

Prospects and challenges for dense matter studies with gravitational waves

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ACFI workshop – 13 Oct 2022



Neutron stars in LVK O3

BNS rate of $\sim 440 \text{ Gpc}^{-3} \text{ yr}^{-1}$

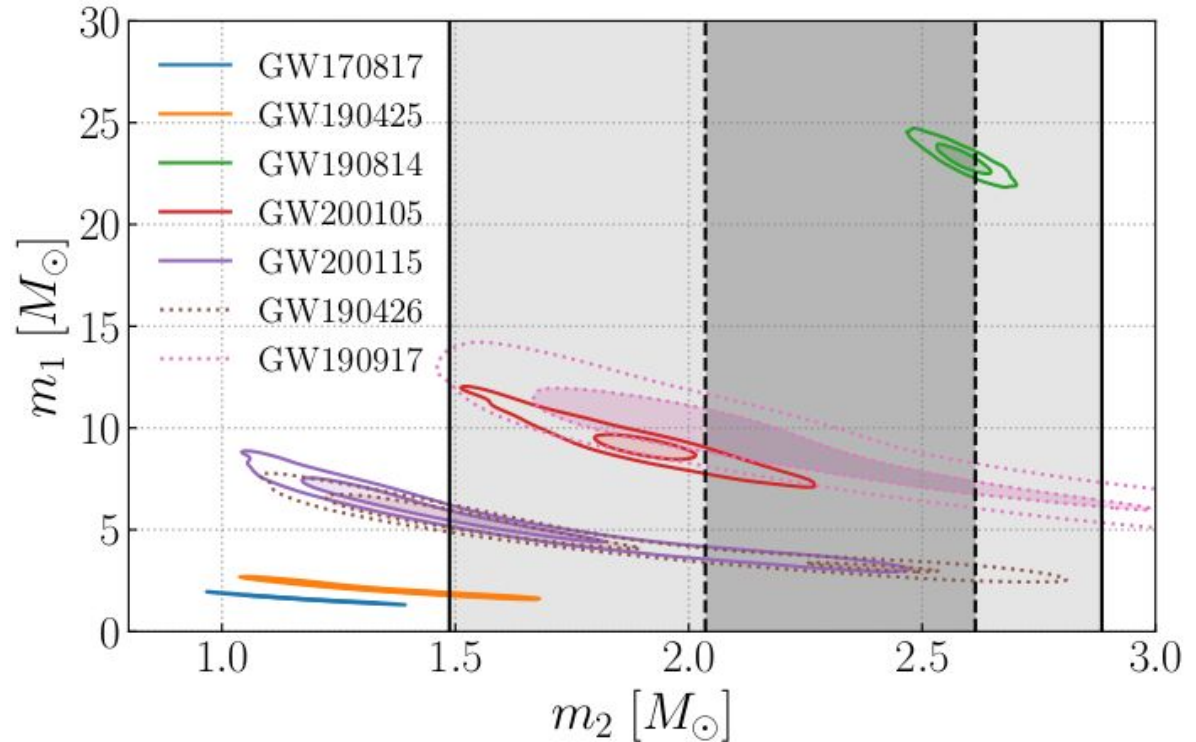
NSBH rate of $\sim 60 \text{ Gpc}^{-3} \text{ yr}^{-1}$

*cf. post-GW170817 BNS rate
estimate of $\sim 3800 \text{ Gpc}^{-3} \text{ yr}^{-1}$*

or BBH rate of $\sim 35 \text{ Gpc}^{-3} \text{ yr}^{-1}$

LVK LRR 2020

LVK arXiv:2111.03634



What to expect in LVK O4?

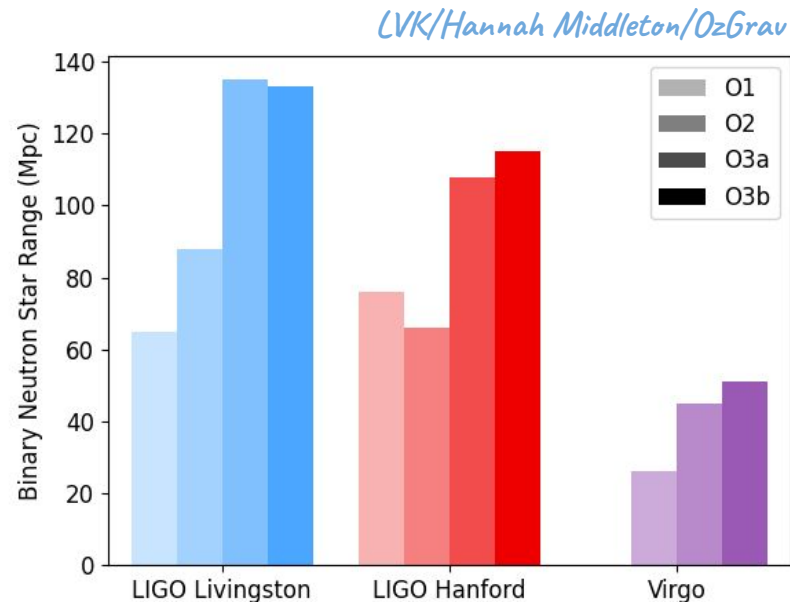
BNS range of ~ 190 Mpc, search VT of ~ 0.016 Gpc³ yr *LVK LRR 2020*

- *~ 7 expected BNS detections*

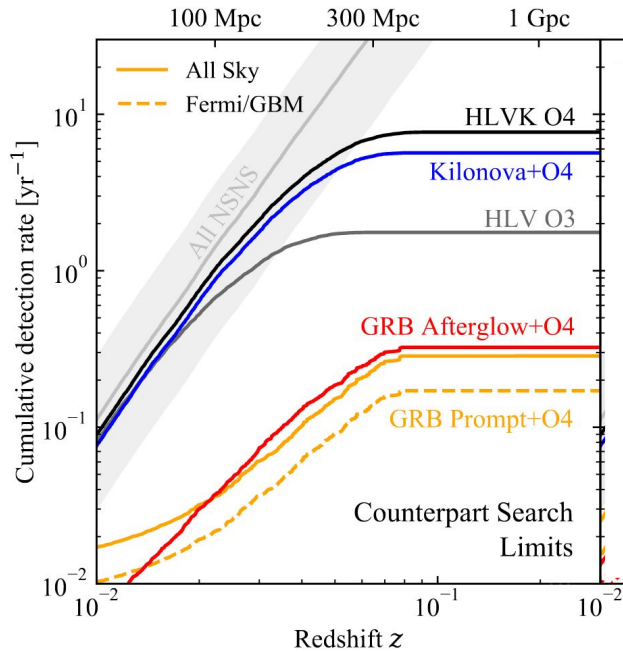
NSBH range of ~ 330 Mpc, ~ 5x larger search VT but ~ 7x lower rate than BNSs

- *~ 5 expected NSBH detections*

LVK O5 (ca. 2025) is projected to increase search VTs by another ~ 5x



Multimessenger forecasting

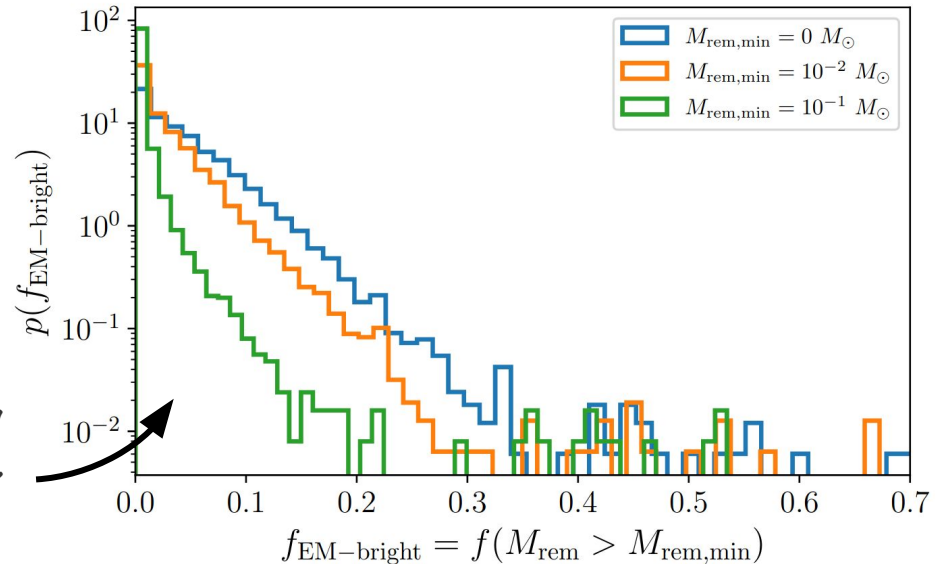


80% of BNS mergers estimated to produce a kilonova, but only ~ 10% will be detectable

Colombo+ Ap] 2022, cf. Patricelli+ MNRAS 2022

reasonable to expect 1 EM counterpart in O4

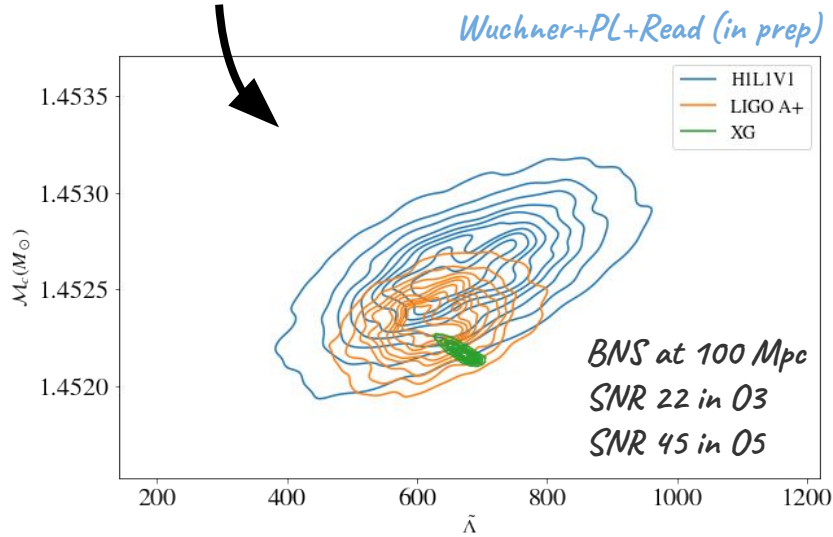
Biscoveanu+PL+Vitale arXiv:2207.01568



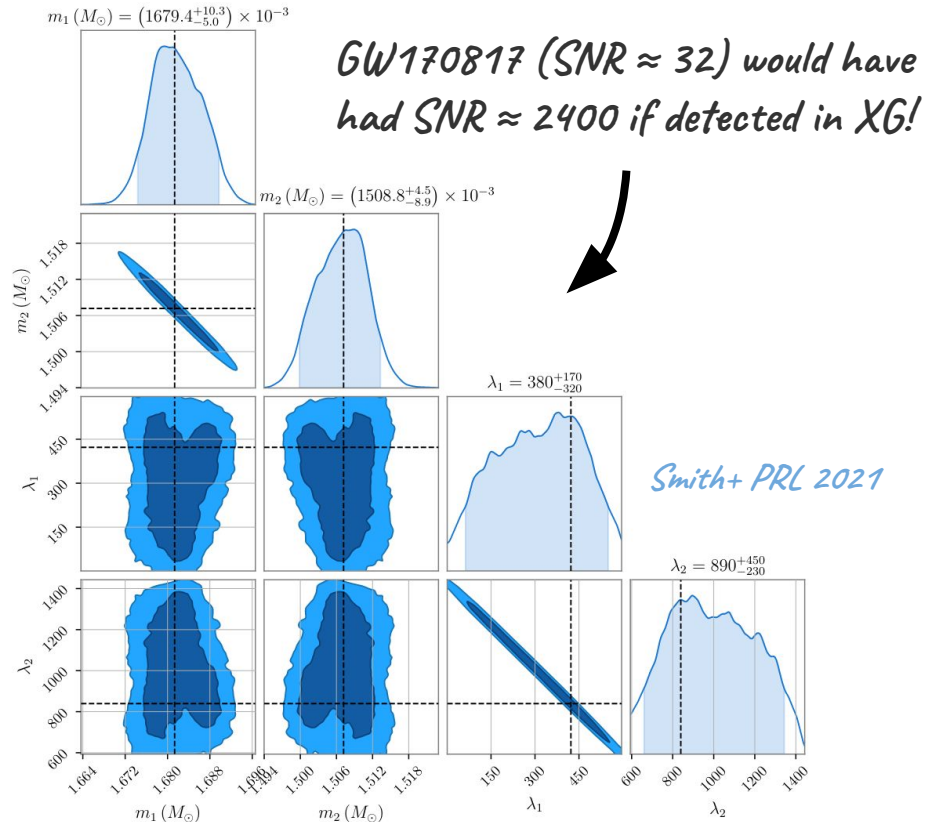
90% credible upper bound on fraction of NSBH population with any ejecta whatsoever is ~ 10%

Future BNS tidal measurements

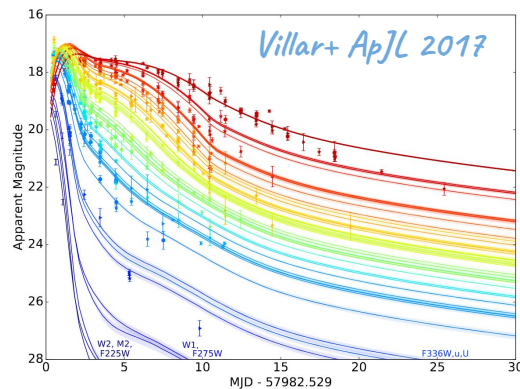
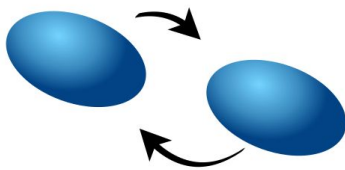
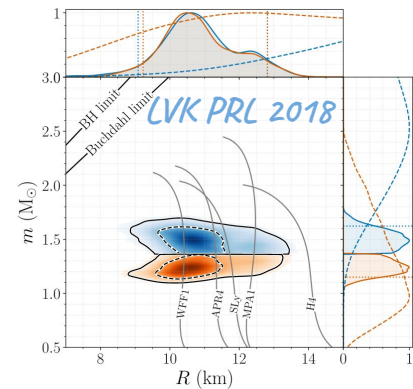
statistical error of ± 100 in binary tidal deformability in O5



SPN tidal parameter $\tilde{\Lambda}$: measurable
 6PN tidal parameter $\delta\tilde{\Lambda}$: not measurable!



Inferring dense matter properties from GWs



hierarchical Bayesian EOS inference from GW observations



$$P(\text{eos} | d) \propto P(\text{eos}) \prod_i \int P(d_i | m_{1,2}^i, \Lambda_{1,2}^i) P(m_{1,2}^i, \Lambda_{1,2}^i | \text{eos}, \text{pop}) dm_{1,2}^i d\Lambda_{1,2}^i$$

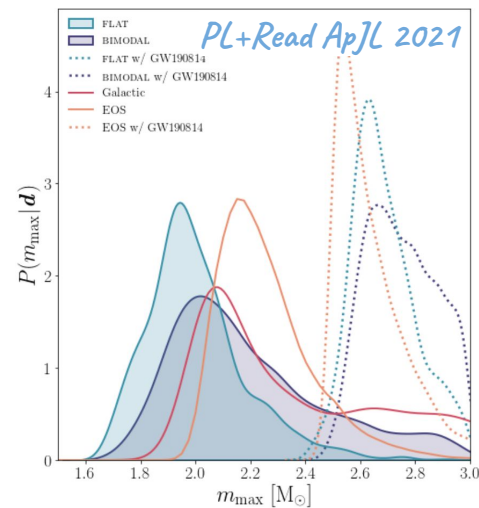
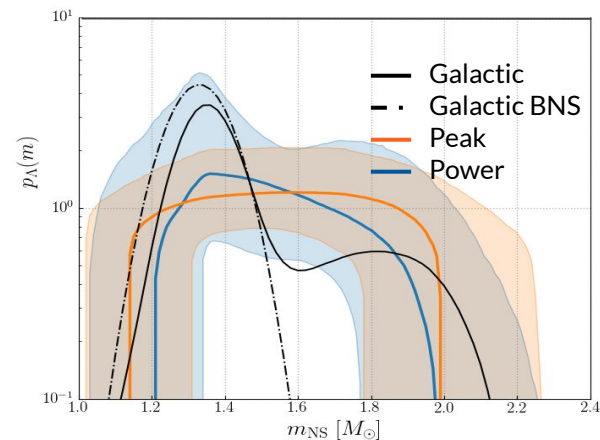
EOS prior

i^{th} GW likelihood

*common prior on
source properties*

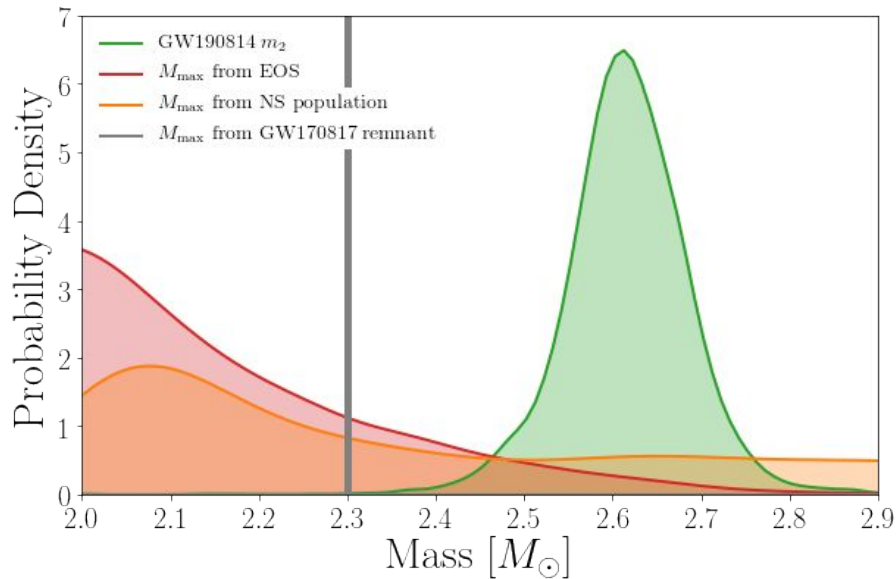
Astrophysical uncertainties

- *shape of the NS mass distribution: uniform or bimodal?*
- *BNS (or NSBH) pairing function: random or preferentially equal-mass?*
- *lower mass gap: disjoint, contiguous, or overlapping NS and BH mass spectra?*
- *maximum mass in the NS population: less than, equal to, or larger than M_{TOV} ?*
- *NS spin distribution: fast or slow?*

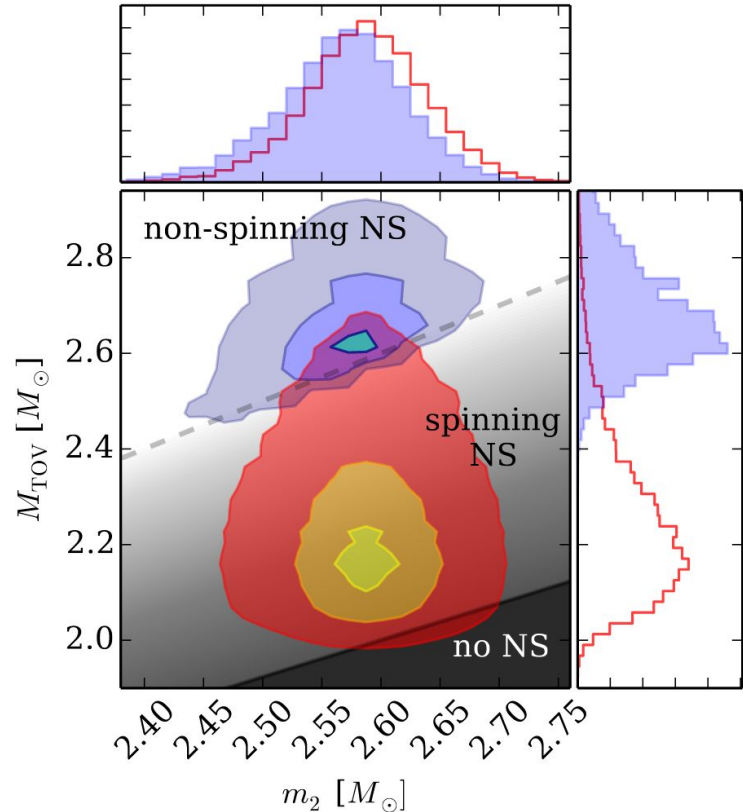


Does the population really matter?

population assumptions inform the classification of GW190814 as a NSBH or a BBH merger, and strongly impact the inferred maximum TOV mass



Essick+PL ApJ 2020



Population assumptions in EOS inference

fixed (but uncertain) population model

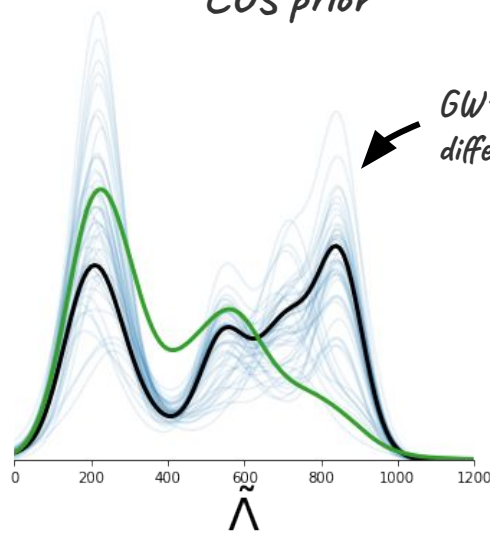


$$P(\text{eos} | d) \propto P(\text{eos}) \prod_i \int P(\text{pop}) P(d_i | m_{1,2}^i, \Lambda_{1,2}^i) P(m_{1,2}^i, \Lambda_{1,2}^i | \text{eos}, \text{pop}) dm_{1,2}^i d\Lambda_{1,2}^i d\text{pop}$$

EOS prior

i^{th} GW likelihood

common prior on source properties



GW170817 tides under different population priors

$$P(m_{1,2}^i, \Lambda_{1,2}^i | \text{eos}, \text{pop}) = P(m_{1,2}^i | \text{pop}) \delta^{(2)}(\Lambda_{1,2}^i - \Lambda(m_{1,2}^i; \text{eos}))$$

mass distribution

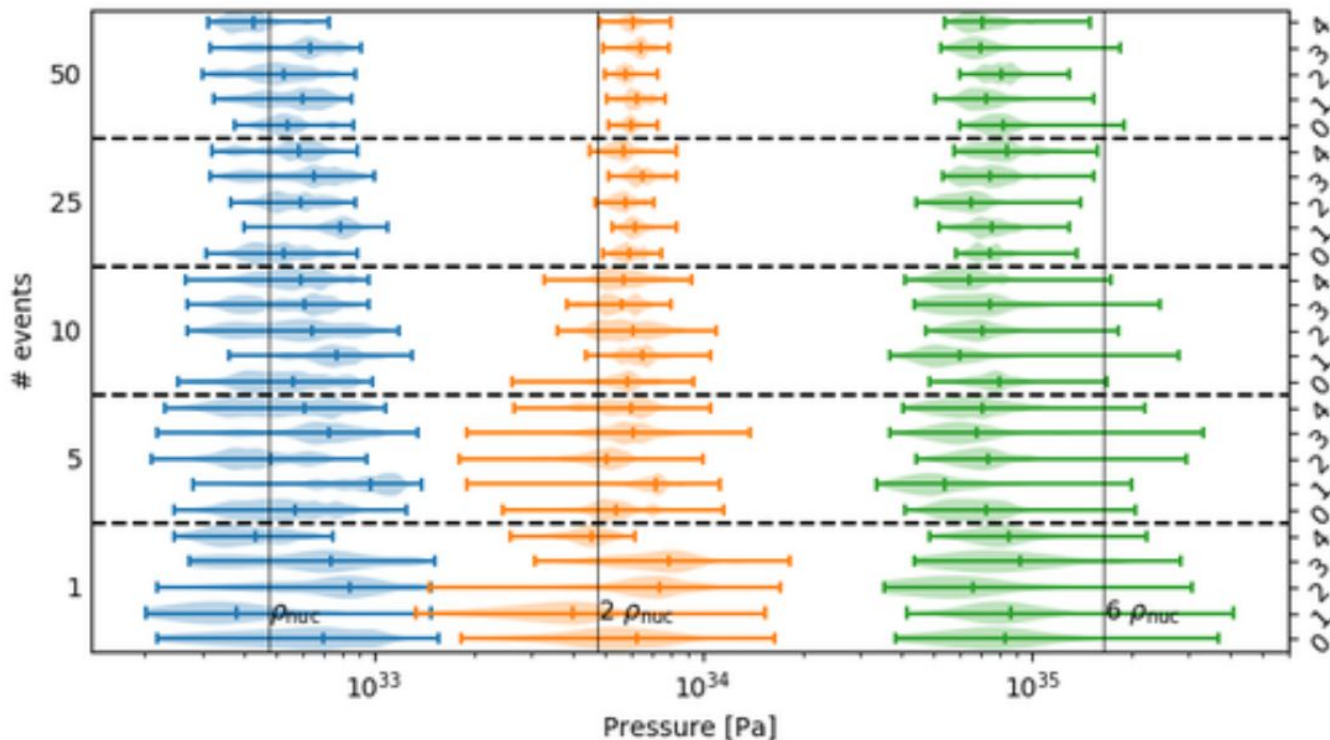
common EOS

Penuliar+PL+Read (in prep)

When does the population matter?

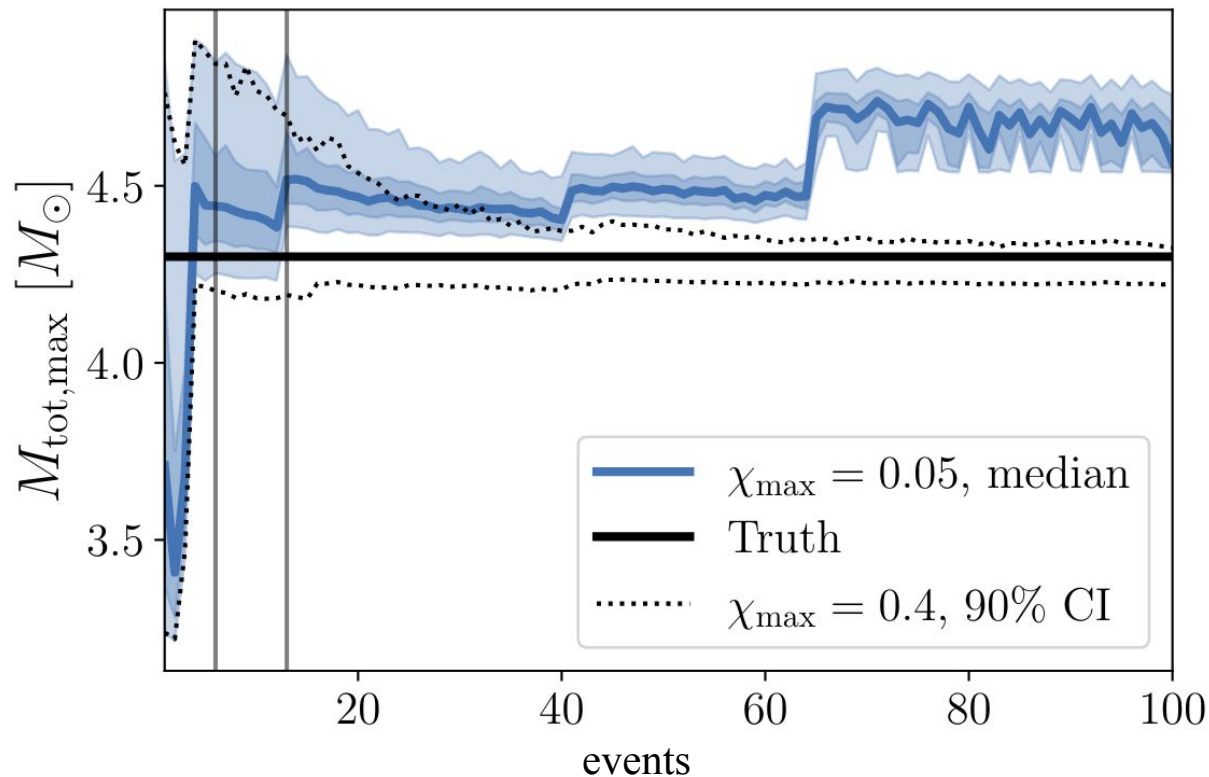
Wysocki+ arXiv:2001.01747

*imposing the wrong
population-level
mass prior can bias
the inferred EOS
after $O(10)$ BNS
observations*



When does the population matter?

Biscoveanu+Talbot+Vitale MNRAS 2022



*incorrectly assuming NSs
spin slowly can bias the
inferred maximum mass in
the population after $O(10)$
BNS observations*

Simultaneous EOS & population inference

population model no longer fixed!



$$P(\text{eos, pop} | d) \propto P(\text{eos, pop}) \prod_i \beta(\text{pop})^{-1} \int P(d_i | m_{1,2}^i, \Lambda_{1,2}^i) P(m_{1,2}^i, \Lambda_{1,2}^i | \text{eos, pop}) dm_{1,2}^i d\Lambda_{1,2}^i$$

*joint EOS +
population prior*

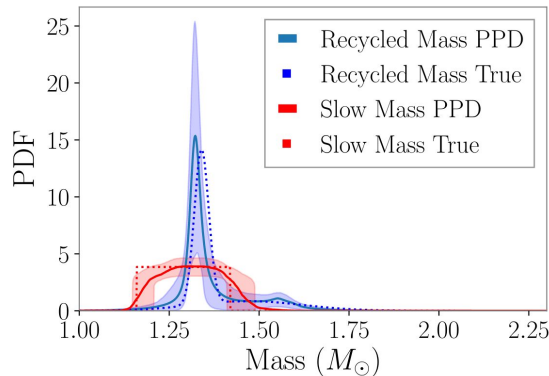
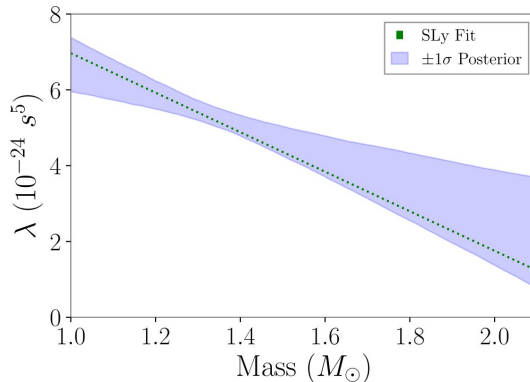
*selection
effects*

i^{th} GW likelihood

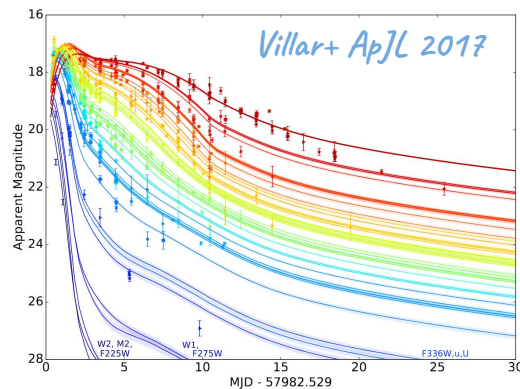
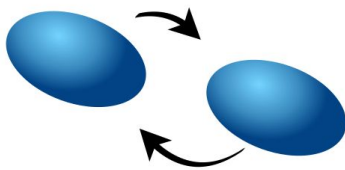
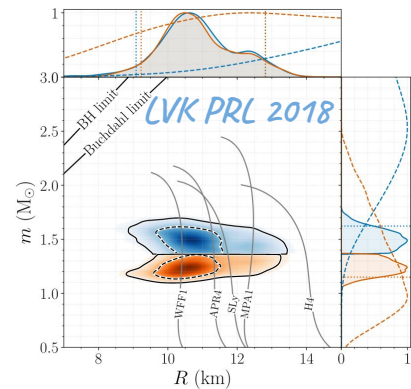
*common prior on
source properties*

*methods in development
by several groups*

*e.g. Golomb+Talbot ApJ 2022,
Wysocki+ arXiv:2001.01747,
Legred+ (incl. PL) (in prep)*



Inferring dense matter properties from GWs



hierarchical Bayesian EOS inference from GW observations

$$P(\text{eos} | d) \propto P(\text{eos}) \prod_i \int P(d_i | m_{1,2}^i, \Lambda_{1,2}^i) P(m_{1,2}^i, \Lambda_{1,2}^i | \text{eos}, \text{pop}) dm_{1,2}^i d\Lambda_{1,2}^i$$

EOS prior

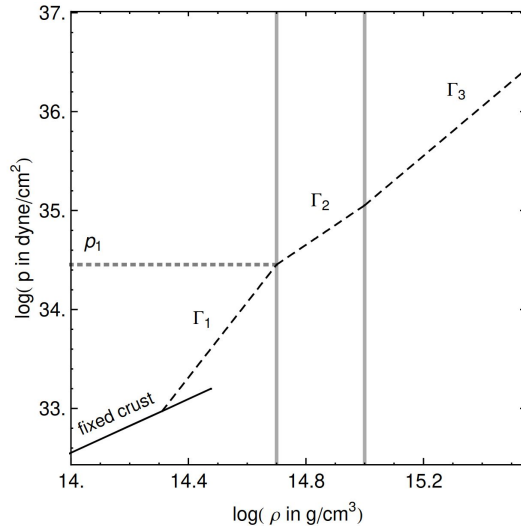
i^{th} GW likelihood

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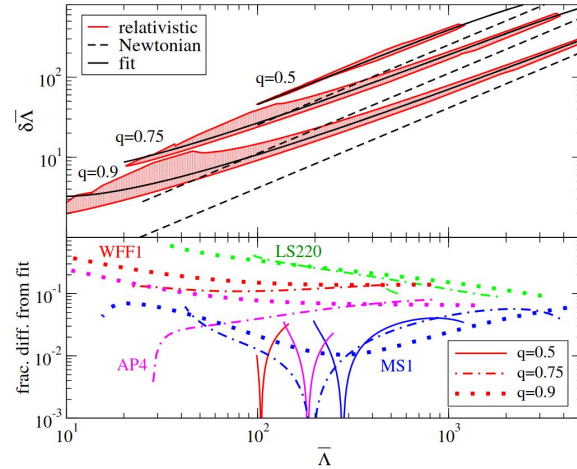
EOS modeling & uncertainty quantification

EOS parameterizations

e.g. piecewise polytropes [Read+ PRD 2008](#),
spectral decomposition [Lindblom PRD 2010](#),
sound speed parameterization [Tews+ PRC 2018](#)



Universal relations



e.g. binary Love relations [Yagi+Yunes CQG 2016](#)

parametric models specify $P(\text{eos})$ via distributions over parameters; URs use fit + residuals to specify effective $P(\bar{\lambda}(m))$

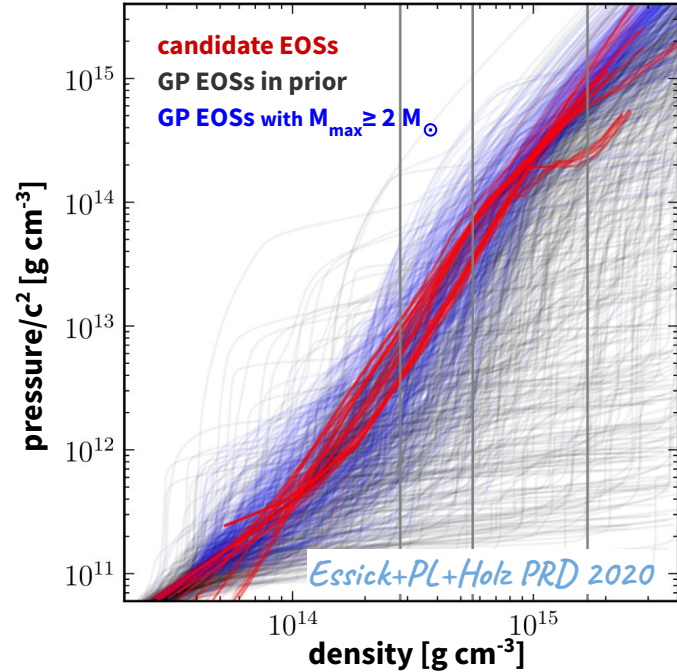
EOS modeling & uncertainty quantification

Nonparametric representations

directly specify the distribution over functions
that is $P(\text{eos})!$

e.g. Gaussian processes [PL+Essick PRD 2019](#),
cf. [Miller+ ApJL 2021](#), also machine learning
models [Morawski+Bejger A&A 2020](#), [Han+ ApJ 2021](#)

the Gaussian process for the EOS is a probability
distribution over causal and thermodynamically stable
functions $c_s^2(p)$ with Gaussian covariance kernel



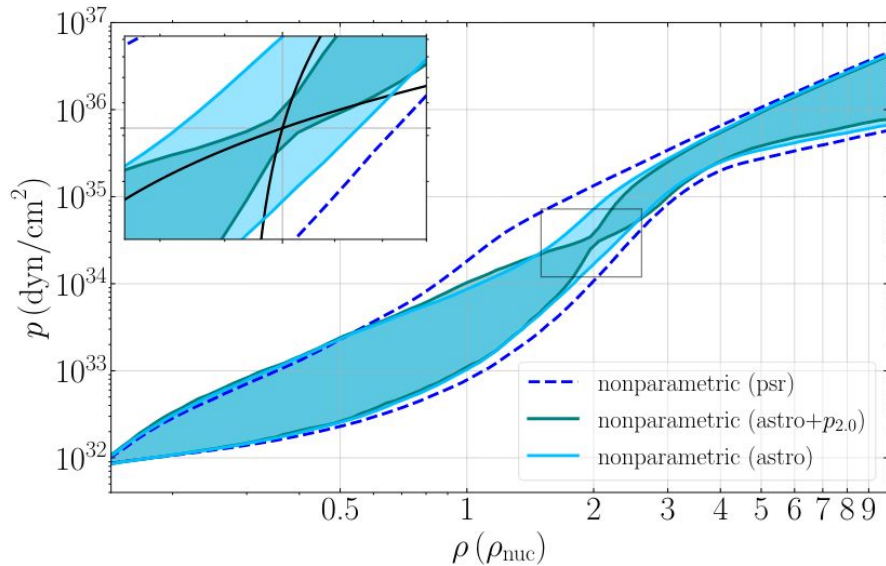
$$K_{\text{se}}(x_i, x_j; \sigma, l) = \sigma^2 \exp\left(-\frac{(x_i - x_j)^2}{2l^2}\right)$$

Designing phenomenological EOS priors

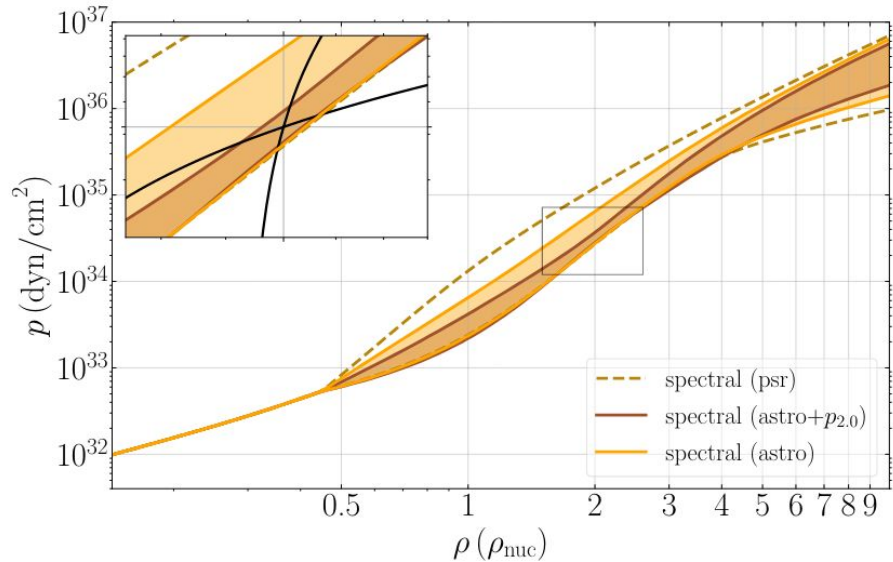
“data-driven” astrophysical EOS inference philosophy: prescribe prior spanning all physically allowed EOSs

↪ beware: parametric EOS models can introduce artificial intra-density correlations [Legred+ \(incl. PL\) PRD 2022](#)

nonparametric



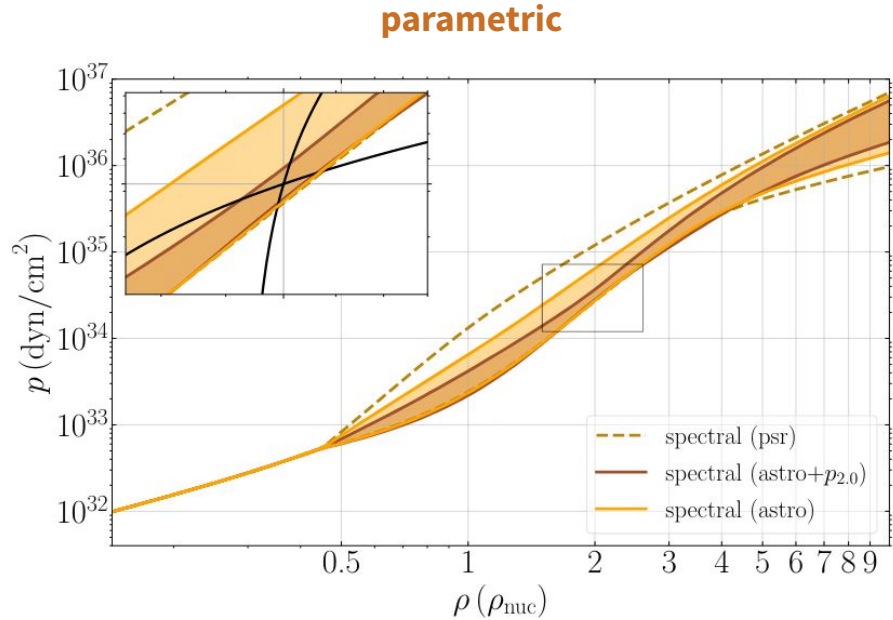
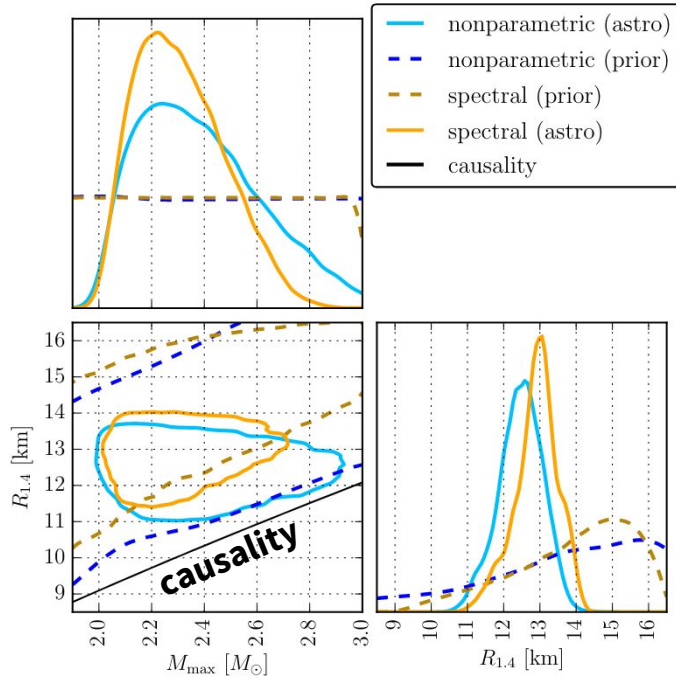
parametric



Designing phenomenological EOS priors

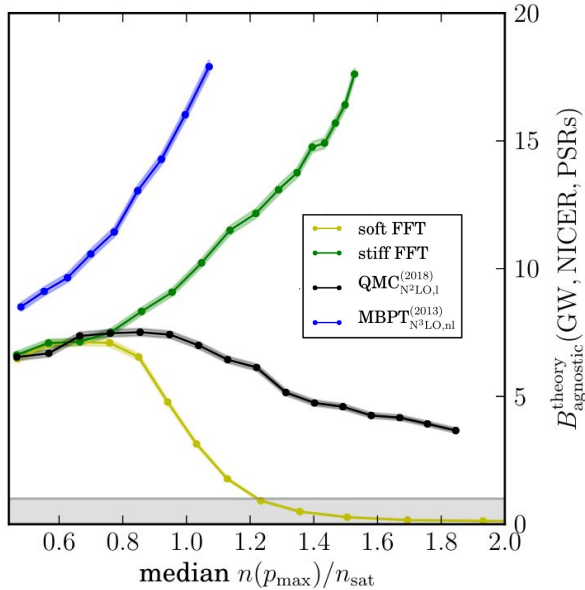
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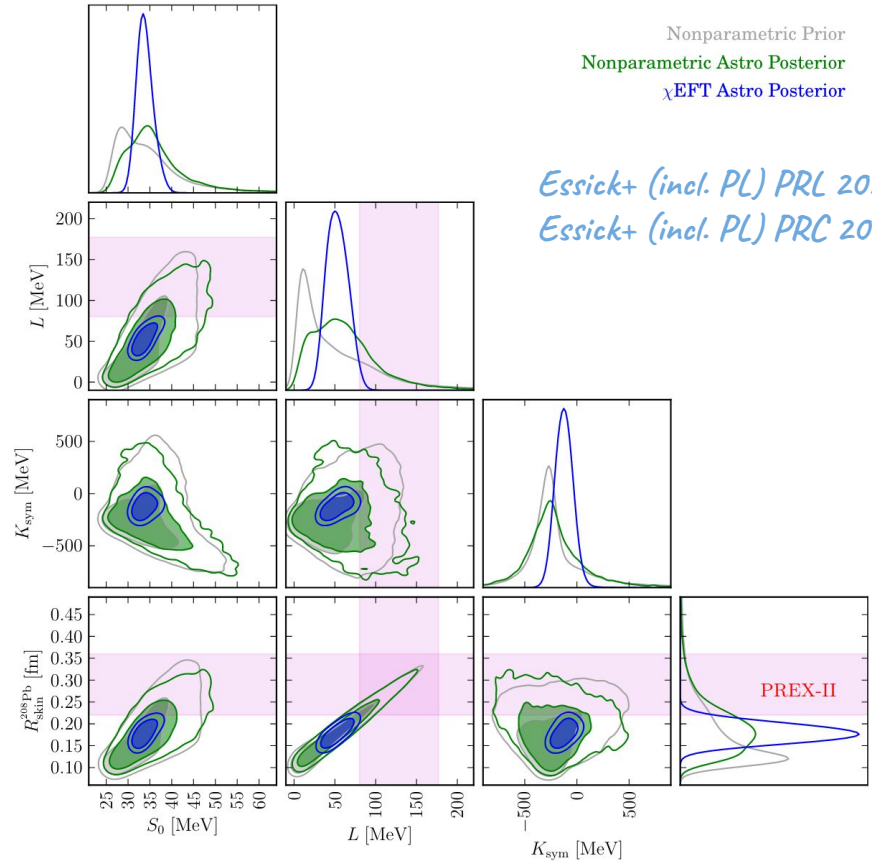


Beyond EOS phenomenology

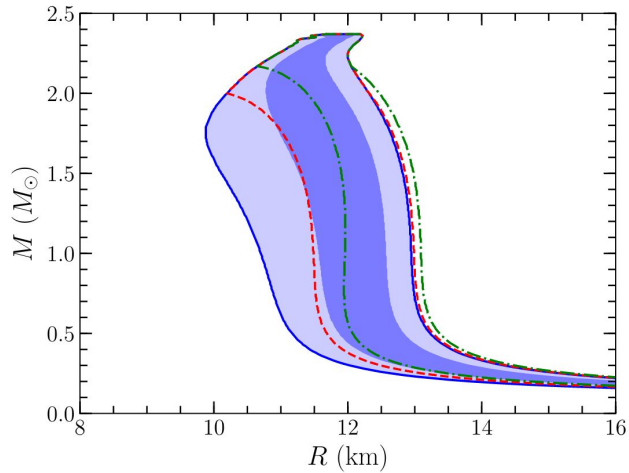
condition a phenomenological EOS prior, like the Gaussian process model, on chiral EFT calculations at low densities



*Essick+ (incl PL)
PRC 2020*

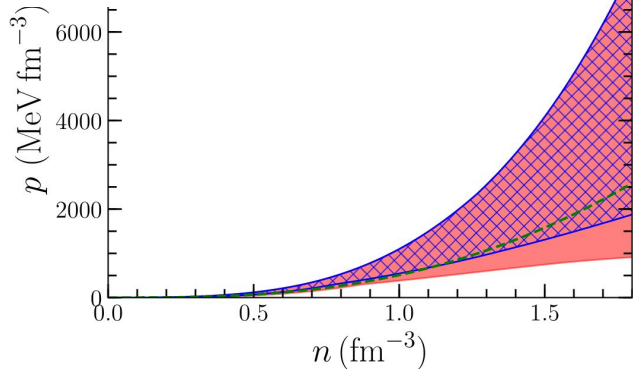


Beyond EOS phenomenology



*parametric energy density
functional model for EOS,
constrained with chiral EFT
and nuclear experiments*

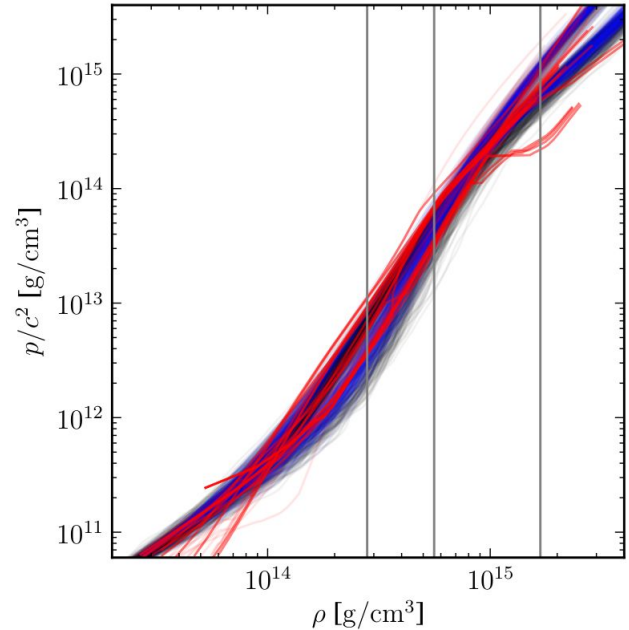
Lim+Holt EPJA 2019



*Gaussian processes
conditioned on
candidate EOSs with
different
compositions*

Essick+PL+Holz PRD 2020

$P(\text{Hadronic} \text{data})$	$P(\text{Hyperonic} \text{data})$	$P(\text{Quark} \text{data})$
28%	16%	56%



Outlook

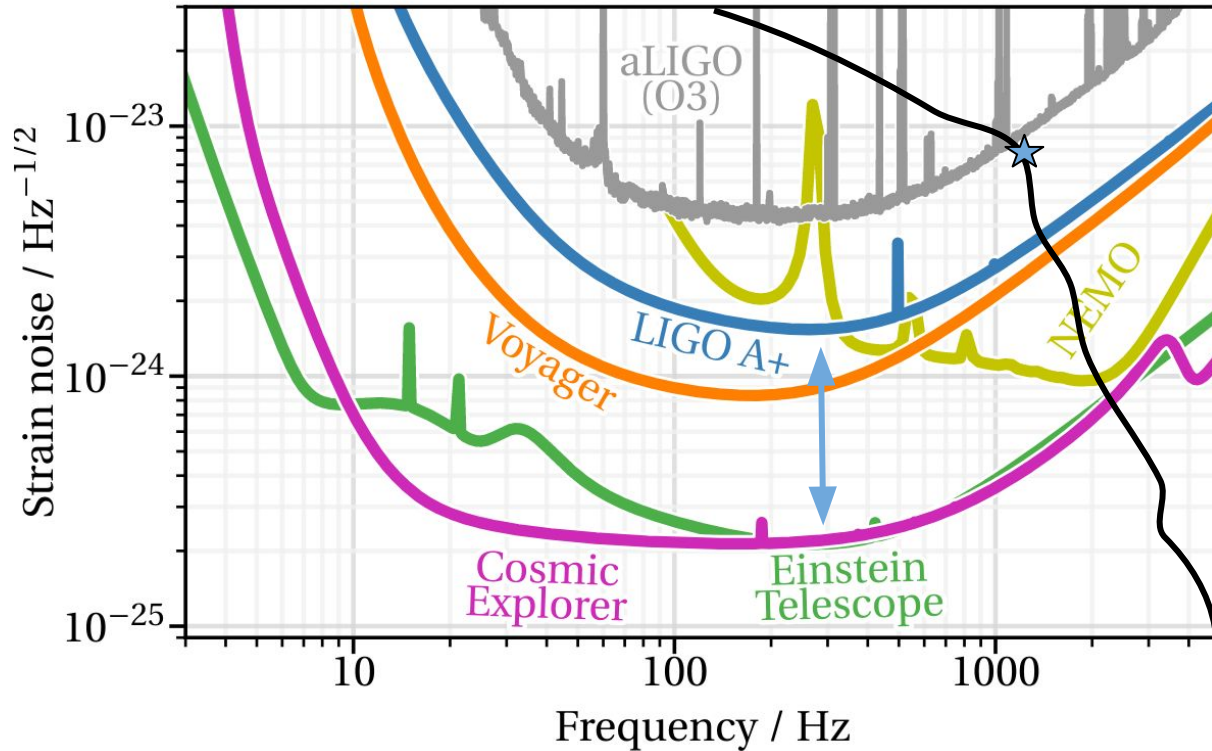
In the next campaign of LVK observations, we can hope for...

- *several BNS mergers with measurable tidal effects*
- *half a dozen more NSBH mergers*
- *maybe an EM counterpart or two*

In order to take full advantage of the dense-matter information these louder, more numerous signals will provide, we need...

- *more careful treatments of systematics in EOS modeling*
- *closer integration between nuclear theory/experiment and astrophysics*
- *simultaneous population and EOS inference*

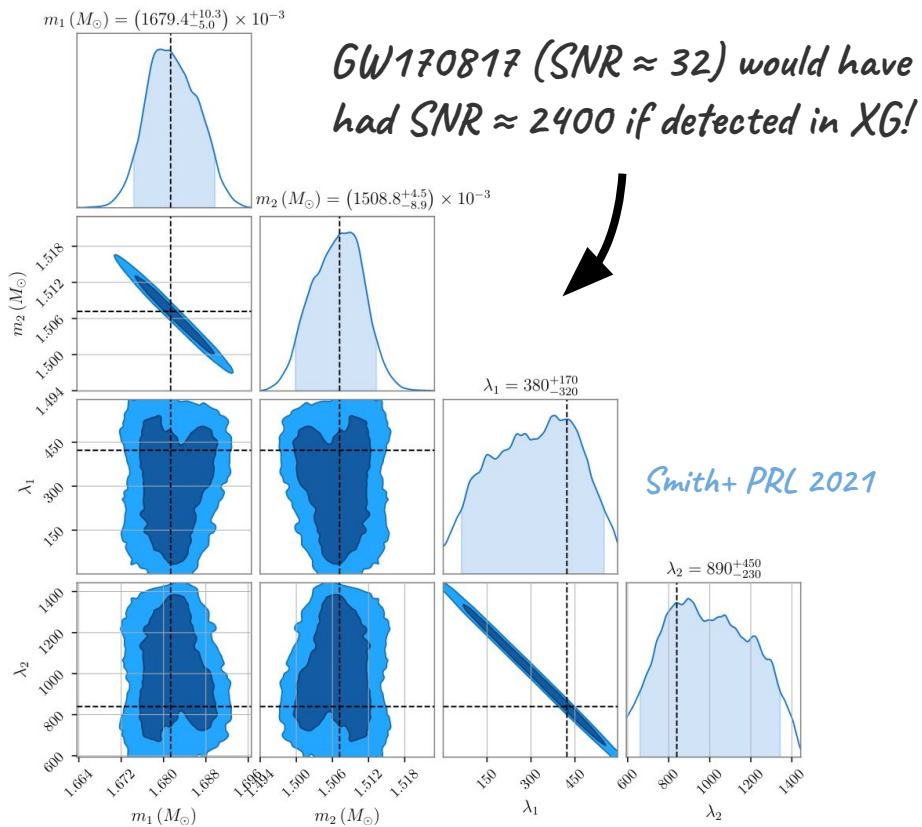
Bonus slide: XG sensitivity



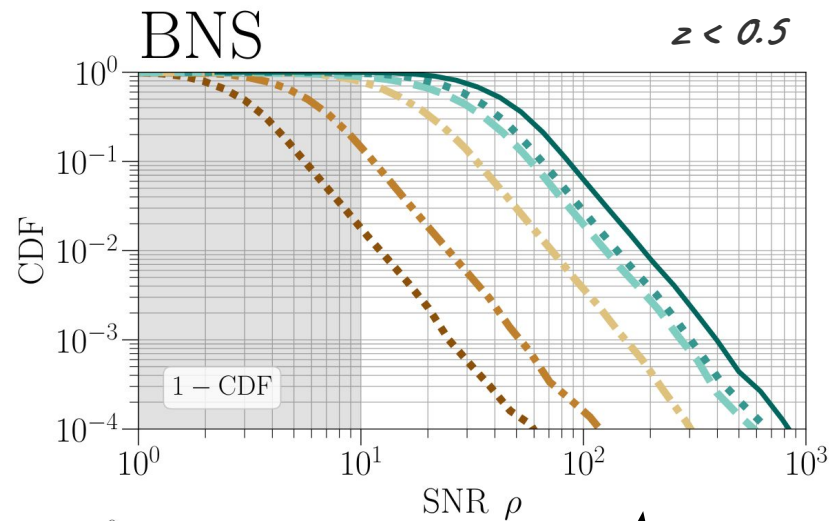
Evans+ arXiv:2109.09882

*XG gives 10x improvement in
broadband sensitivity*

Bonus slide: XG BNS observations

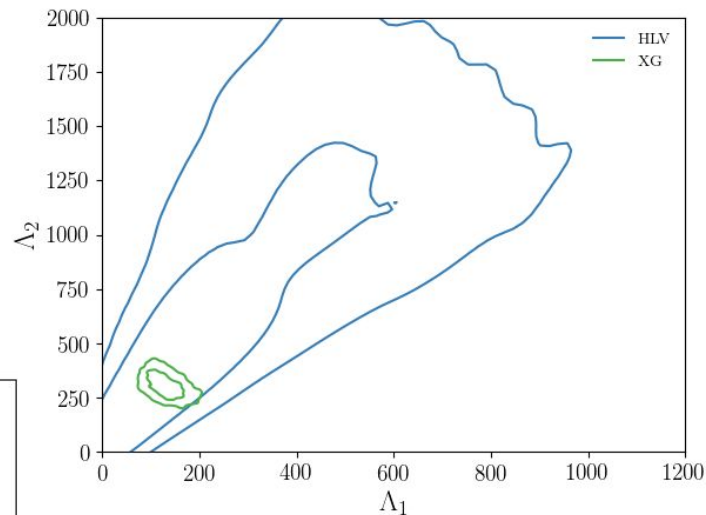
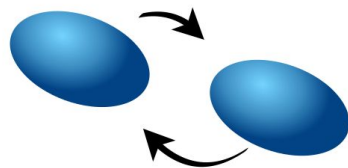
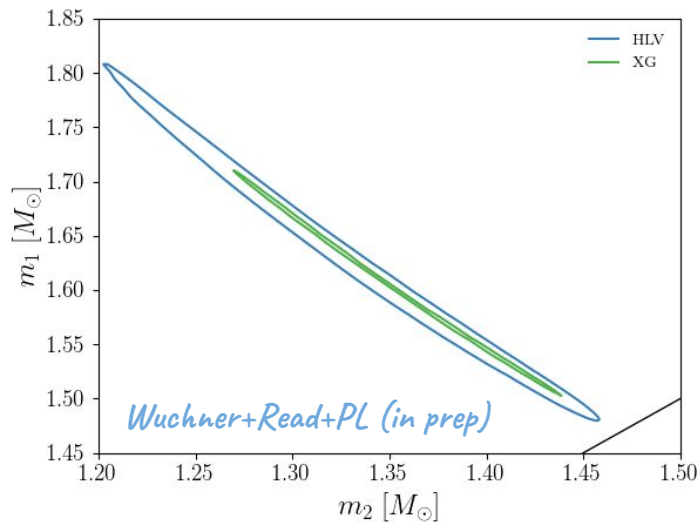


Borhanian+Sathyaprakash arXiv:2202.11048

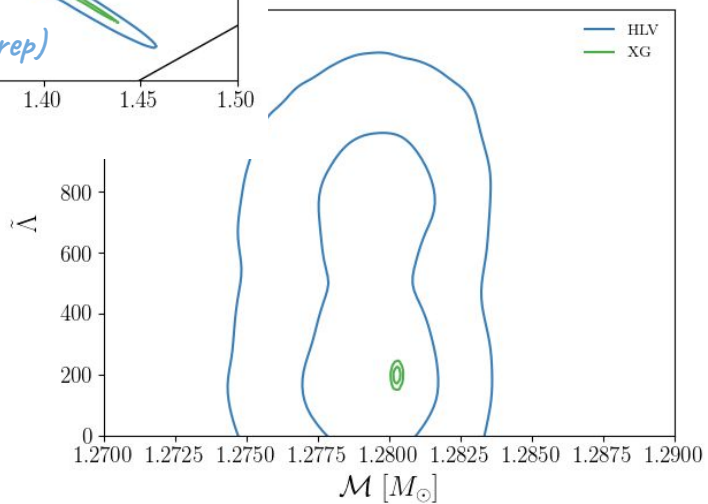


50% of nearby BNSs have SNR > 100 in XG

Bonus slide: XG tidal measurements

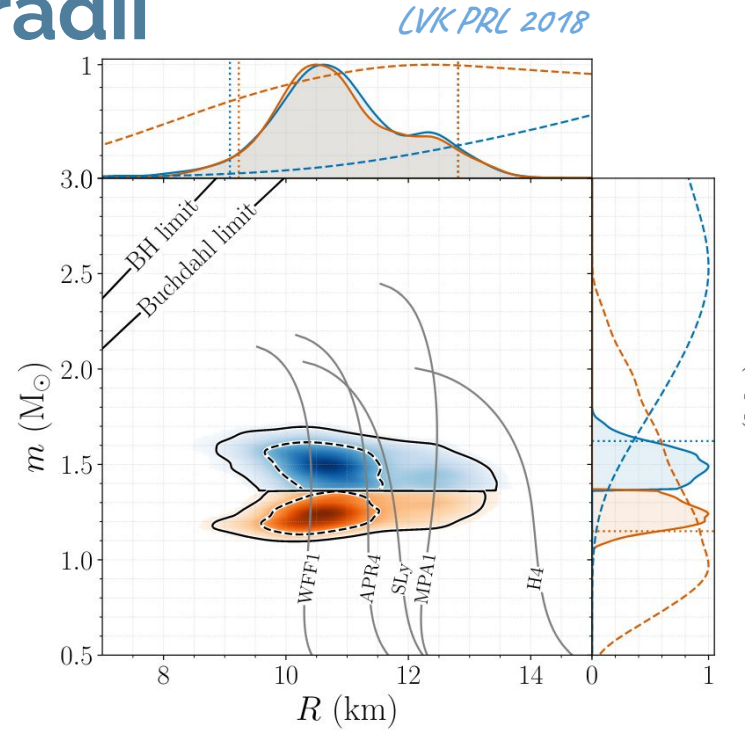
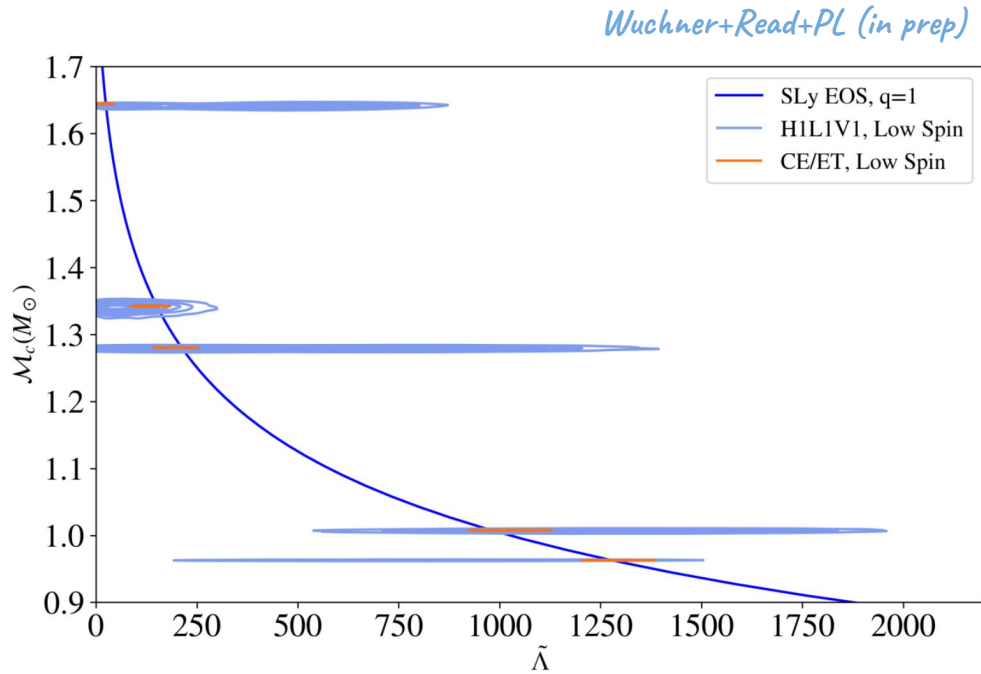


*2.9 M_\odot total mass
BNS at 110 Mpc:
SNR 15 in HLV,
SNR 400 in XG*



*Λ measures tidal deformation;
 $\tilde{\Lambda}$ is a mass-weighted average
of Λ_1 and Λ_2*

Bonus slide: XG NS tides & radii



correlations currently used to map between tides and radii break down in XG era!

