

# Incorporating neutron star physics into gravitational wave inference with neural priors

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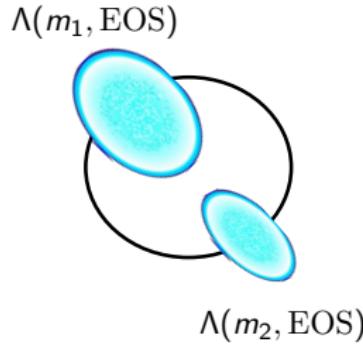
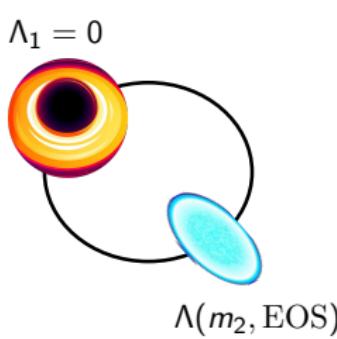
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# Motivation

- Posterior  $\propto$  likelihood  $\times$  prior
- Prior: usually agnostic, but can encode valuable information!
  - Theory
  - Observations outside of GW

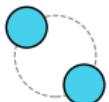
# Motivation

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- Prior: usually agnostic, but can encode valuable information!
  - Theory
  - Observations outside of GW
- Neutron stars – key GW observables?
  - Masses  $m_1, m_2$
  - Tidal deformabilities  $\Lambda_1, \Lambda_2 \leftarrow$  equation of state (EOS)

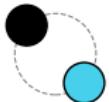


# Key idea

Source



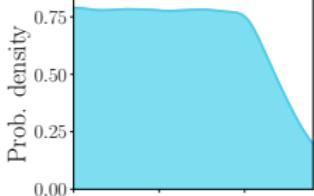
BNS



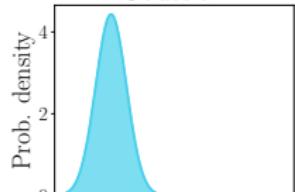
NSBH

Population

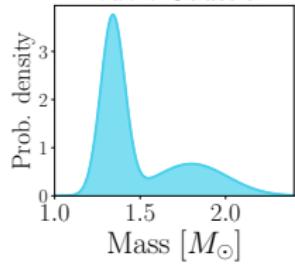
Uniform



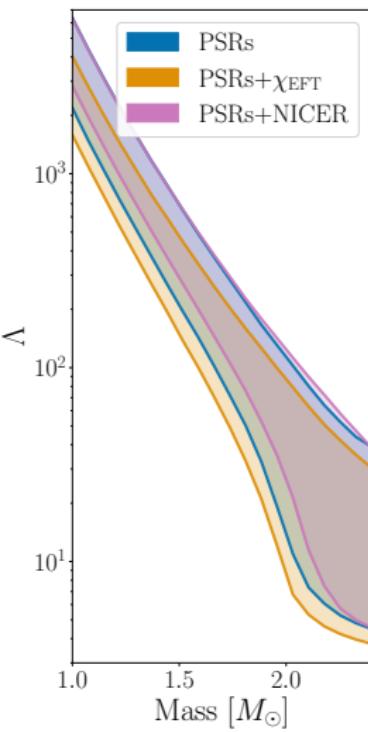
Gaussian



Double Gaussian

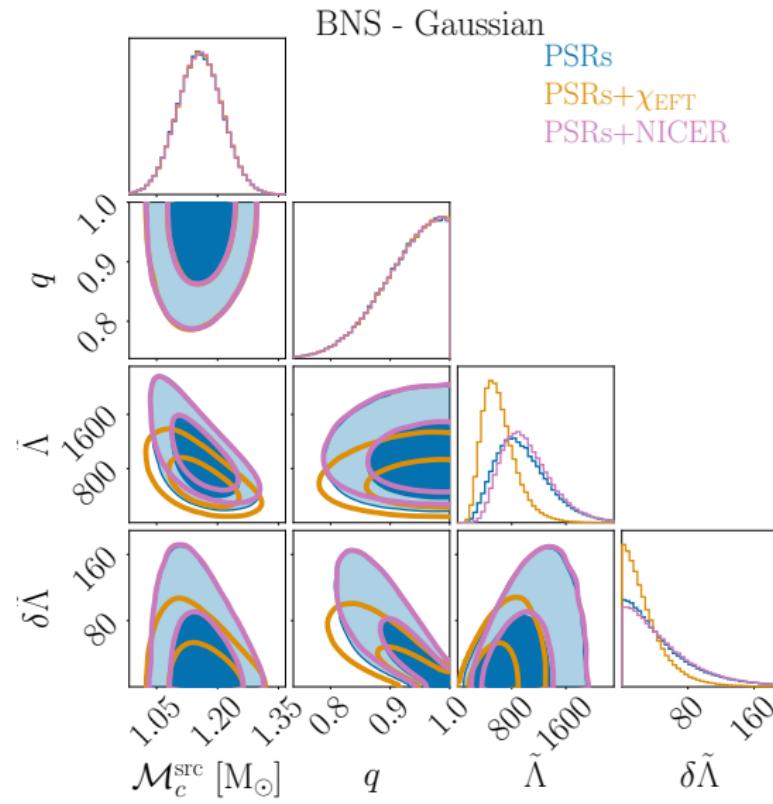


EOS

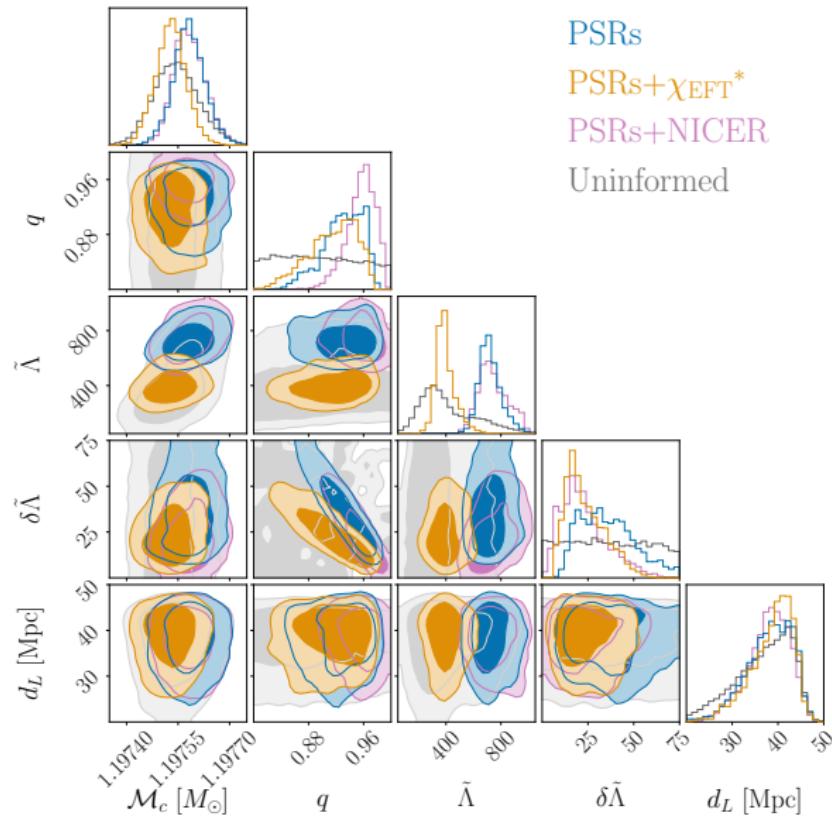


$$p_{\text{NF}}(m_1^{\text{src}}, m_2^{\text{src}}, \Lambda_1, \Lambda_2 | \mathcal{H})$$

# Example prior



# GW170817 posteriors with EOS informed priors



# Closing thoughts

- **Neural priors:** encode non-trivial prior information (including uncertainties)
- Implemented in BILBY, but works with any stochastic sampler
  - Appealing for GPU-based samplers (FLOWMC, BLACKJAX)
- Bayesian source classification:
  - GW170817: BNS with soft EOS
  - GW230529: NSBH favored over BNS
- ML for ‘informed’ sampling (cf. Michael Williams’s talk)
- Code is open source
  - [Θ ThibeauWouters/neural-priors](https://github.com/ThibeauWouters/neural-priors)
  - [Θ ThibeauWouters/bilby/tree/neural\\_prior\\_bilby\\_pipe](https://github.com/ThibeauWouters/bilby/tree/neural_prior_bilby_pipe)



**Thanks for listening!**

