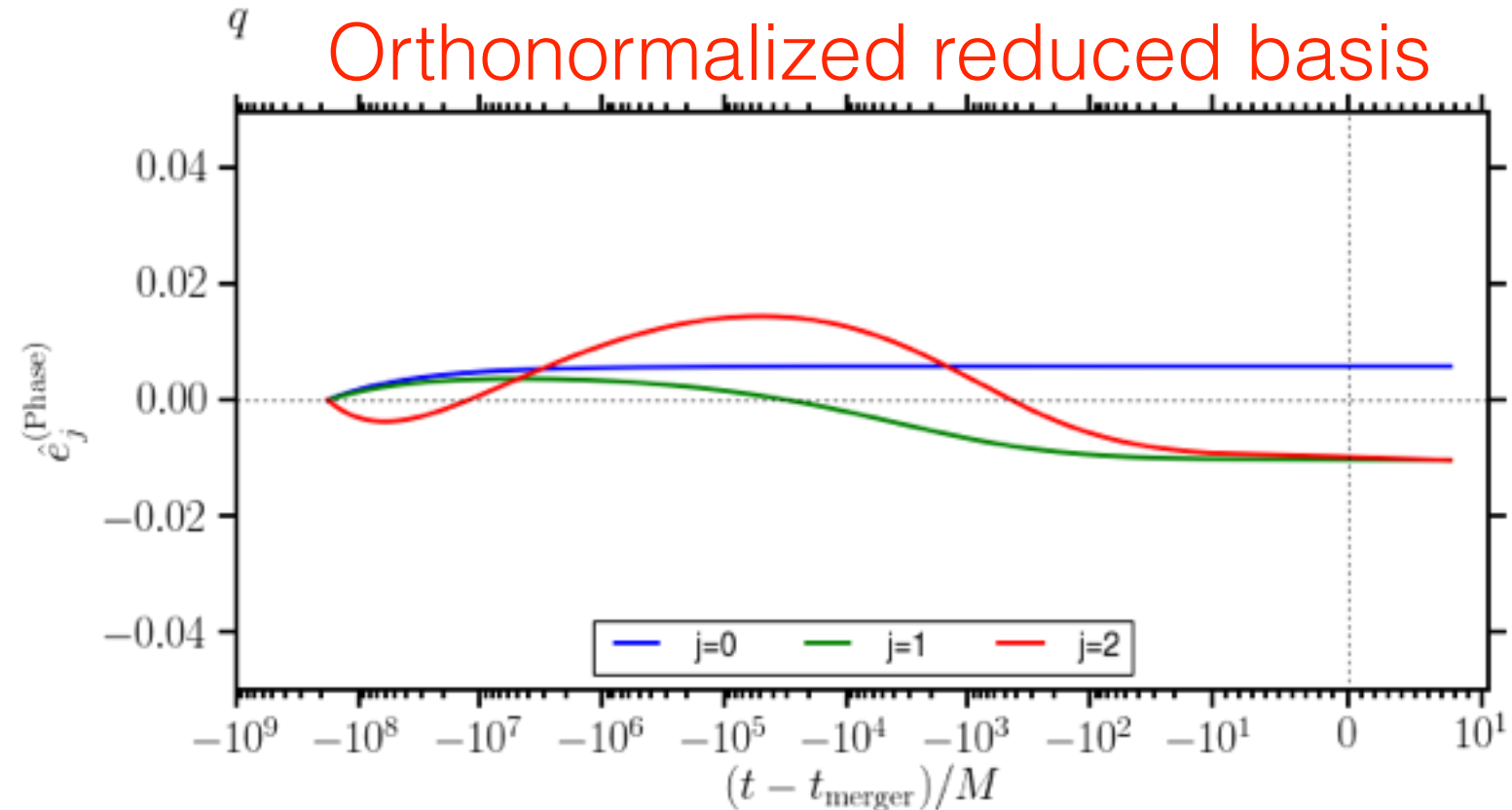
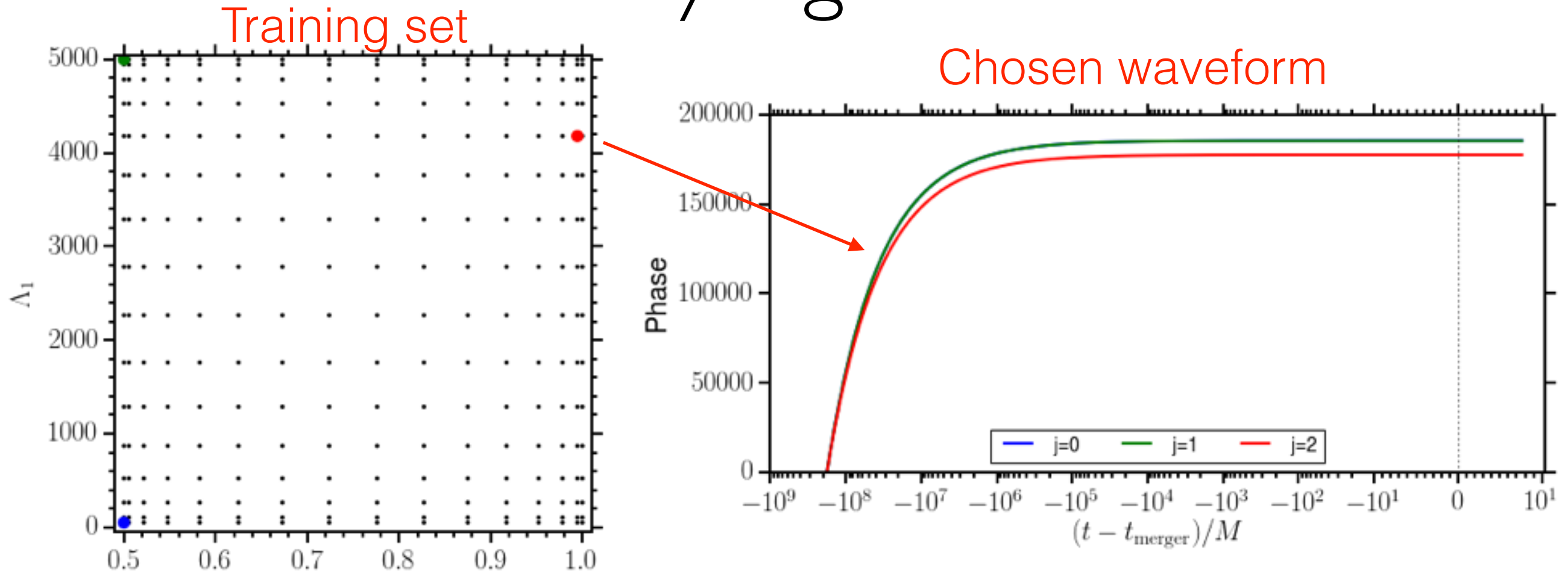
Orthonormalized reduced basis



# Greedy algorithm

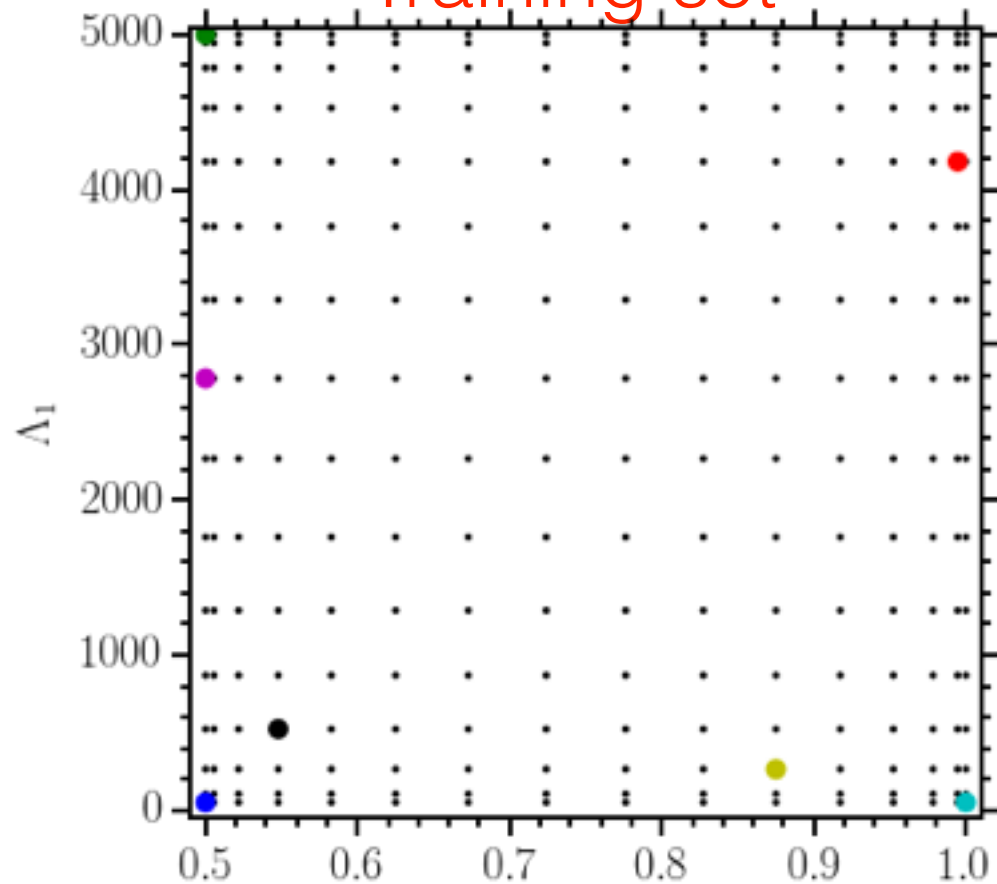


# Greedy algorithm

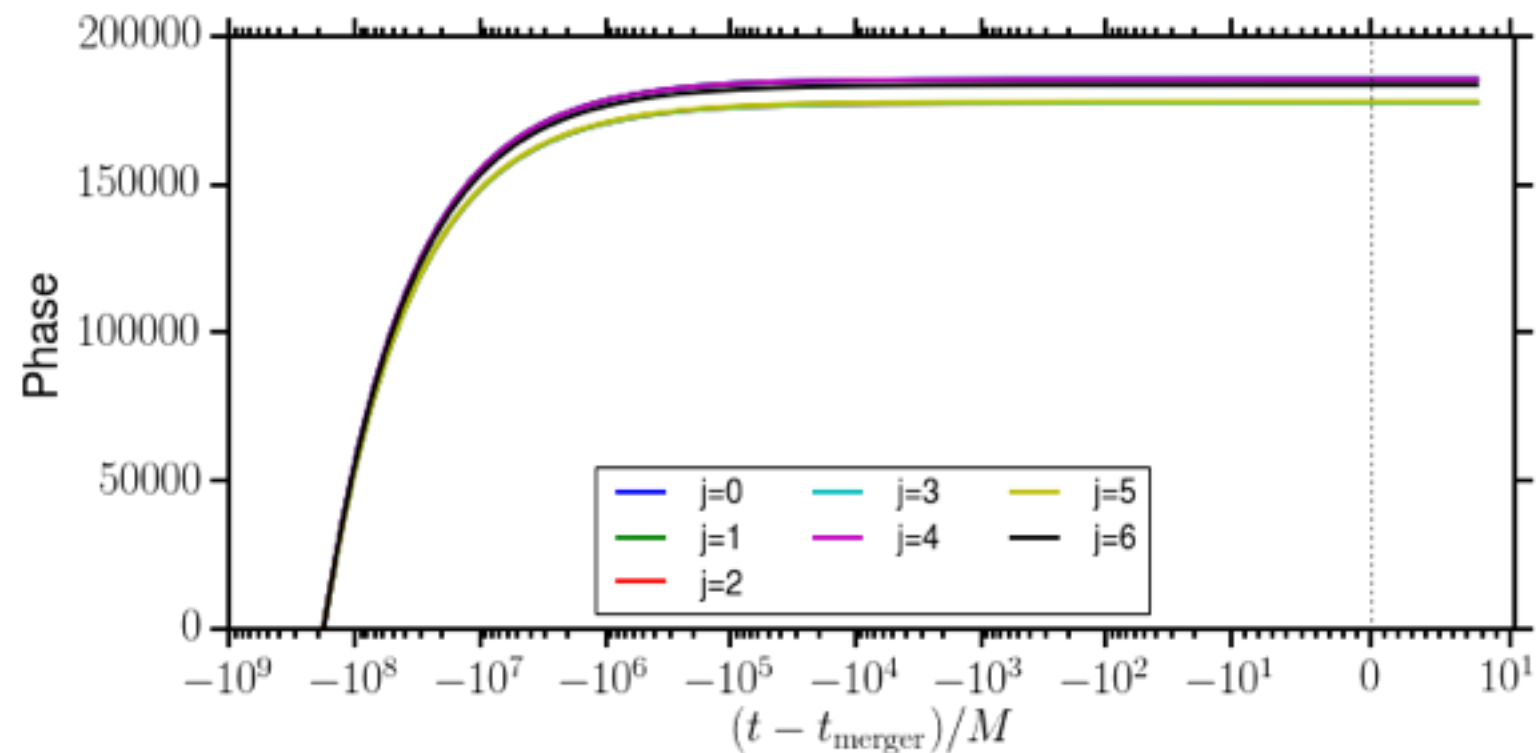


# Greedy algorithm

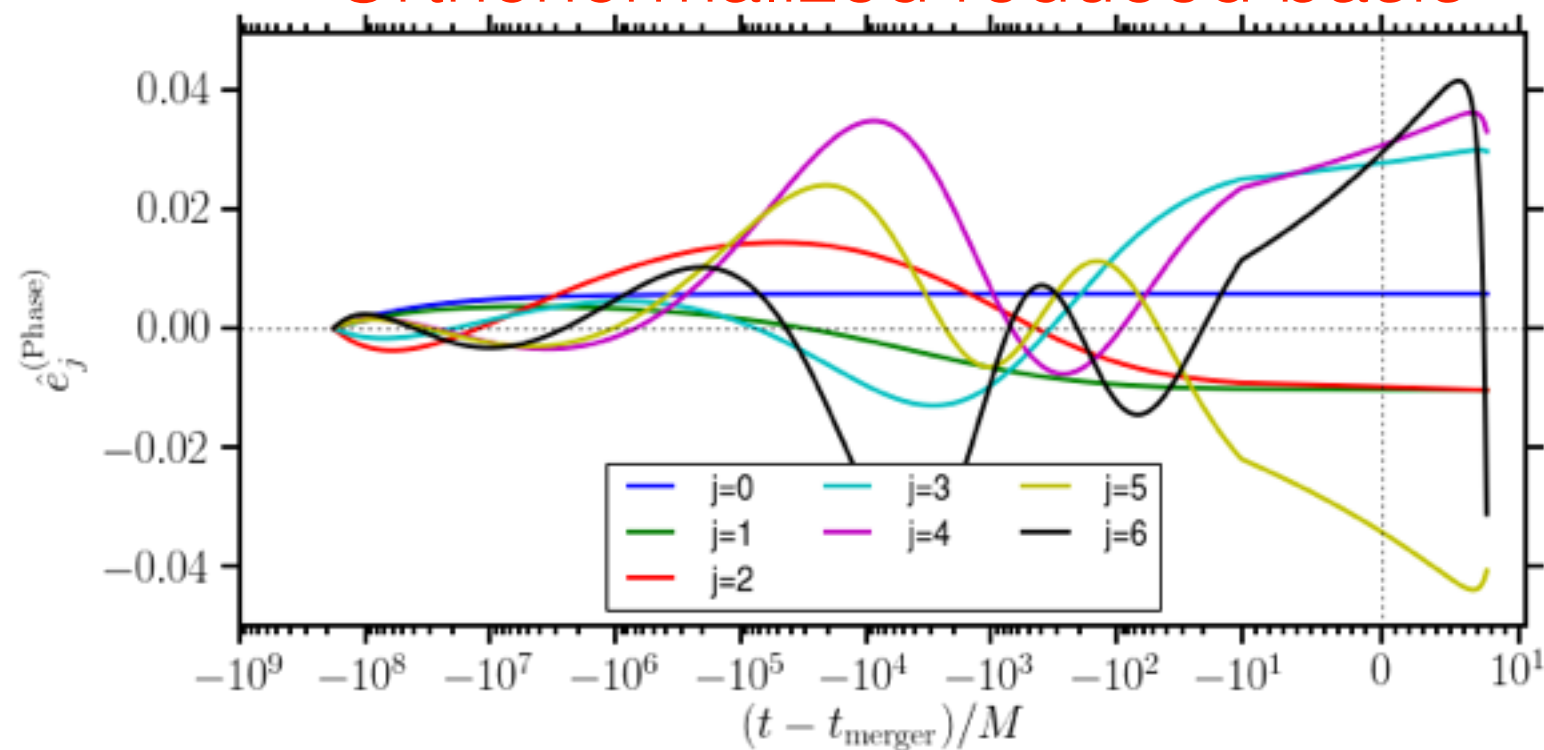
Training set



Chosen waveform



Orthonormalized reduced basis

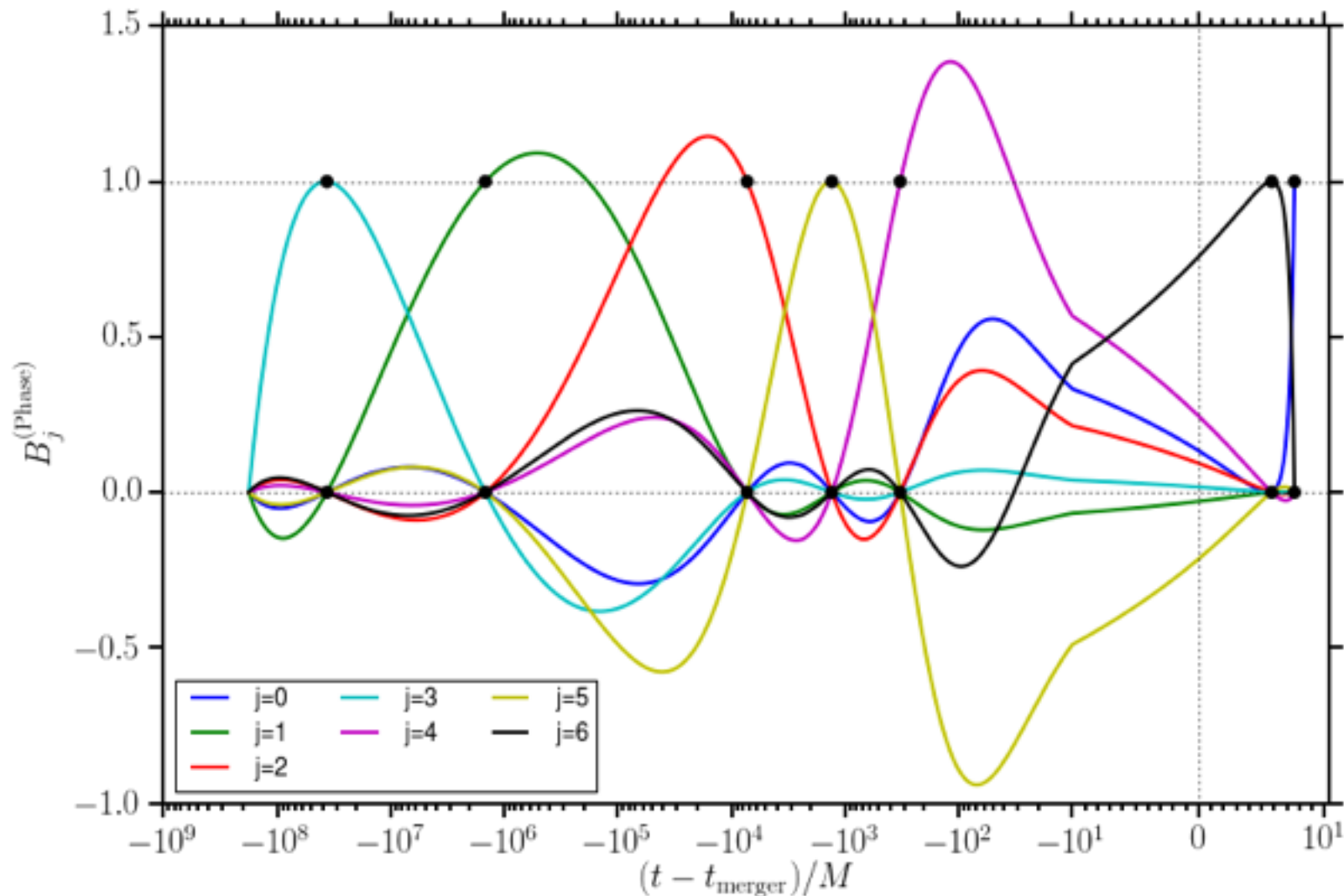


# Empirical interpolation method

- Can re-express phase basis  $\{e_i(t)\}$  in terms of waveform evaluated at empirical times  $T_j$  and empirical interpolants  $B_j(t)$

$$\Phi(t; \vec{\theta}) \approx \sum_{j=1}^n \Phi(T_j; \vec{\theta}) B_j(t)$$

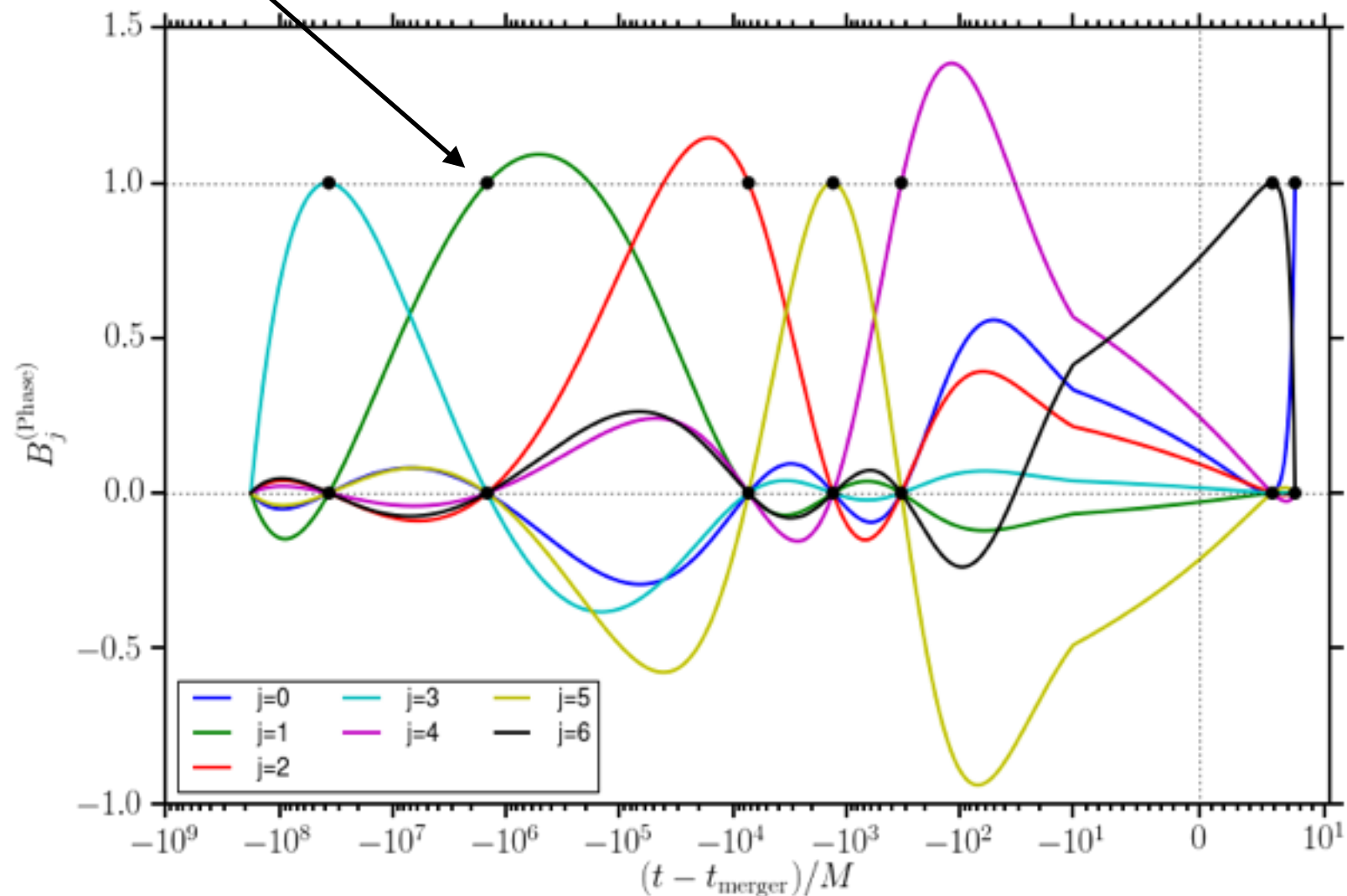
- where  $B_j(t) = \sum_{i=1}^n \hat{e}_i(t) (V^{-1})_{ij}$  and  $V_{ji} = \hat{e}_i(T_j)$



# Interpolating waveform parameters

- Waveform at each empirical node  $T_j$  interpolated with Chebyshev polynomials
- Coefficients  $b_{lmn}$  calculated with Gaussian quadrature

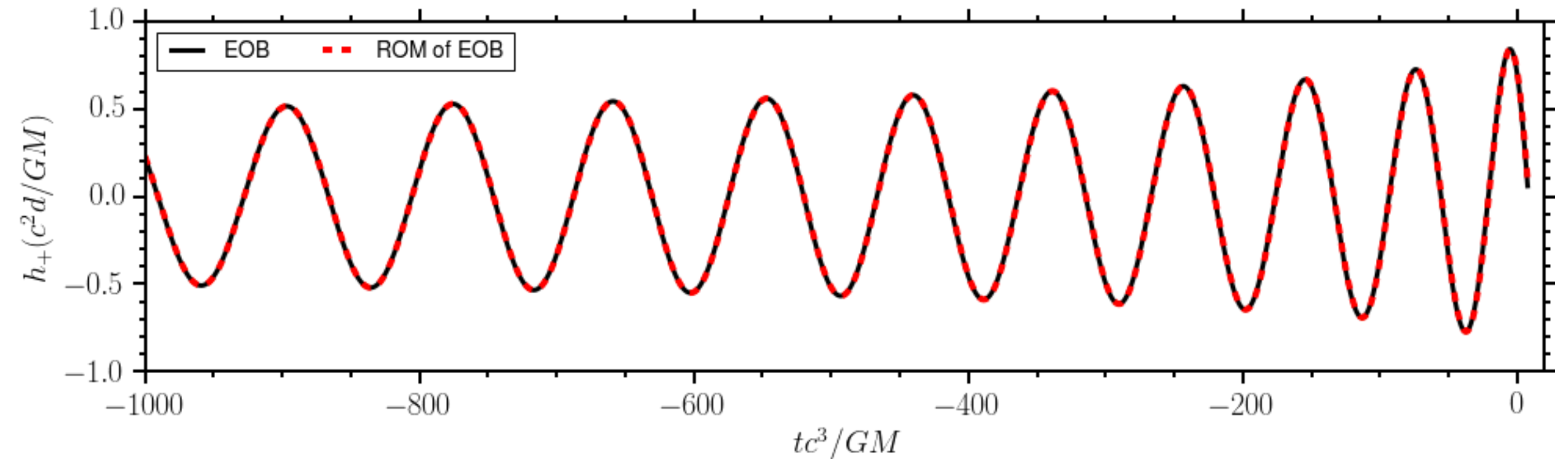
$$\Phi(T_j; \vec{\theta}) = \sum_l \sum_m \sum_n b_{lmn} T_l(q) T_m(\Lambda_1) T_n(\Lambda_2)$$





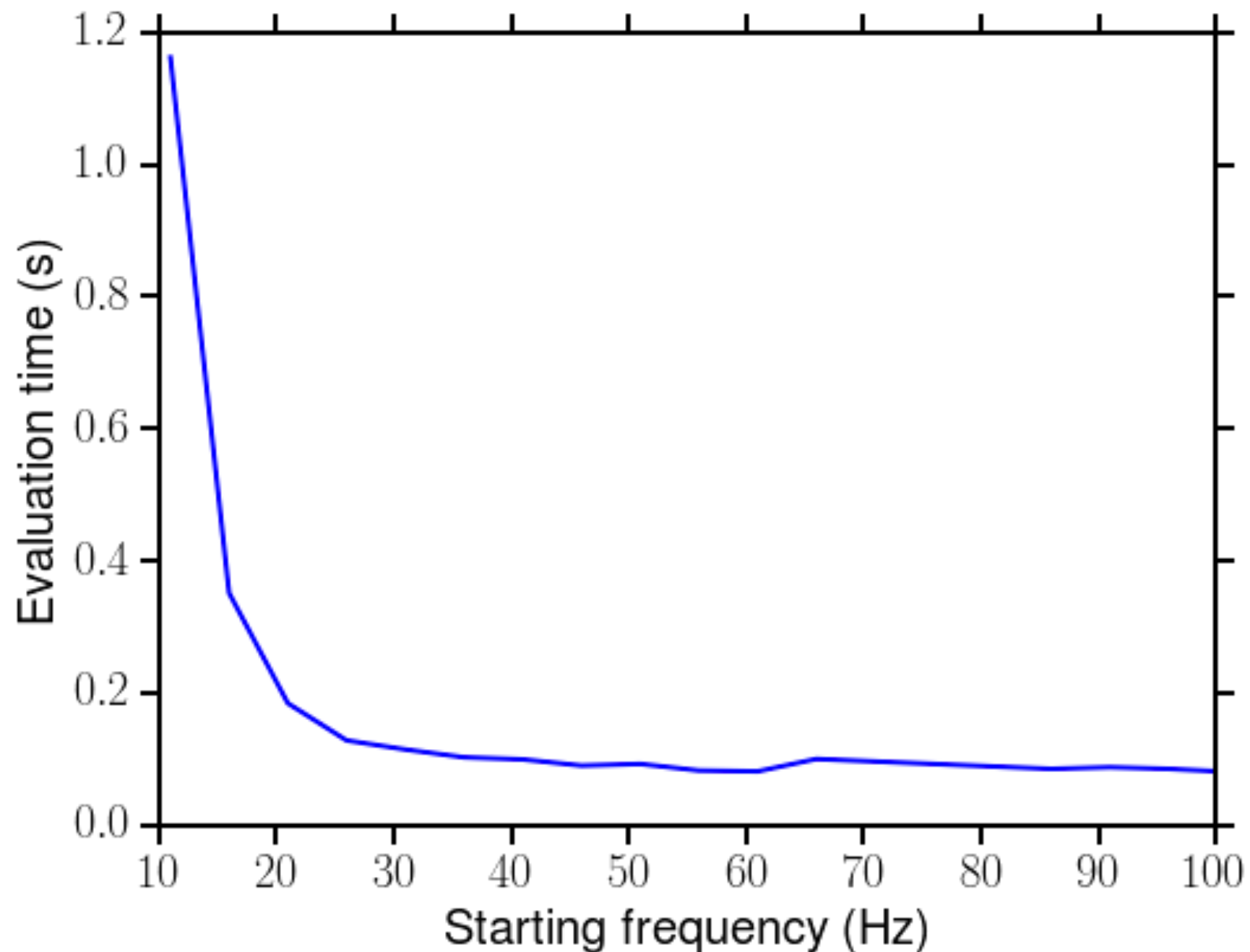
# Accuracy of ROM

- Compare ROM to 10,000 waveforms randomly sampled in parameter space
  - Maximum amplitude error:  $\sim 2\%$  (0.2% before last cycle)
  - Maximum phase error:  $\sim 0.04$  radians
    - Much less than the tidal effect of  $\sim 5-10$  radians



# Speed of ROM

- Implementation of ROM in LAL (written in C)
  - Faster than original Matlab tidal EOB code by a factor of  $\sim 1000$
  - Faster than all time-domain waveforms in LAL

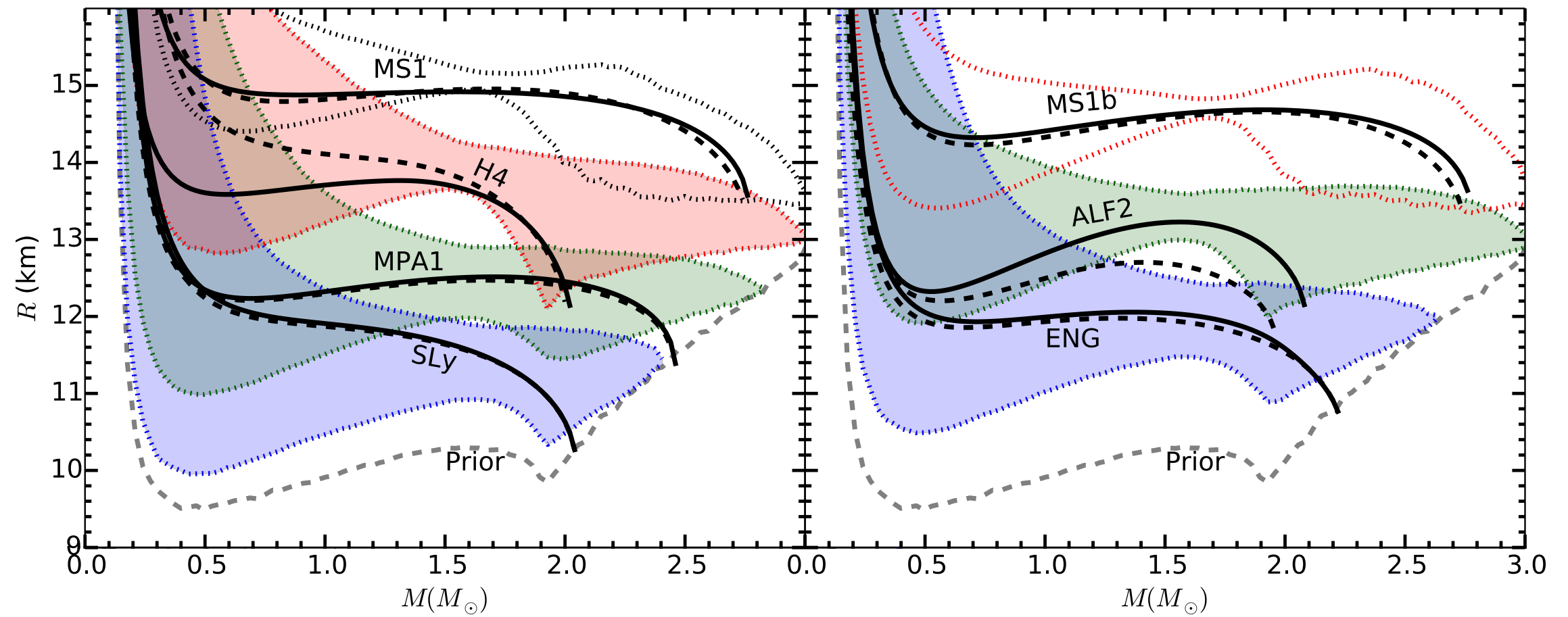


# Conclusions

- EOS and NS structure can be recovered with  $\sim 1$  year of BNS observations
- But, it's crucial that we have accurate and fast waveform models

Thank you

# Other EOS models



- 95% credible regions

# Other EOS models

