

The Sins of the Priors: Inverse Problems and Statistical Estimation

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Plan of Talk

- To be provocative: why we shouldn't talk about inverse problems
- How much do priors matter in our inferences about the properties of dense matter?

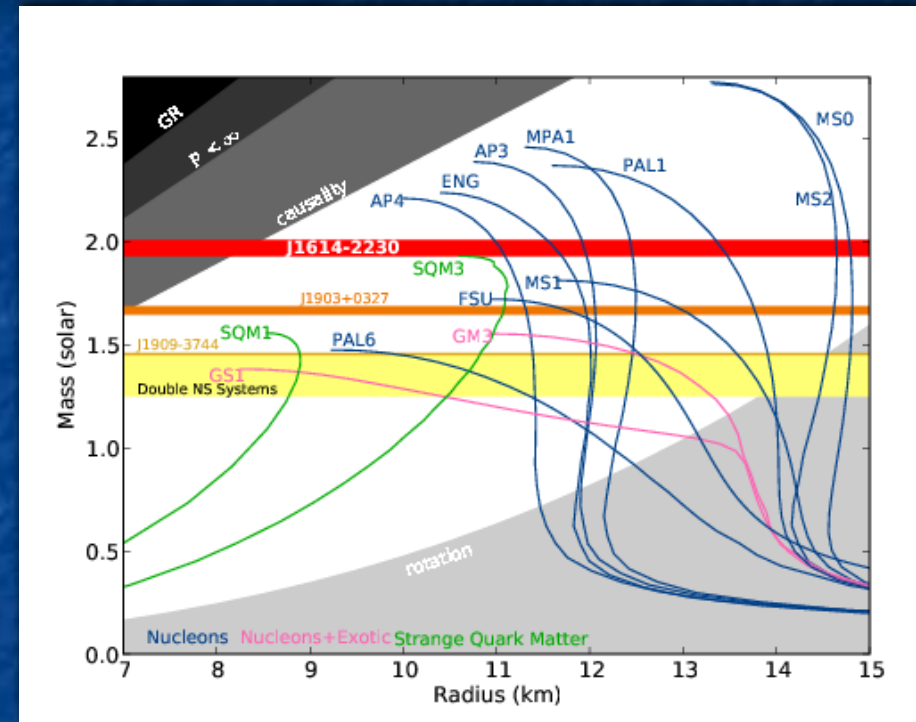
Dense Matter to an Astronomer

- $T=0$ ($10^9 \text{ K} \ll 10^{12} \text{ K}$)
- Equilibrium
- Equation of state, not composition
- Thus $P=P(\rho)$ (or $P(n)$, ...)
- Example: polytrope

$$P=K\rho^\Gamma$$

Piecewise polytrope

- Or smooth slope distribution, e.g., spectral



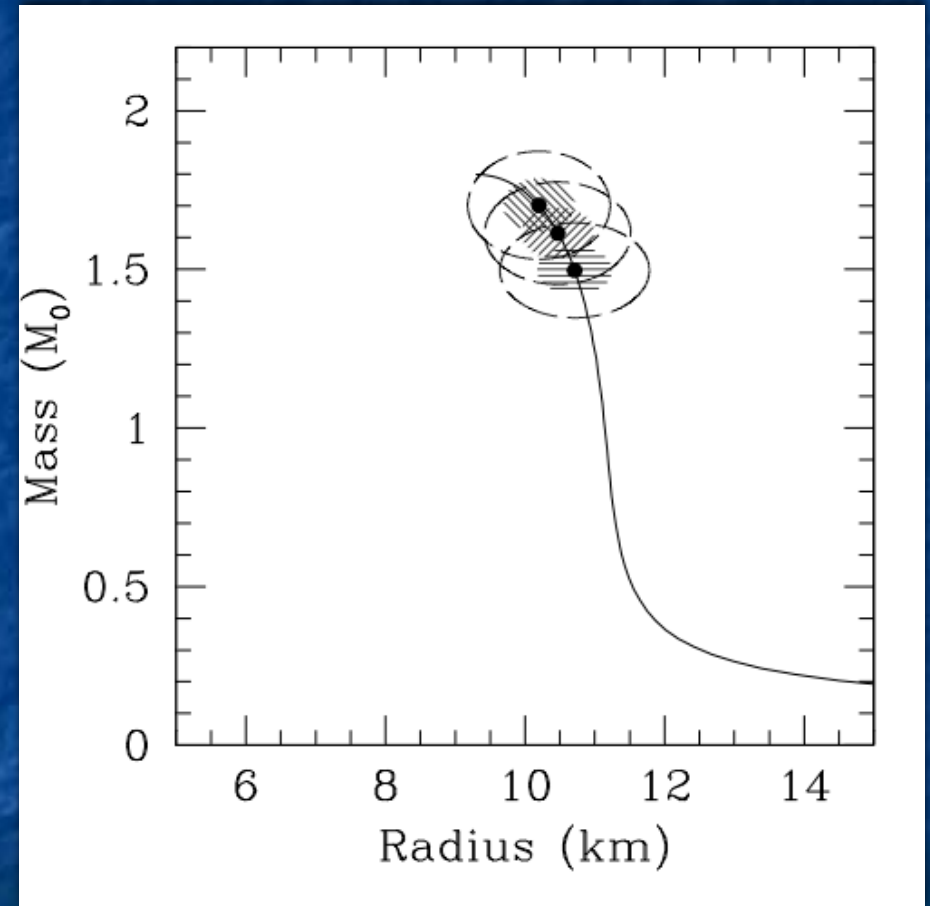
Demorest et al. 2010

Inverse Temptations

- There is a strong temptation to look at the path from measurements to constraints on dense matter as a mathematical problem: just invert the system
- For example, if we have N measurements and an N -parameter model, we just invert. The more measurements the better
- What could be simpler?

Example inversion proposal

- The idea is that we have a parameterized EOS model with three parameters. Three measurements will determine those parameters.
- Simple and clean, right? Just use Jacobians

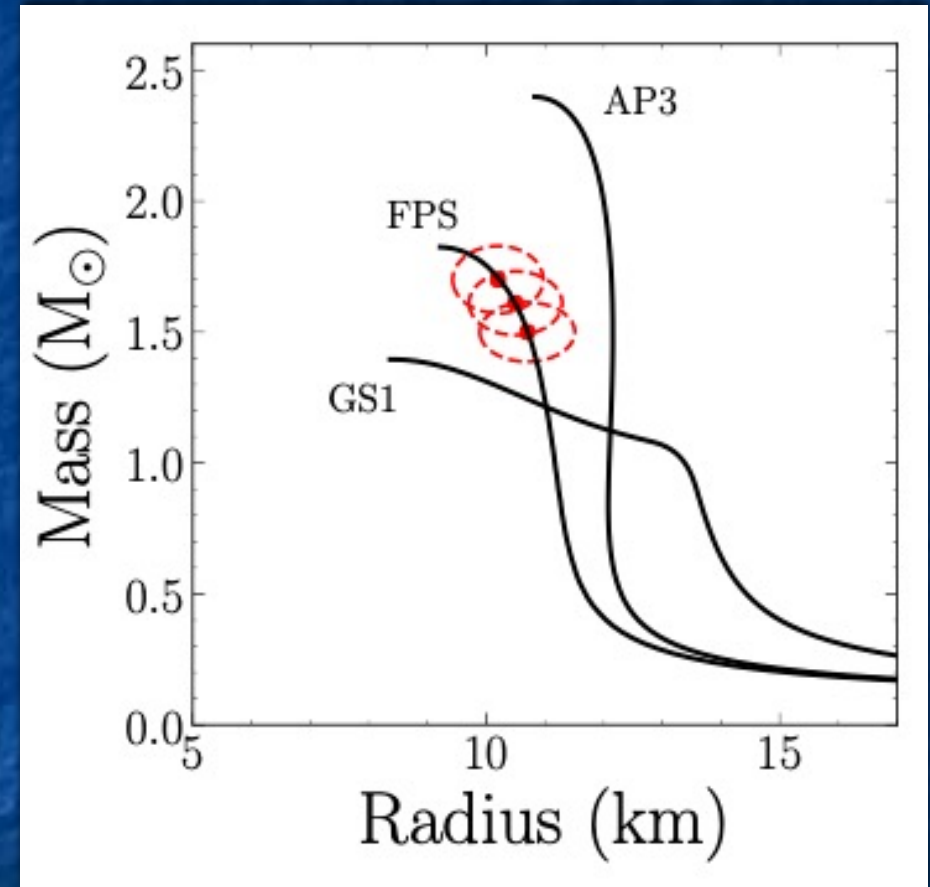


Özel and Psaltis 2009

Problems with inversion

- Several issues:
- The inversion is usually singular (crossing points)
- Some NS might not reach highest density
- Want more measurements than parameters!

Otherwise we don't test the model



Raaijmakers+ 2018

What should we do instead?

- It may be less exciting (assuming you get excited by mathematical inversion...), but the right approach is statistical estimation
- Standard Bayesian analysis: start with priors, update with likelihood of data
- But this brings out an issue: how dependent are we on our priors?

Priors for EOS inference

- People have their own preferences...
- Our group uses three frameworks:
 - Piecewise polytrope (many groups)
 - Spectral decomposition (e.g., Lindblom+)
 - Gaussian processes (e.g., Landry+Essick)
- Many others are possible
- A point from Andrew Steiner: flat in one domain (e.g., M, R) \neq flat in another (P, ρ)

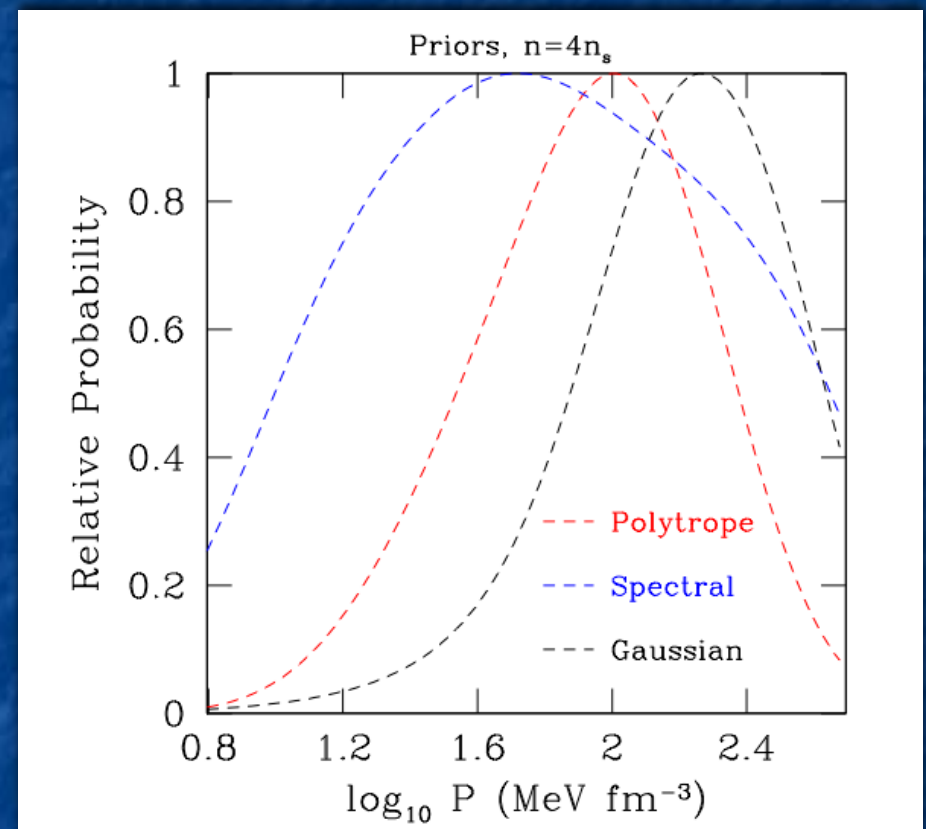
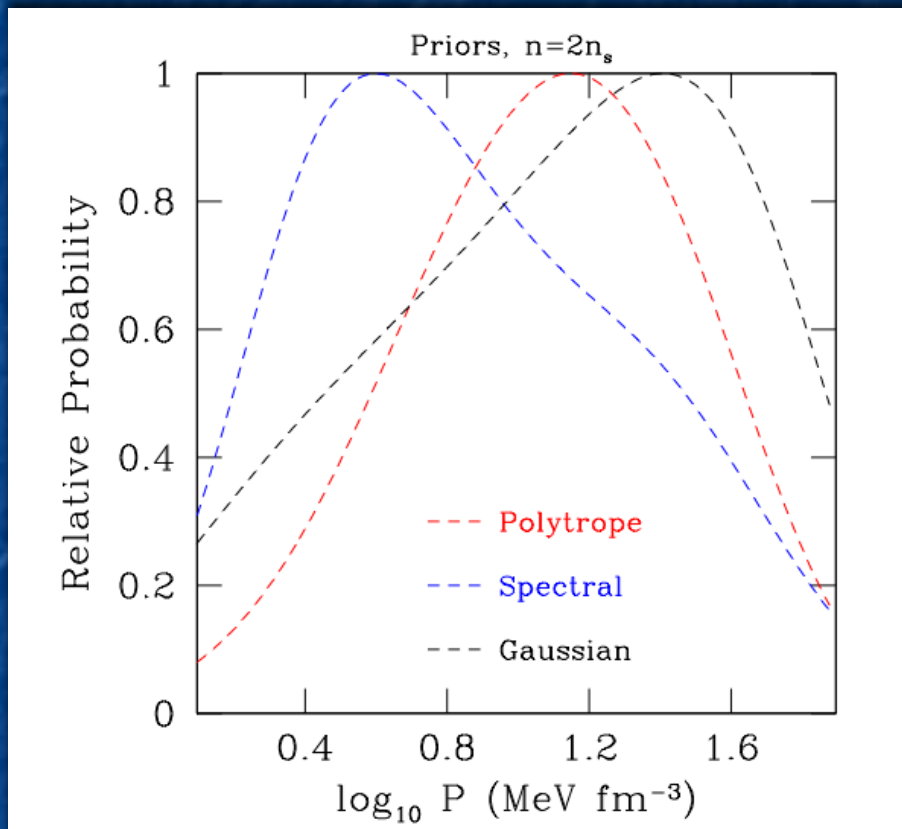
Constraints that we apply

- Symmetry energy at $n_{\text{sat}}=0.16 \text{ fm}^{-3}$
Roughly approximated by $e/n - m_n c^2 + 16 \text{ MeV}$
Assume $E_{\text{Sym}}=32 \pm 2 \text{ MeV}$ at n_{sat}
- Masses of three $\sim 2 M_{\text{sun}}$ NS
- Tidal deformability from GW170817
- Mass-radius of J0030 from NICER
- Mass-radius of J0740 from NICER
- Formalism: Miller, Chirenti, Lamb 2020

A diversion on divergences

- Common to use K-L divergence, $P|Q$:
 $D_{KL}(P|Q) = \sum P_i \log(P_i/Q_i)$, i is over data set
- Asymmetric! Useful for progressive constraints but not as much for distributions on the same footing, e.g., different priors
- Quote both directions for prior comparison
- Shannon entropy: $S = -\sum p_i \log(p_i)$
Use entropy over set of 100,000 EOS
Normalize: $S=1$ if uniform, $=0$ for only 1 EOS

Priors at $2n_{\text{sat}}$ and $4n_{\text{sat}}$

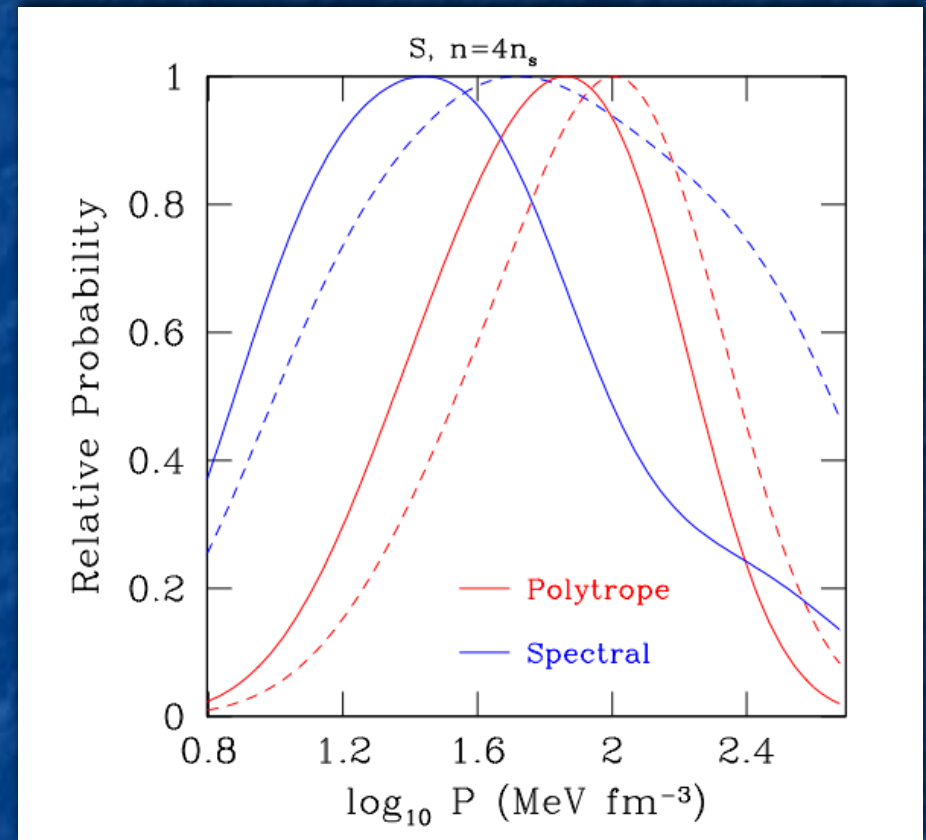
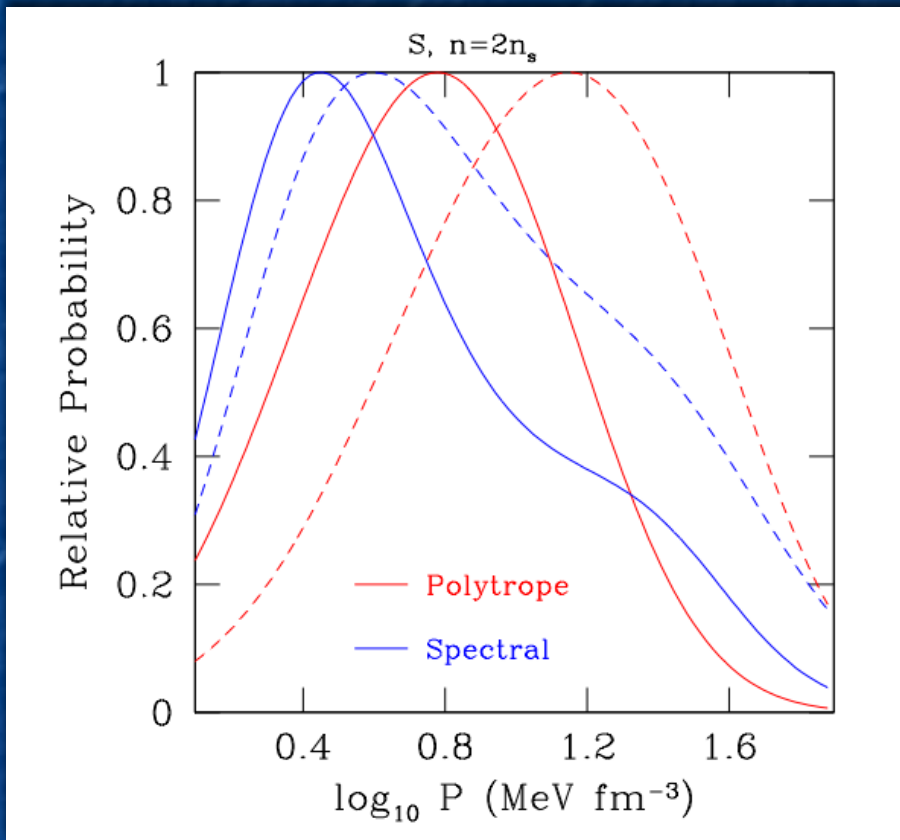


$$D_{\text{KL}}(P,S)=0.15 \text{ or } 0.17$$

$$D_{\text{KL}}(P,S)=0.17 \text{ or } 0.23$$

Entropy = 1 in EOS set, by normalization

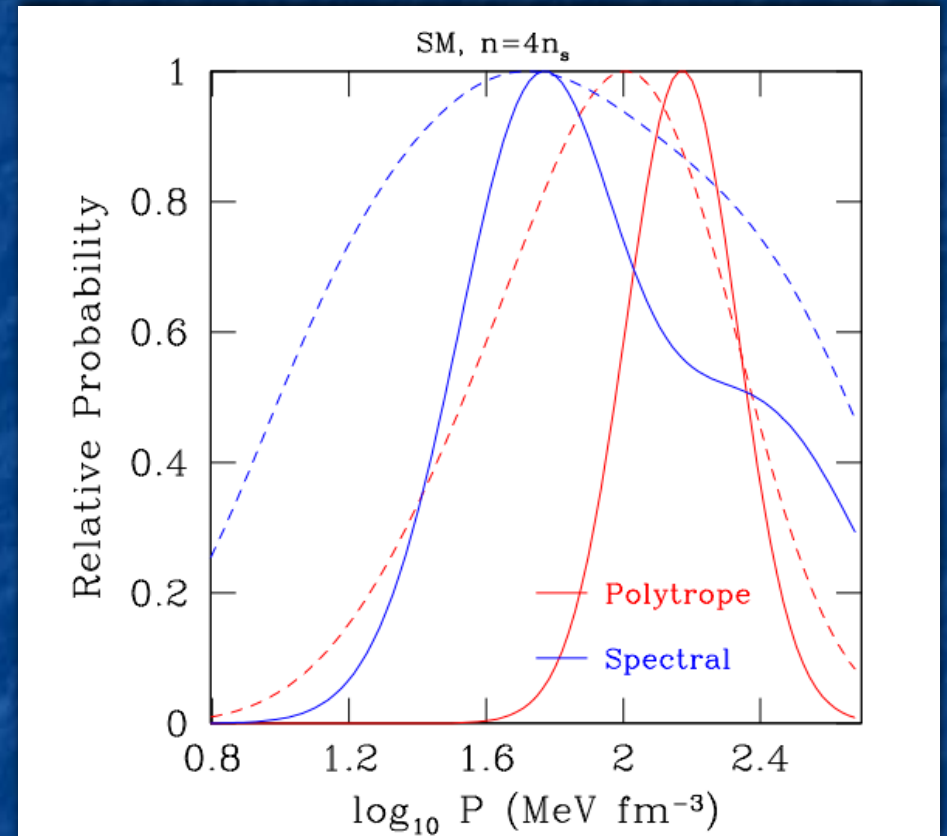
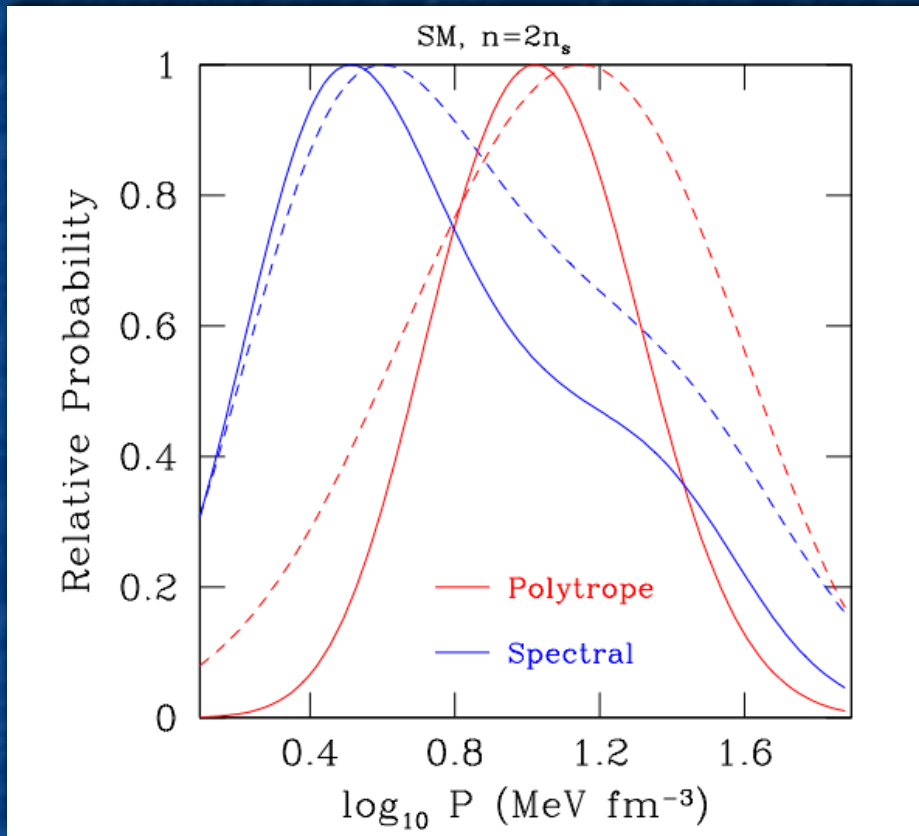
+Symmetry Energy



D_{KL} from prior: $P=0.39$; $S=0.06$ D_{KL} from prior; $P=0.08$; $S=0.13$

Entropy in EOS set is 0.92 for P, 0.97 for S

+High Masses

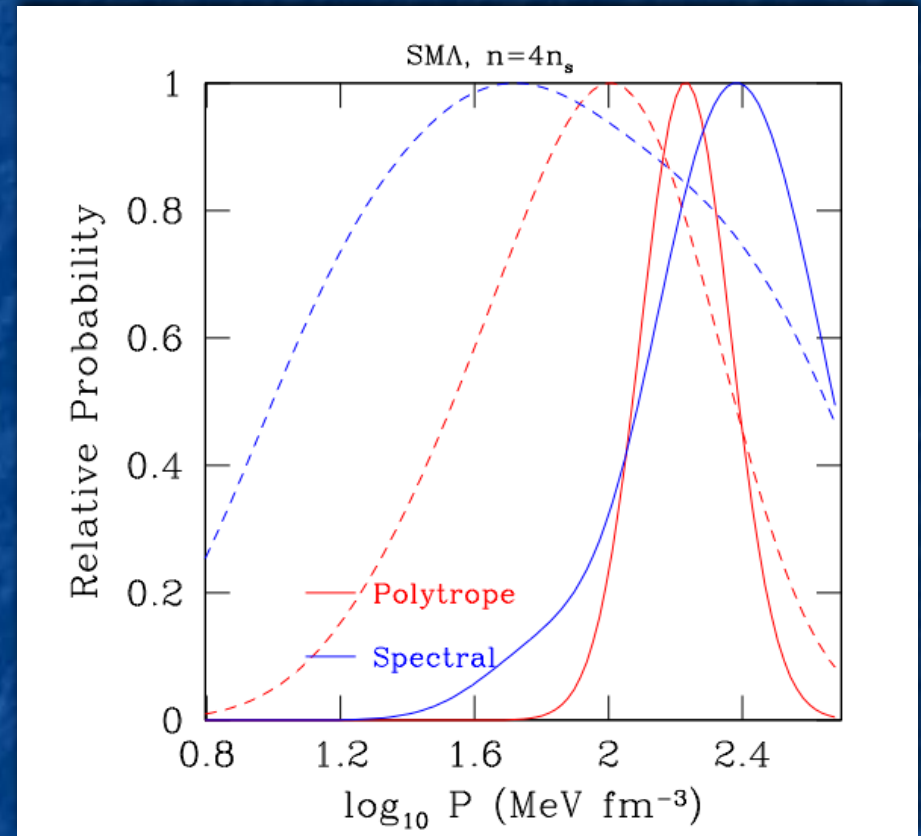
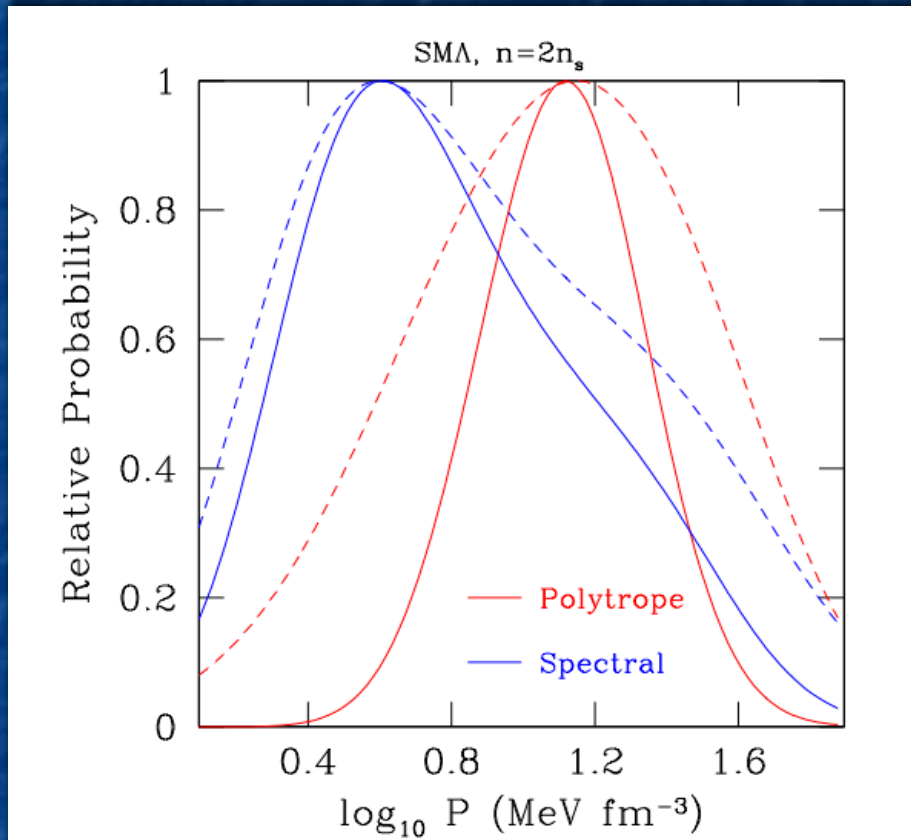


D_{KL} from prior: $P=0.21$; $S=0.03$

D_{KL} from prior; $P=1.94$, $S=0.45$

Entropy in EOS set is 0.83 for P, 0.96 for S

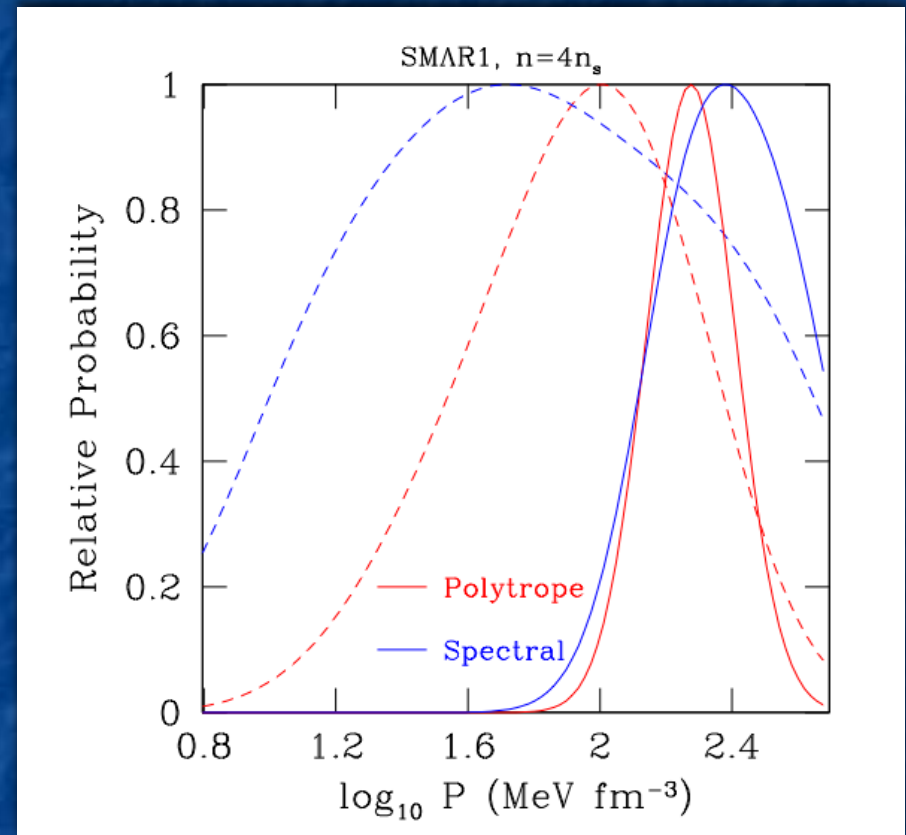
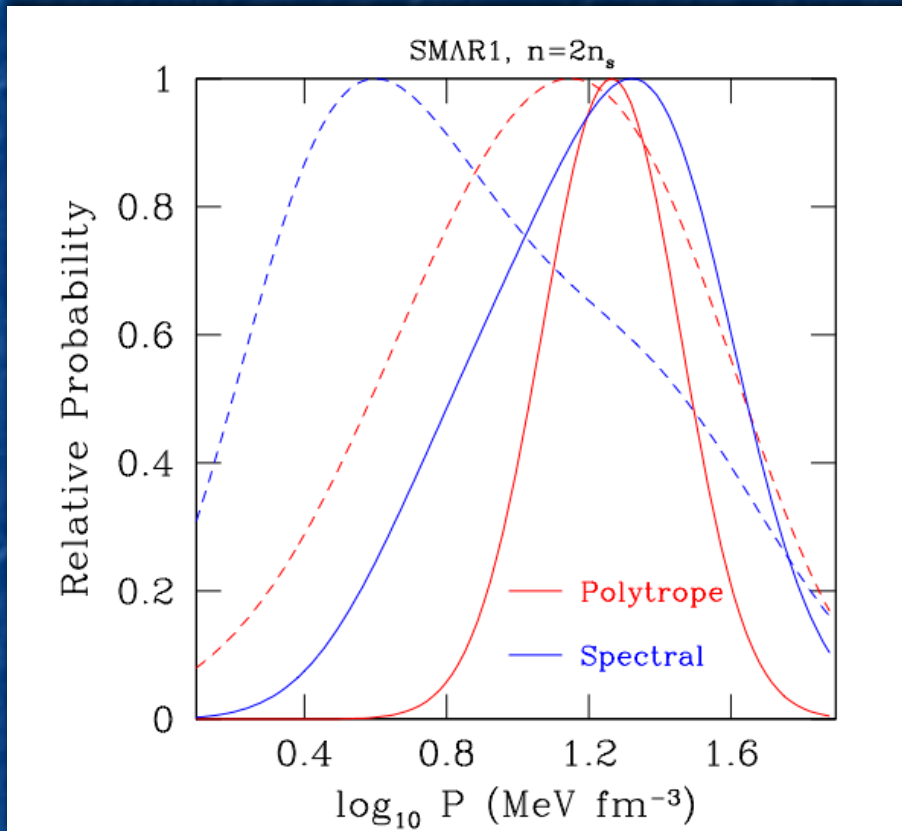
+Tidal Deformability



D_{KL} from prior: $P=0.46$; $S=0.04$ D_{KL} from prior; $P=3.98$, $S=2.16$

Entropy in EOS set is 0.82 for P, 0.95 for S

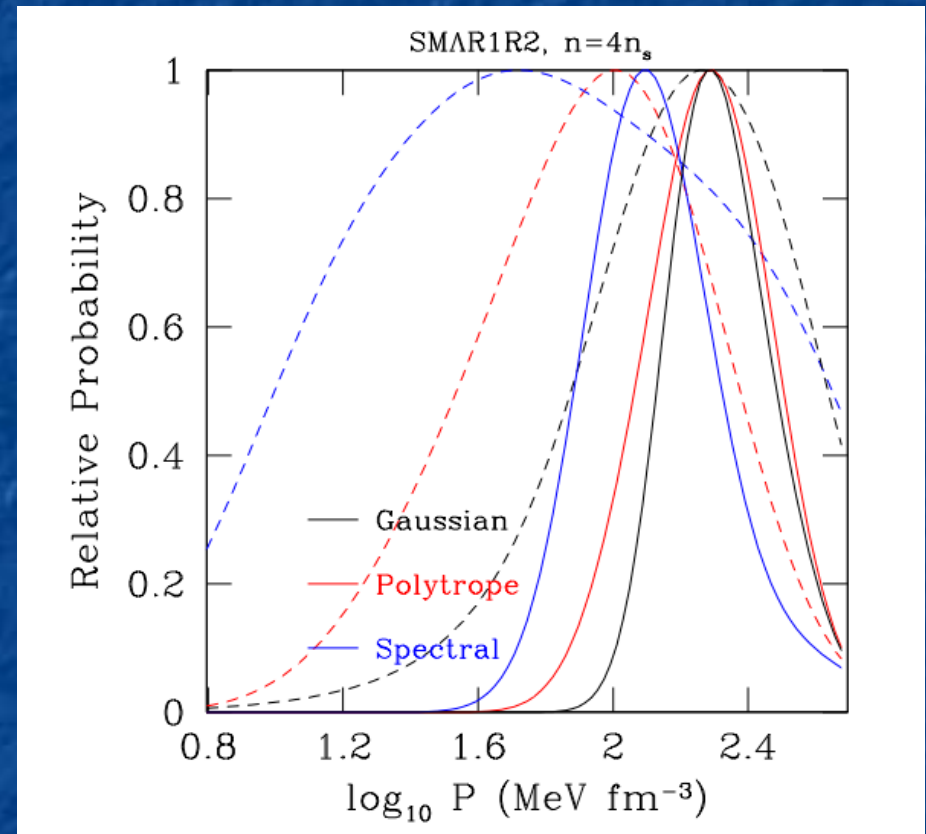
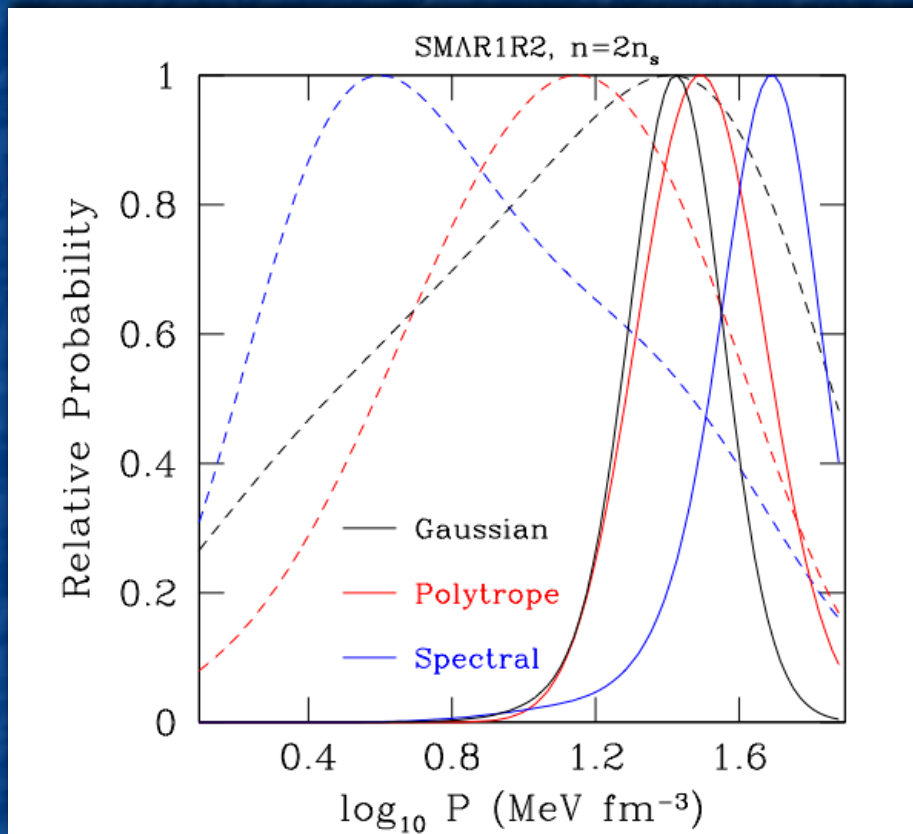
+Radius of J0030



D_{KL} from prior: $P=1.27$; $S=0.60$ D_{KL} from prior; $P=5.59$, $S=6.43$

Entropy in EOS set is 0.81 for P, 0.91 for S

+Radius of J0740



D_{KL} from prior: P=4.77; S=4.27 D_{KL} from prior; P=2.78, S=4.19

Entropy in EOS set is 0.77 for P, 0.70 for S

What does all this tell us?

- Priors can have complicated effects
- If the likelihood pushes us to a tail of the prior, those effects can be amplified
- Polytrope-Spectral K-L divergence is greater in posterior than in prior!
- But this is mainly because the posterior is substantially tighter

Conclusions

- Residual differences between posteriors can be larger than we might think
- But it doesn't mean that we're not making progress!
- It does mean that we need to be careful with our priors