

Reconstruction of nuclear matter parameters in a Bayesian approach

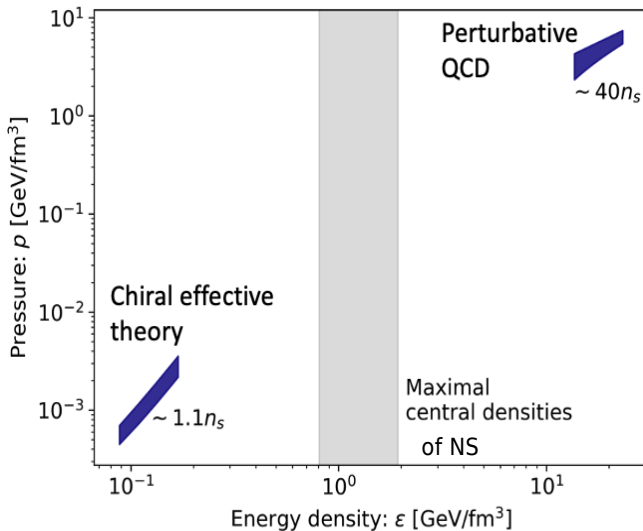
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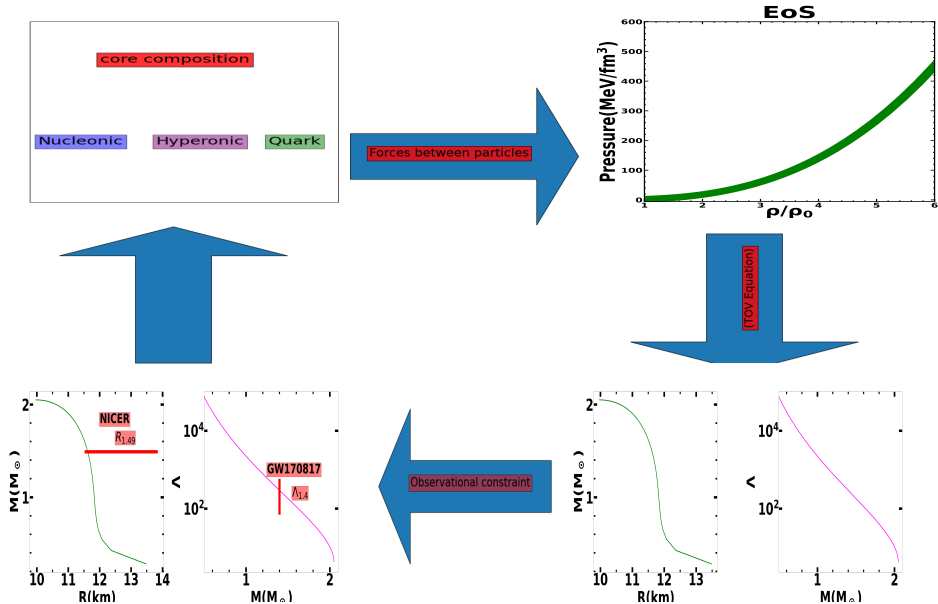
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Introduction



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- charge neutrality : $\rho_p = \rho_e + \rho_\mu$
- β -equilibrium : $\mu_n = \mu_p + \mu_e$
 $\mu_e = \mu_\mu$
- using these equations we can calculate the particle fraction and $(\rho_n - \rho_p)/\rho = \delta$: Isospin asymmetry parameter

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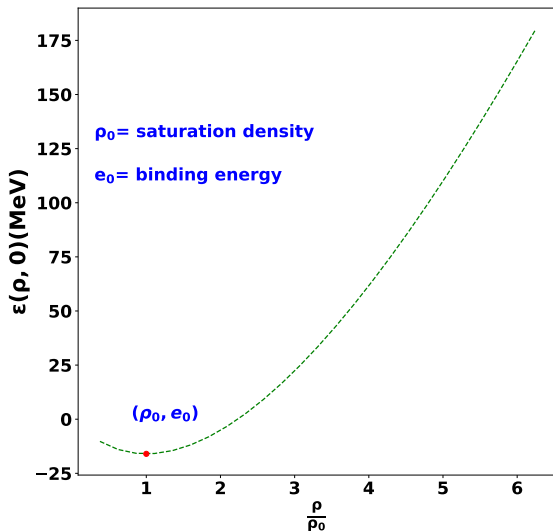
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- $\rho_n(\rho_p)$:Neutron(Proton) density

Variation of energy of SNM with baryon density



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- $\varepsilon(\rho, \delta) = \sum_n \frac{1}{n!} (a_n + b_n \delta^2) x^n$

■ **SNM parameters :**

■ $a_0 = e_0 \equiv$ Binding Energy at ρ_0

■ $a_1 = 0$

■ $a_2 = K_0 \equiv$ Incompressibility Coefficient at ρ_0

■ $a_3(a_4) = Q_0(Z_0) \equiv$ Third(Fourth)order Derivative at ρ_0

■ **Symmetry Energy parameters :**

■ $b_0 = J_0 \equiv$ Symmetry Energy at ρ_0

■ $b_1 = L_0 \equiv$ Slope of Symmetry Energy at ρ_0

■ $b_2 = K_{sym,0} \equiv$ Symmetry Energy Curvature at ρ_0

■ $b_3(b_4) = Q_{sym,0}(Z_{sym,0}) \equiv$ Third(Fourth)order Derivative at ρ_0

n/3 Expansion

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$$\begin{pmatrix} e_0 \\ 0 \\ K_0 \\ Q_0 \\ Z_0 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 3 & 4 & 5 & 6 \\ -2 & 0 & 4 & 10 & 18 \\ 8 & 0 & -8 & -10 & 0 \\ -56 & 0 & 40 & 40 & 0 \end{pmatrix} \begin{pmatrix} a'_0 \\ a'_1 \\ a'_2 \\ a'_3 \\ a'_4 \end{pmatrix}$$
$$\begin{pmatrix} J_0 \\ L_0 \\ K_{\text{sym},0} \\ Q_{\text{sym},0} \\ Z_{\text{sym},0} \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 3 & 4 & 5 & 6 \\ -2 & 0 & 4 & 10 & 18 \\ 8 & 0 & -8 & -10 & 0 \\ -56 & 0 & 40 & 40 & 0 \end{pmatrix} \begin{pmatrix} b'_0 \\ b'_1 \\ b'_2 \\ b'_3 \\ b'_4 \end{pmatrix} .$$

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- The parameters which are almost unknown:
 $Q_0, Z_0, K_{sym,0}, Q_{sym,0}, Z_{sym,0}$

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- We want to reconstruct these parameters from a given EoS(known precisely)

Bayesian estimation

- This approach is mainly based on the Bayes theorem which states that,
- $P(\theta|D) = \frac{\mathcal{L}(D|\theta)P(\theta)}{\mathcal{Z}}$
- θ :model parameters and D :Data
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- $P(\theta_i|D)$: Marginalised posterior distribution of the parameters $(\theta_i) = \int P(\theta|D) \prod_{k \neq i} d\theta_k$

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Table: The values of nuclear matter parameters (in MeV) which are employed to construct various pseudo data using the Taylor and $\frac{n}{3}$ expansions.

N	Symmetric nuclear matter		Symmetry energy	
0	e_0	-16.0	J_0	32.0
1			L_0	50.0
2	K_0	230	$K_{\text{sym},0}$	-100
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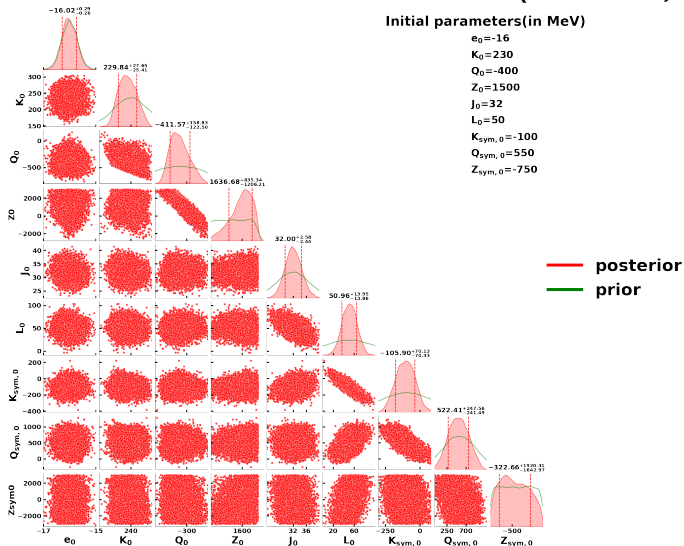
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- Causality ($\frac{dP}{d\epsilon} \leq 1$)
- Maximum Mass for that EoS $\geq 2M_{\odot}$

Table: Two different sets P1 and P2 for the prior distributions of the nuclear matter parameters (in MeV). The saturation density ρ_0 is taken to be 0.16 fm^{-3} .

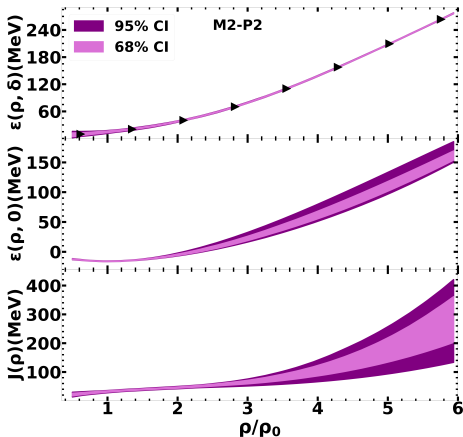
Parameters	P1			P2		
	Pr-Dist	μ min	σ max	Pr-Dist.	μ min	σ max
e_0	G	-16	0.3	G	-16	0.3
K_0	G	240	100	G	240	50
Q_0	U	-2000	2000	G	-400	400
Z_0	U	-3000	3000	U	-3000	3000
J_0	G	32	5	G	32	5
L_0	U	20	150	G	50	50
$K_{\text{sym},0}$	U	-1000	1000	G	-100	200
$Q_{\text{sym},0}$	U	-2000	2000	G	-550	400
$Z_{\text{sym},0}$	U	-3000	3000	U	-3000	3000

Posterior Distribution of the Parameters (n/3 model,P2):

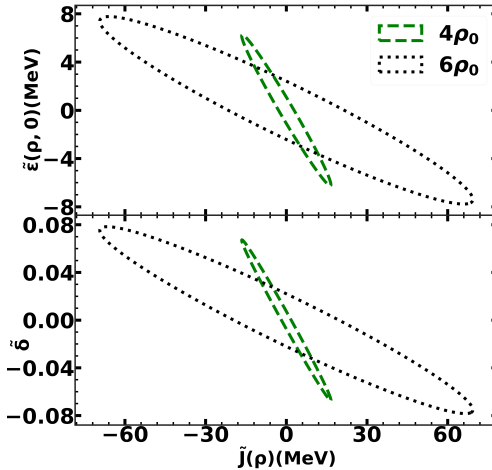


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■ Posterior distribution of the EoS (n/3 model,P2):



■ Confidence ellipse (n/3 model, P2) :



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- The median values of second or higher order NMPs show sizeable deviations from their true values and associated uncertainties are larger.
- The sources of these uncertainties are intrinsic in nature, identified as:
 - (i) Correlations among various NMPs and
 - (ii) The balance between the EoS of symmetric nuclear matter, symmetry energy, and the neutron-proton asymmetry

Thank You!
For Your Attention