

Projections for Neutron Star EOS Constraints with the LVK Network in A+ Era

Alexis Boudon - IP2I Lyon team
With: Hong Qi, Jean-François Coupechoux, Philippe Landry, and Viola Sordini







Canadian Institute for Theoretical Astrophysics

L'institut Canadien d'astrophysique théoriqu

Overview- NS EOS

Current constraints

Abbott et al. 2017 / 2018 / 2019

- **GW170817:** First BNS merger with GWs provided key constraints on the NS EOS through Λ measurements
- Multi-Messenger Astronomy: Combining data from observations of pulsars, nuclear experiments, and theoretical predictions has refined EOS constraints

Capano et al. 2020, Legred et al. 2021, Huth et al. 2022

Limitations

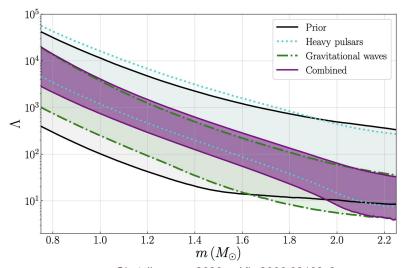
Low Detection Rate: Only 2 BNS mergers Abbott et al. 2017 / 2020 (GW170817 and GW190425)

Future Observing Prospects

Abbott et al. 2018

- O5 Observing Run: Upgraded LVK network is expected to increase the detection range for BNS mergers
- **Next-Generation:** Precision constraints on NS matter will likely require future observatories (ET, CE, ...)

Maggiore et al. 2020, Evans et al. 2021



Chatziioannou 2020, arXiv:2006.03168v2

Aim: Constraining NS EOS with LVK Network in A+

Objectives

- Assess how well we can constrain NS EOS recovery using simulated BNS mergers detectable by LVK Network
- Compare different EOS models: soft (hqc18), intermediate (sly230a), and stiff slope (mpa1), under "realistic" observation scenarios

 Baym et al. 2018

 Chabanat et al. 1997, Danielewicz and Lee 2009, Gulminelli and Raduta 2015

Methodology

- Simulate BNS Populations: Inject large number of BNS events (1,000 events per EOS) into A+ sensitivity
- Parameter Estimation: Use ROQs for rapid GW signal analysis

Canizares et al. 2015, Smith et al. 2016

Constrain EOS: Apply Bayesian inference (LWP) to extract EOS constraints

EOS Inference Approach (LWP)

- Post-process Parameter Estimation: Use spectral EOS inference with large prior sets
- **Sequential Inference**: Combine EOS information across events, using posteriors from one event as priors for the next

Determine how well the injected EOSs are recovered in all scenarios, based on the number of detected BNS events

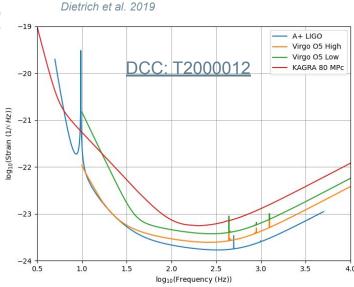
Methodology Overview: Simulations of GWs

Injection Setup

Landry and Read 2021, Abbott et al. 2023

- **BNS Population Model:** Uniform NS mass distribution with random pairing, low spin scenario and sources up to 460 MPc (comoving volume)
- Waveform: Use IMRPhenomPv2_NRTidalv2 model, including both spin precession and tidal effects.

Parameter (Symbol) [Unit]	Injection configuration	Parameter estimation prior
Source-frame primary NS (m_1) $[M_{\odot}]$	Uniform $[1, m_{\text{max}}^*]$ with $m_1 \ge m_2$	Determined by \mathcal{M}_{c} and q
Source-frame secondary NS (m_2) $[M_{\odot}]$	Uniform $[1, m_{\text{max}}]$	Determined by \mathcal{M}_c and q
Source-frame chirp mass (\mathcal{M}_c) $[M_{\odot}]$	$\frac{(m_1 m_2)^{3/5}}{(m_1 + m_2)^{1/5}}$	Uniform [1.6, 2.6] M_{\odot} for 128 s
		Uniform [0.98, 1.7] M_{\odot} for 256 s
Source-frame mass ratio (q)	m_2/m_1	Uniform $[0.125, 1]$
Dimensionless primary NS spin (a_1)	Uniform $[0, 0.05]$	Uniform $[0, 0.05]$
Dimensionless secondary NS spin (a_2)	Uniform $[0, 0.05]$	Uniform $[0, 0.05]$
Primary NS tilt (θ_1) [radian]	Uniform Sine $[0, \pi]$	Same as injection
Secondary NS tilt (θ_2) [radian]	Uniform Sine $[0, \pi]$	Same as injection
Relative spin azimuthal angle (ϕ_{jl}) [radian]	Uniform $[0, 2\pi]$	Same as injection
Spin phase angle (ϕ_{12}) [radian]	Uniform $[0, 2\pi]$	Same as injection
Luminosity distance (d_L) [Mpc]	Uniform in square [10, 460] Mpc	Square power law [1, 1000] Mpc
Right ascension (α) [radian]	Uniform $[0, 2\pi]$	Uniform $[0, 2\pi]$
Declination (δ) [radian]	Uniform Cosine $[-\pi/2, \pi/2]$	Uniform Cosine
Inclination angle $(\theta_{\rm JN})$ [radian]	Uniform Sine $[0, \pi]$	Uniform Sine $[0, \pi]$
Polarization (Ψ) [radian]	Uniform $[0,\pi]$	Uniform $[0, \pi]$
Coalescence phase (ϕ) [radian]	Uniform $[0, 2\pi]$	Marginalized
Geocenter time (t_c) [second]	Trigger time	Uniform [trigger time - 0.1, trigger time + 0.1]
Tidal deformability of primary NS (Λ_1)	Determined by m_1 and EOS	Uniform [0, 5000]
Tidal deformability of secondary NS (Λ_2)	Determined by m_2 and EOS	Uniform [0, 5000]



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Methodology Overview: PE and EOS Inference

Parameter Estimation of GWs from BNS Mergers

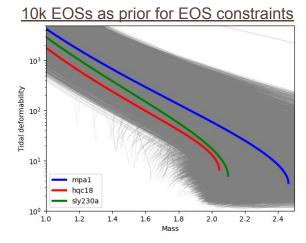
Qi and Raymond 2020, Soichiro et al. 2023, Ashton et al. 2019

- **Technique:** Apply ROQ-accelerated PE using Bilby (1000 events PE completed in 6 days)
- Detection Thresholds: Events detectable with SNR > 11.2 in the LVK network (around 600-700 events for each EOS)
- Recovered Parameters: Full parameters, including mass and tidal deformability distributions

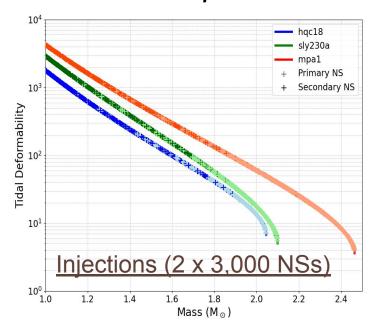
EOS Inference

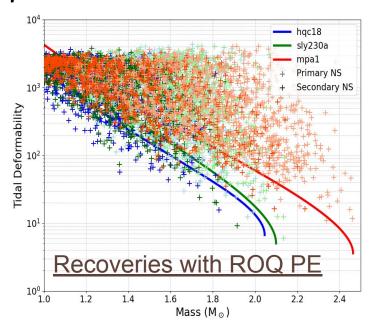
Landry et al. 2020, Legred et al. 2021

- Bayesian Framework: Inference using LWP software (EOS modeled as a Gaussian process based on 10,000 EOS samples)
- Likelihood Calculation: EOS likelihood calculated for each GW event,
- **Expected Results:** Constraints on EOS parameters and NS observables (e.g., NS radius, pressure at saturation density)



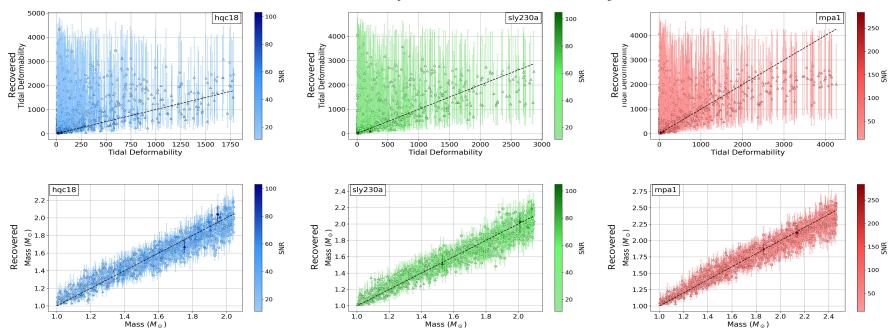
Distribution of Mass and Tidal Deformation





- Injected vs Recovered values: good coverage of the mass-Λ parameter space, posteriors cluster tightly
- After ROQ PE: Distinguishing EOS models is visually difficult

Medians and Error Bars of Recovered Injections

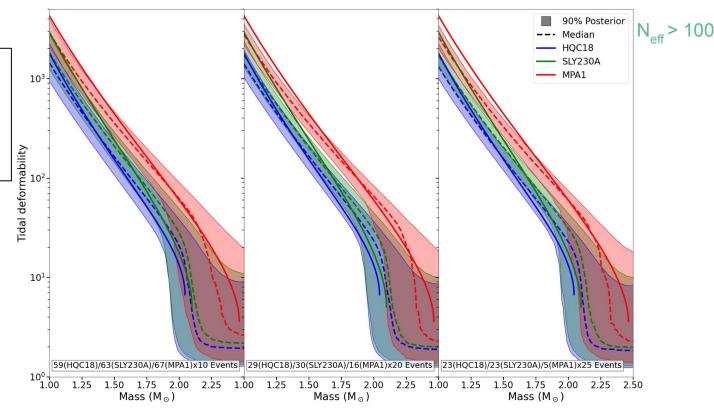


- **Tidal Deformability:** Spread uncertainties, overestimation at low Λ and underestimation at high Λ
- Mass: Injected vs. posterior masses align well for all EOS models, with small errors and good accuracy
- **Impact on EOS Recovery:** Biases in the Λ -mass relationship due to systematic errors in Λ estimates

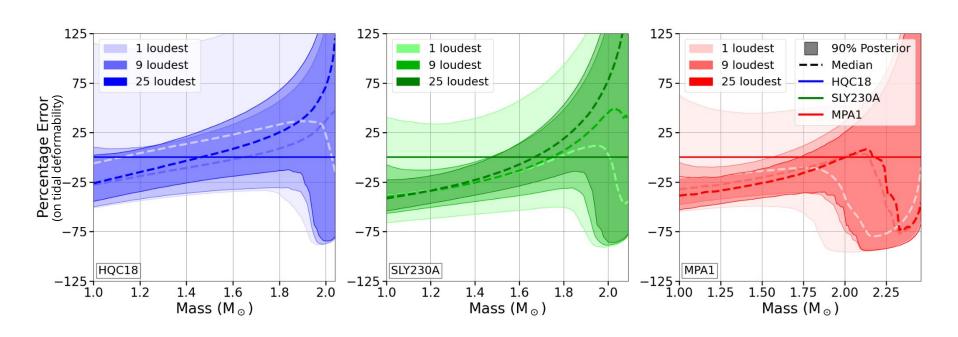
EOS Constraints with Different Numbers of Detections

Combined posteriors for the three EOS for two different size groups (10, 20, 25 events) + averaging

All EOS show biases, even for groups of 10 events. Bias increases with larger group sizes.



Evolution of EOS Constraints: N Loudest Events



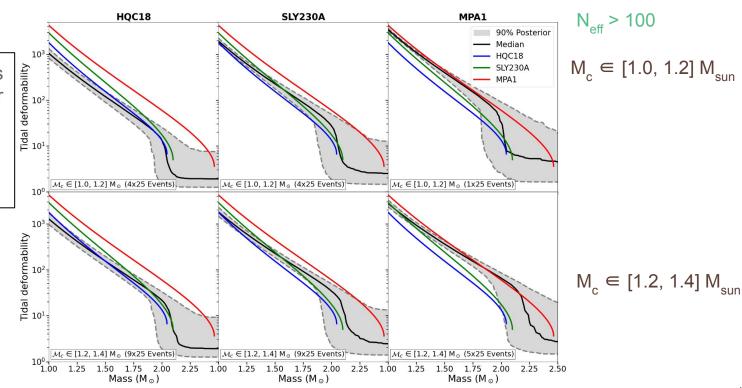
Conclusions

 EOS Recovery with O5 LVK Network: Simulated BNS mergers show promising constraints on NS EOS (can expect roughly 25 BNS detections if the merger rate lies near its median estimate)

- ROQ reduced computation time from months to days, making such population studies possible. The LWP framework allowed us to combine hundreds of events consistently and quantify how these biases propagate to EOS constraints.
- Biases in Tidal Deformability: Systematic underestimation at low masses and overestimation at high masses affects accurate EOS recovery (need to take it into account + might be interesting to find out the reasons)

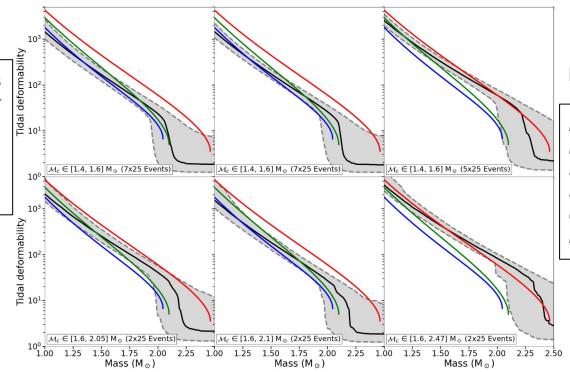
EOS Constraints Varying w.r.t. Chirp Mass Ranges (1/2)

Combined posteriors for the three EOS for different chirp mass ranges considering groups of 25 + averaging



EOS Constraints Varying w.r.t. Chirp Mass Ranges (2/2)

Combined posteriors for the three EOS for different chirp mass ranges considering groups of 25 + averaging



 $N_{\rm eff} > 100$

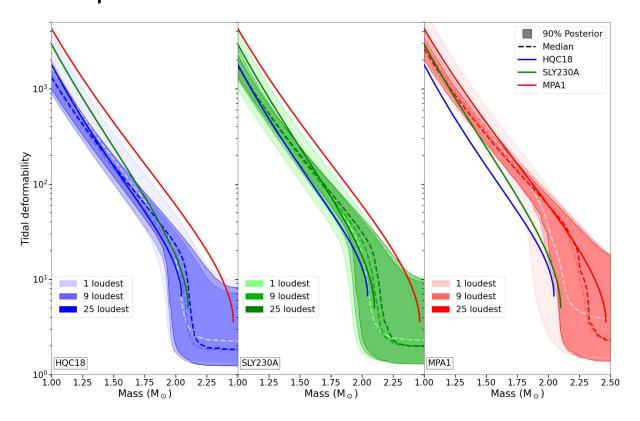
 $M_{c} \in [1.4, 1.6] M_{sun}$

hqc18 is well recovered. sly230a and mpa1 show biases, especially at lower masses.

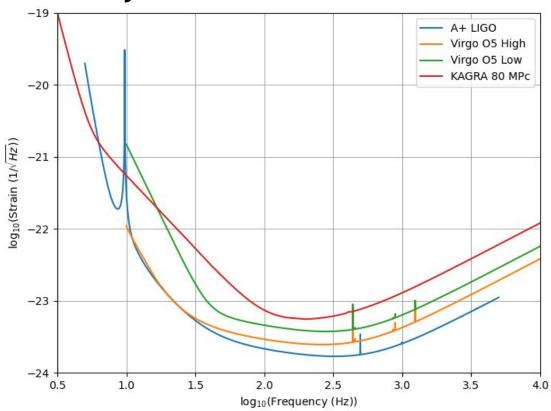
 $M_c \in [1.6, Max] M_{sun}$

- **Tidal Deformability Bias:** Underestimation at low masses and overestimation at high masses
- Method Limitations: Recovery varies by EOS and mass range, better performance at higher M_c

Evolution of EOS Constraints: N Loudest Events



Sensitivity curves used



DCC link to sensitivity curves

DCC: T2000012

- APlusDesign.txt
- avirgo_O5high_NEW.txt
- avirgo_O5low_NEW.txt
- kagra_80MPc.txt