

Applying DNN to extract nuclear matter properties from neutron stars

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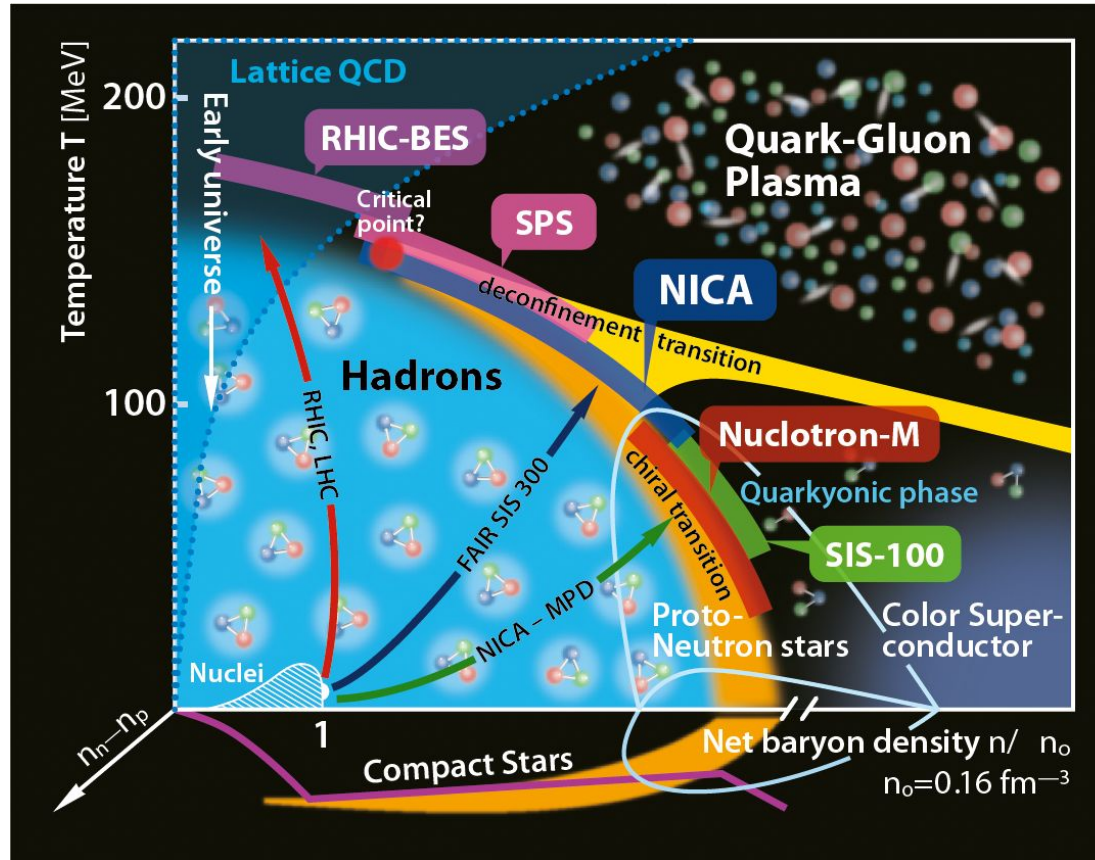
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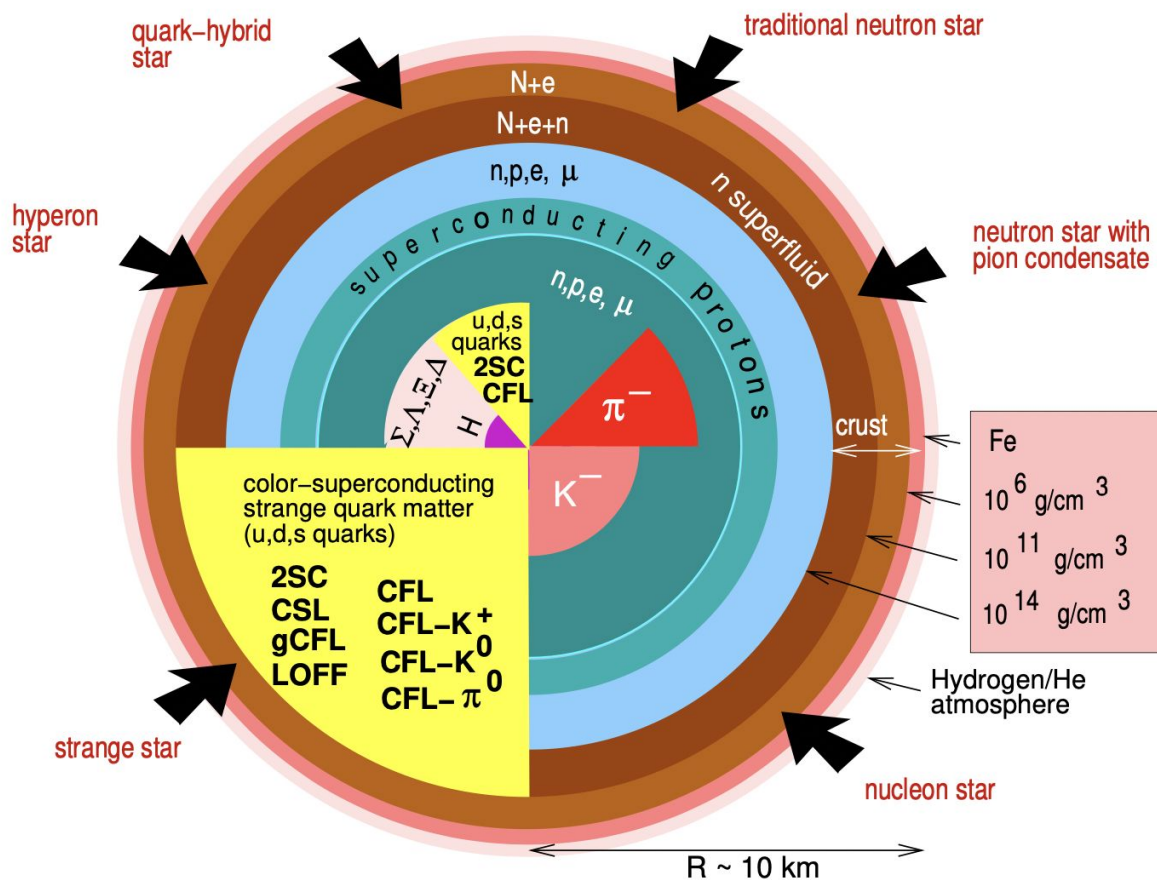
Phase Diagram of QCD



- The different QCD matter phases are still unknown
- Lattice QCD simulations cannot be performed at finite chemical potential
- **Neutron stars can help us in understanding the low T and dense region**

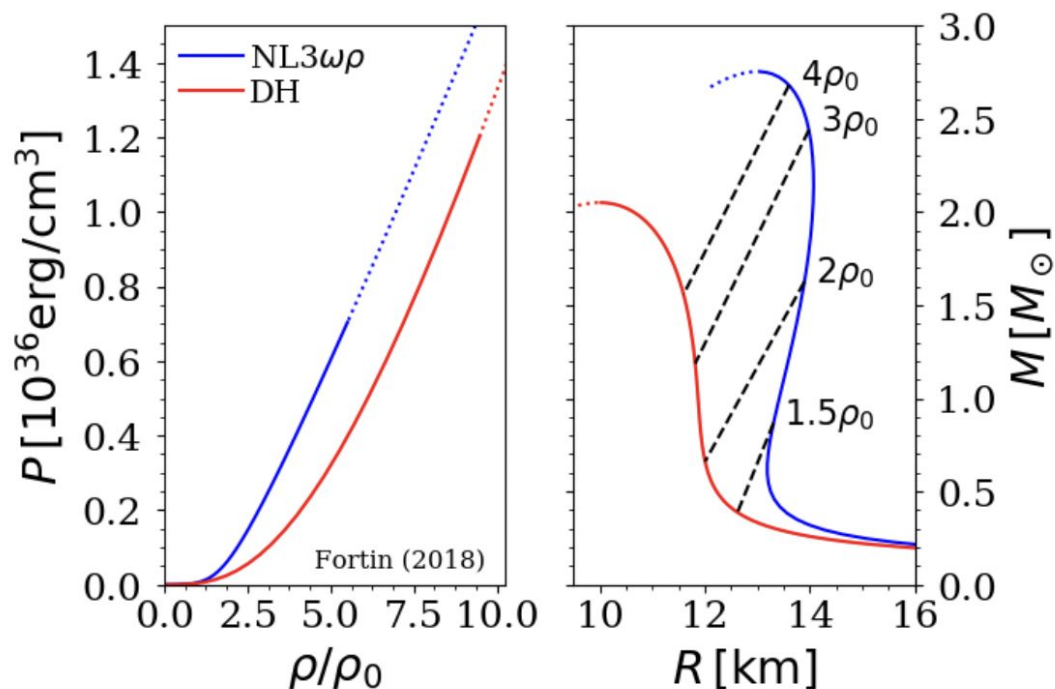
Several hypothesis for neutron star matter

- However, as the QCD solution is still unknown, there are several possible scenarios for NS matter



From the EOS to the mass-radius relation

- We need a microscopic model of nuclear matter: $\mathbf{P}(\mathbf{n})$ and $\mathbf{E}(\mathbf{n})$
- Neutron star matter is cold matter ($T=0$) in beta equilibrium
- Solve **Tolman–Oppenheimer–Volkoff equation**: $M(R)$ sequence



Observational constraints: neutron stars M/R

- **Highest NS mass measured**

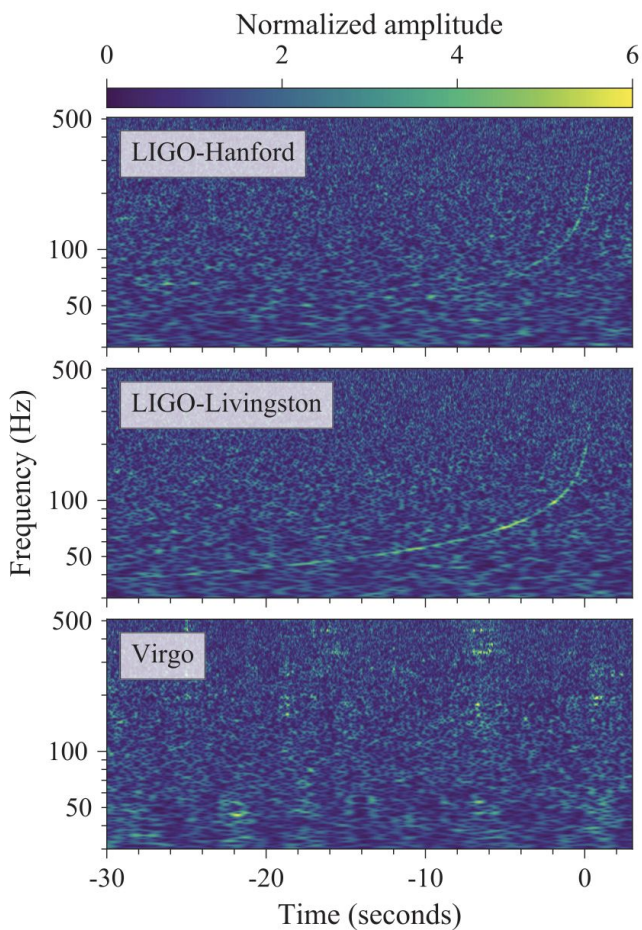
- $M = 1.908 \pm 0.016 M_{\odot}$ (J1614 - 2230)
[Z. Arzoumanian et al., *Astrophys. J., Suppl. Ser.* 235, 37 (2018)]
- $M = 2.08 \pm 0.07 M_{\odot}$ (J0740 + 6620)
[E. Fonseca et al., *Astrophys. J. Lett.* 915, L12 (2021)]
- $M = 2.13 \pm 0.04 M_{\odot}$ (J1810+171)
[R. Romani, et. al., *Astrophys. J. Lett.*, 908, L46]

- **NS masses and radii**

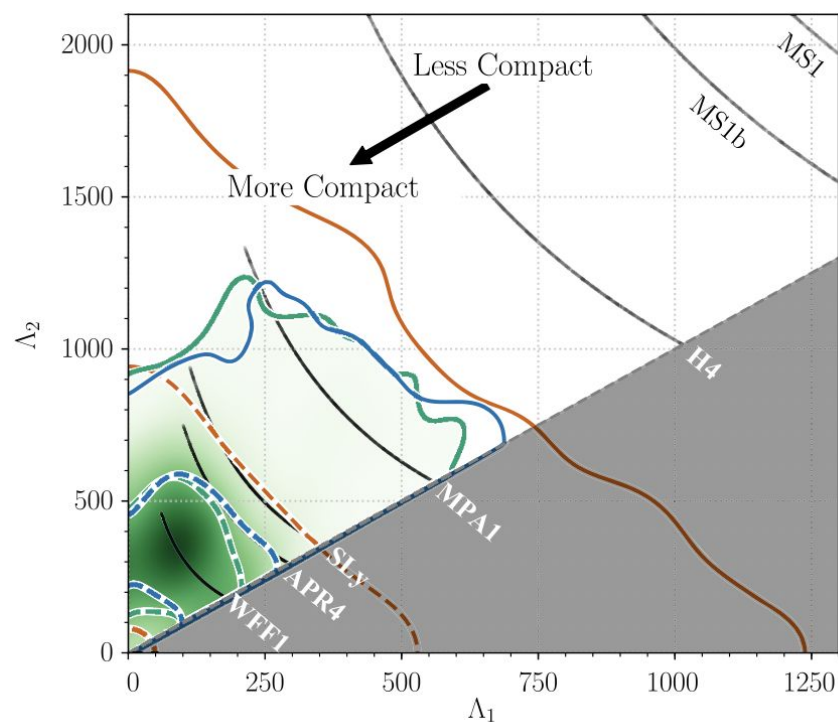
- $M = 1.34_{-0.16}^{+0.15} M_{\odot}$ and $R = 12.71_{-1.19}^{+1.14}$ km (J0030 + 0451)
[T. E. Riley et al., *Astrophys. J. Lett.* 887, L21 (2019)]
- $M = 1.44_{-0.14}^{+0.15} M_{\odot}$ and $R = 13.02_{-1.06}^{+1.24}$ km (J0030 + 0451)
[M. C. Miller et al., *Astrophys. J. Lett.* 887, L24 (2019)]
- $M = 2.072_{-0.066}^{+0.067} M_{\odot}$ and $R = 12.39_{-0.98}^{+1.30}$ km (PSR J0740+6620)
[T. E. Riley, et. al., *Astrophys.J.Lett.* 918 (2021) 2, L27]

Any reliable nuclear model must satisfy these constraints

GW constraints: BNS inspiral (GW170817)



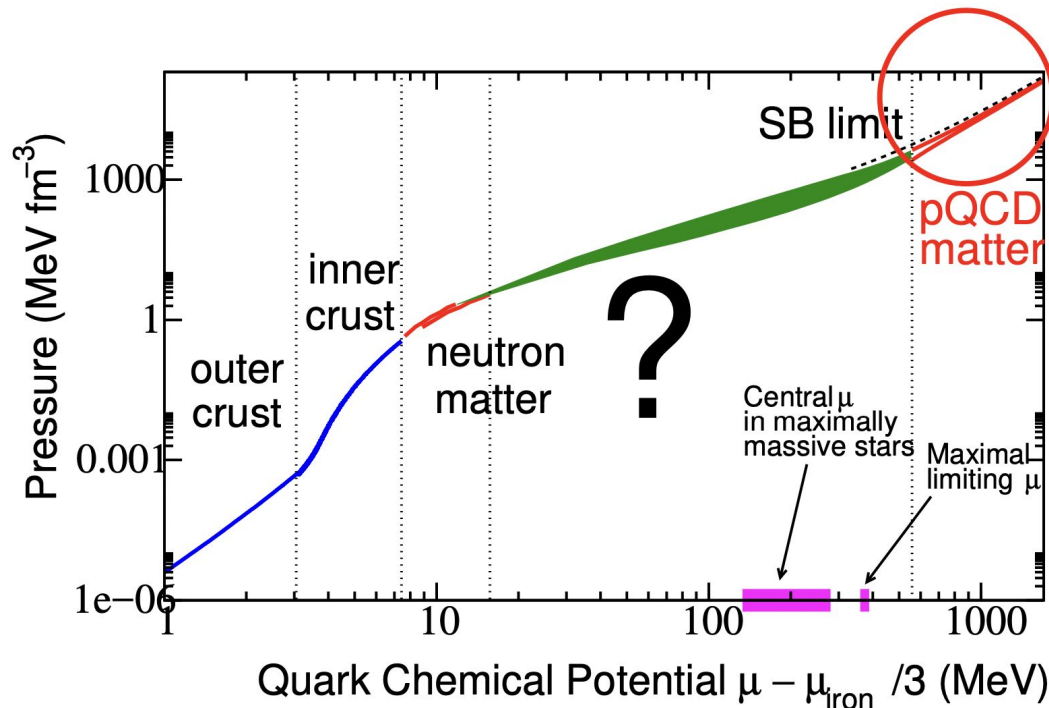
Λ (tidal deformability) quantifies how deformed is the NS when subject to an external tidal field.



PRL 119, 161101 (2017), PRL 121, 161101 (2018)

Allows to restrict the set of EOS of NS matter

High density constraints: perturbative QCD



Kurkela ApJ 789, 2014

- Perturbative QCD at zero temperature
- Good convergente at around $\mu_B = 2.6$ GeV

Kurkela et. al. PRD81 2010

Nuclear matter properties

- Nuclear matter properties are usually characterized by

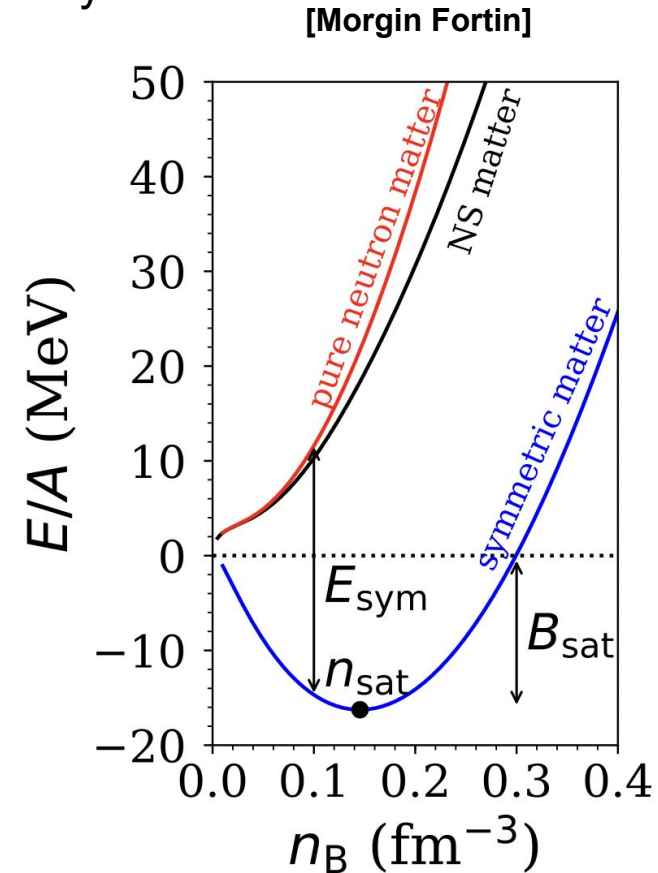
$$\frac{E_{nuc}}{A}(n, \delta) = \frac{E_{SNM}}{A}(n) + E_{sym}(n) \delta^2$$

- $n = n_n + n_p$ is the baryonic density
- $\delta = (n_n - n_p)/(n_n + n_p)$ is the asymmetry

$$\frac{E_{SNM}}{A}(n) = E_{sat} + \frac{K_{sat}}{2} \eta^2 + \frac{J_{sat}}{3!} \eta^3 + \frac{Z_{sat}}{4!} \eta^4$$

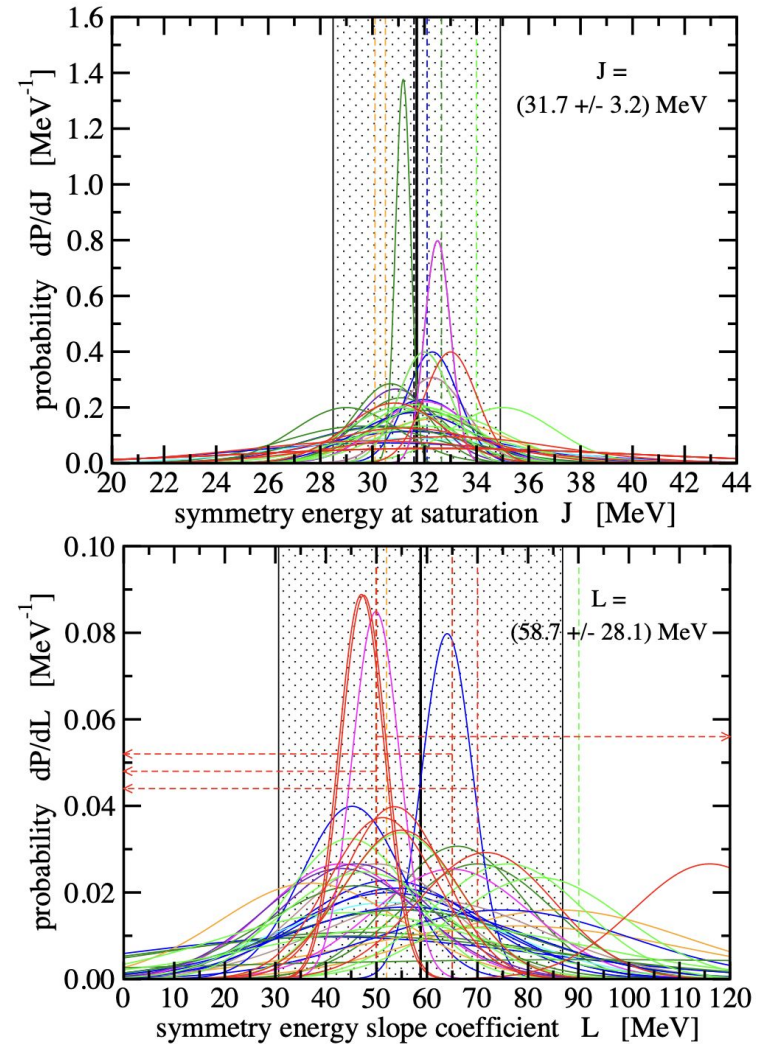
$$E_{sym}(n) = E_{sym} + L_{sym} \eta + \frac{K_{sym}}{2} \eta^2 + \frac{J_{sym}}{3!} \eta^3 + \frac{Z_{sym}}{4!} \eta^4$$

- $\eta = (n - n_0)/(3n_0)$.



Constraints on the symmetry energy

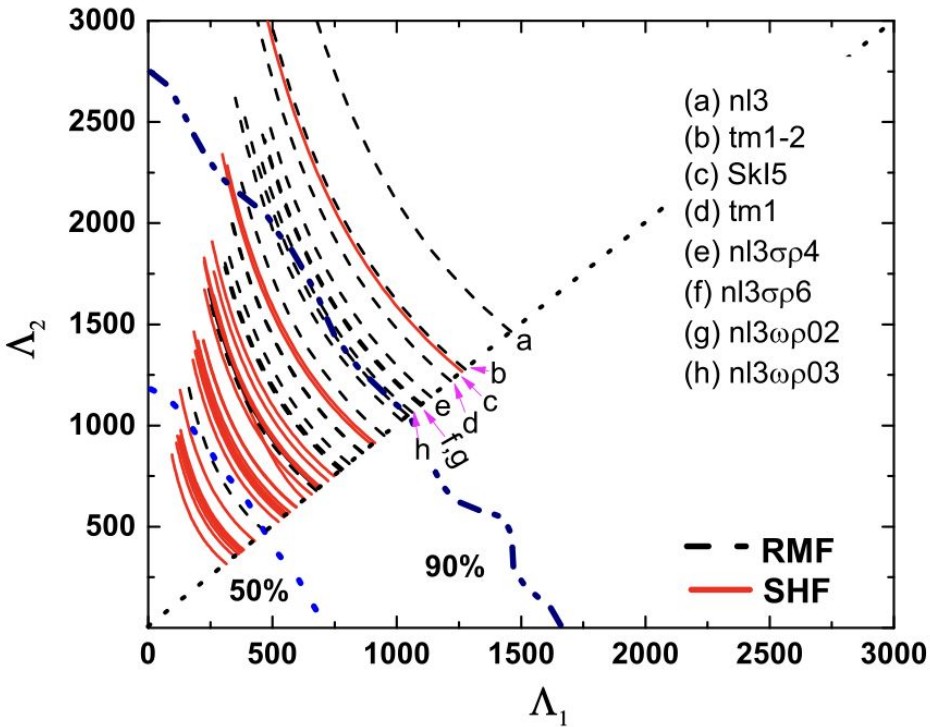
- systematic of nuclear masses
- nuclear structure information
- nuclear resonances
- dipole polarizability of nuclei
- heavy-ion collisions
- theoretical calculations
- properties of neutron stars



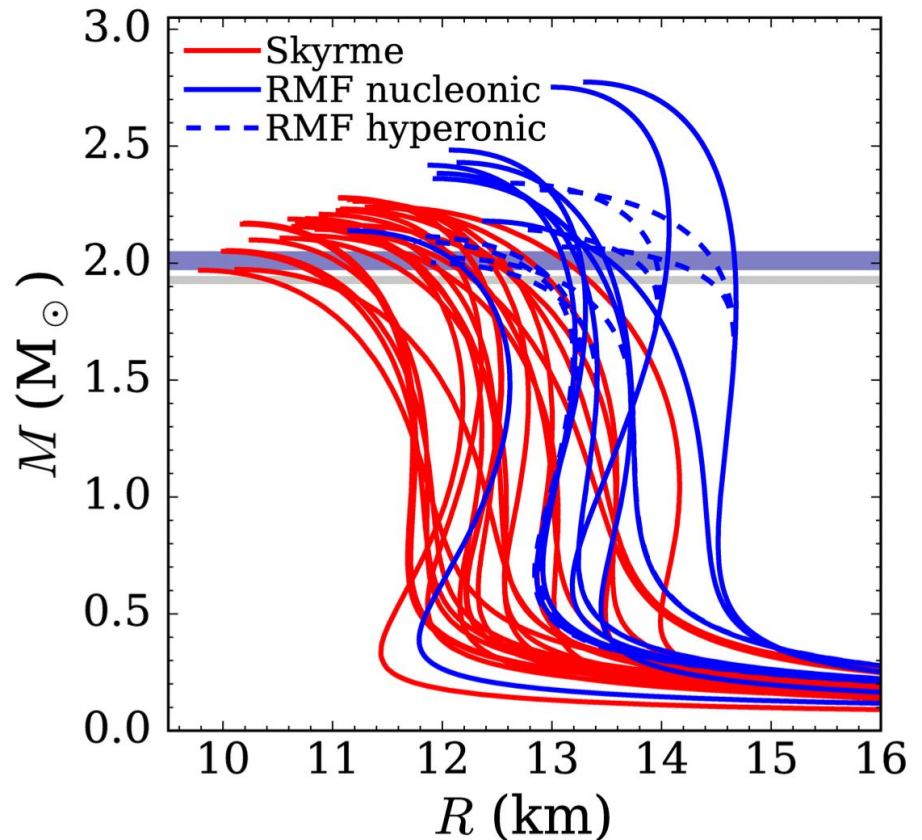
Uncertainties on nuclear parameters

- E_{sat} and n_{sat} known to a few %
- E_{sym} and K_{sat} known within 10%
- L_{sym} known within 50%
- **Other remain almost unknown**

Nuclear models vs. NS observations



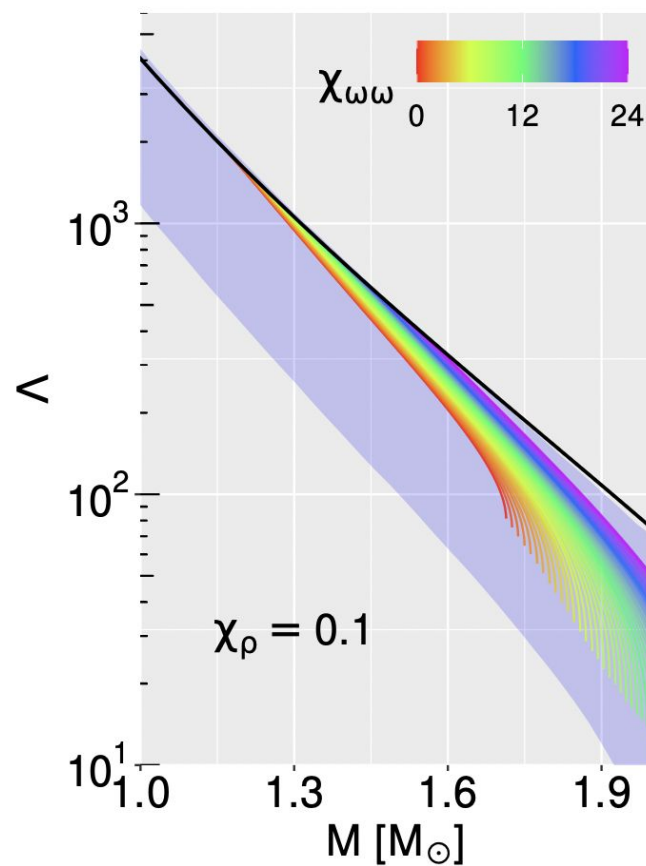
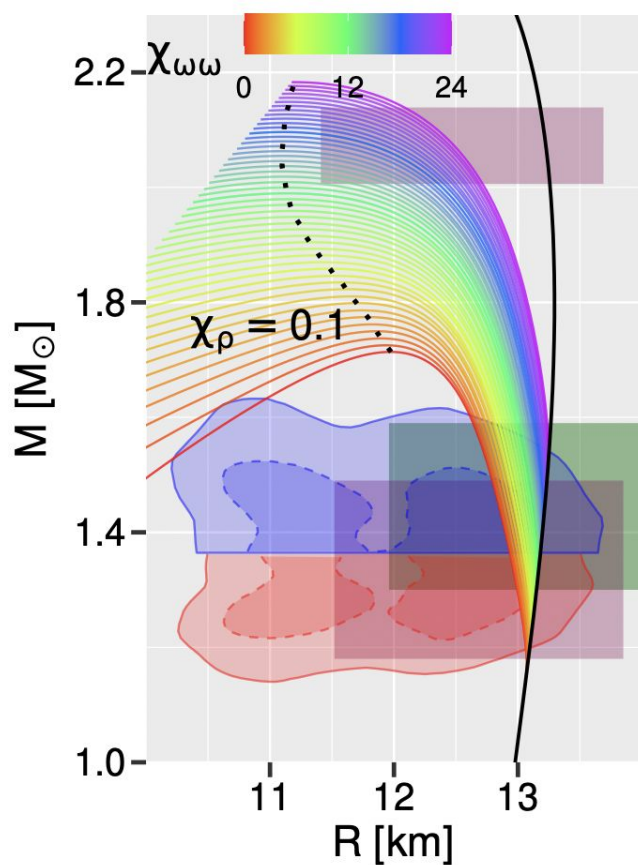
Tuhin Malik, et. al., *Phys.Rev.C* 98 (2018) 3, 035804



M. Fortin, et.al., *Phys.Rev.C* 94 (2016) 3, 035804

Different scenarios for NS matter

- **Hybrid stars:** first-order phase transition from hadronic to quark matter

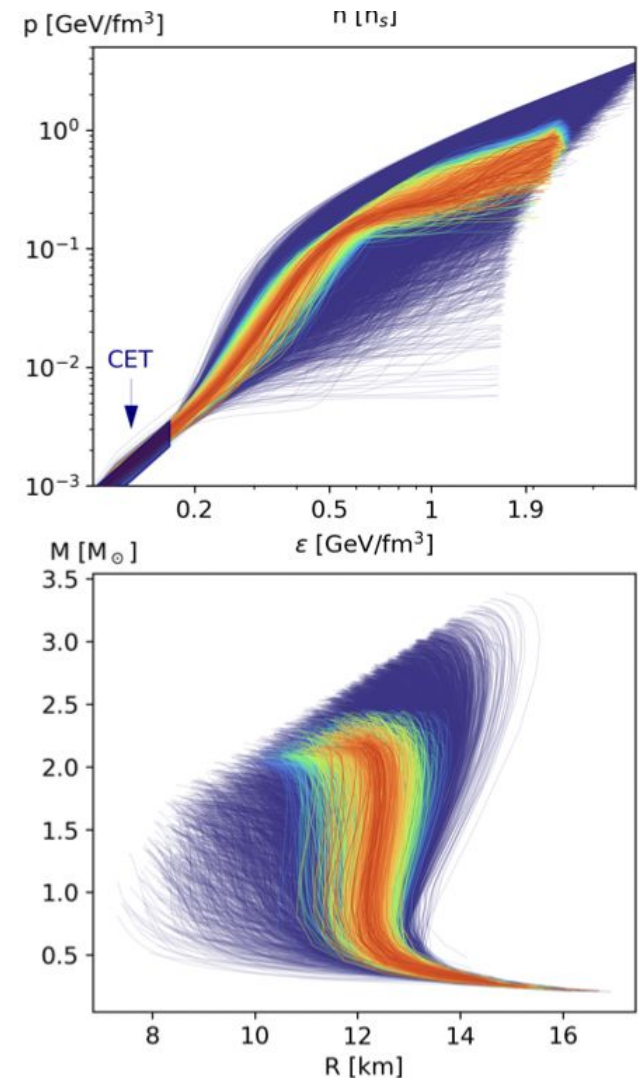


EOS parametrization and inference

- Parametrize the EOS: Gaussian processes
Prior: $P(\text{EOS})$
- Bayesian inference:
 $P(\text{EOS}|\text{data}) = P(\text{data}|\text{EOS})P(\text{EOS})/P(\text{data})$
- The data corresponds to astrophysical observations and theoretical calculations

The future observations will help to constrain the EOS of NS

Can we determine the nuclear matter properties from the NS EOS?



Generating the dataset

- Supervised problem: $\mathbf{y}(\mathbf{x})$
 - \mathbf{X} : energy density of beta-equilibrium matter

$$\mathbf{x}_i = [e(n_1), e(n_2), e(n_3), \dots, e(n_{41})]$$

$$0.08 \leq n_i / \text{fm}^{-3} \leq 0.30$$

[Tovar et. al., *Phys.Rev.D* 104 (2021) 12, 123036]

- \mathbf{Y} : Properties of nuclear matter

$$\mathbf{y}_i = (E_{sat}, K_{sat}, J_{sat}, E_{sym}, L_{sym}, K_{sym}, J_{sym}, n_0)$$

| | [min, max] | # of points |
|-----------|----------------|-------------|
| E_{sat} | [-16.7, -14.9] | 6 |
| K_{sat} | [170, 290] | 6 |
| J_{sat} | [-900, 150] | 6 |
| Z_{sat} | [-3500, 2500] | 6 |
| E_{sym} | [26, 38] | 6 |
| L_{sym} | [15, 105] | 6 |
| K_{sym} | [-400, 200] | 6 |
| J_{sym} | [-1200, 1200] | 6 |
| Z_{sym} | [-3500, 2500] | 6 |
| n_0 | [0.145, 0.166] | 16 |

Dataset size: 9×10^7 EOS

Deep Neural Network

- We want a good interpolating model

- Loss Function (mse)

$$L(\mathbf{w}) = \sum_{i=1}^M (\hat{\mathbf{y}}_i(\mathbf{w}) - \mathbf{y}_i)^2$$

- Adam optimizer: lr =0.001

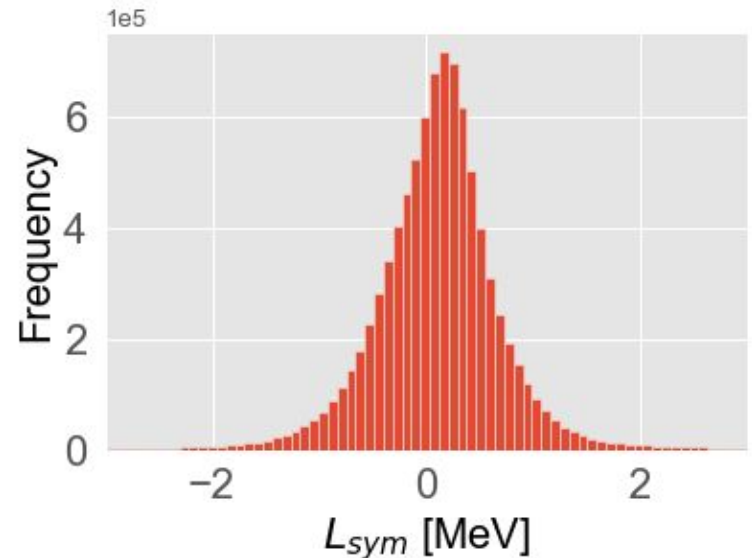
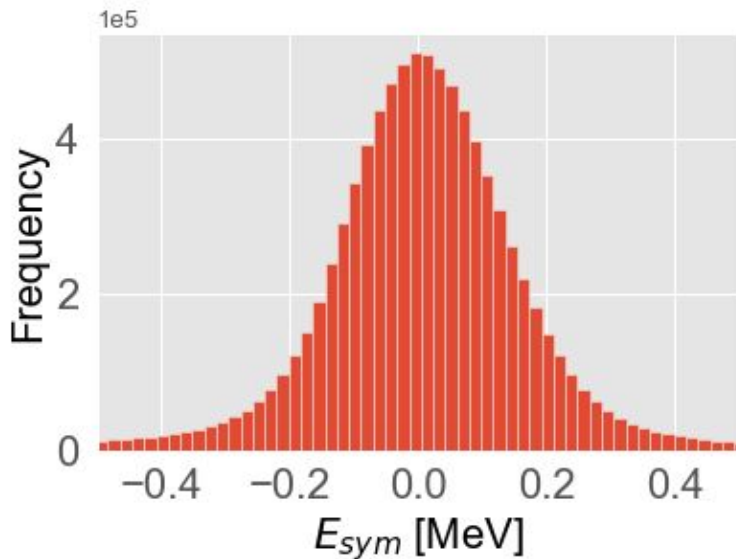
- Validation/training split of 0.2

| Layer | Activation function | size |
|-------|---------------------|------|
| 0 | - | 41 |
| 1 | Sigmoid | 80 |
| 2 | Sigmoid | 80 |
| 3 | Sigmoid | 80 |
| 4 | Sigmoid | 40 |
| 5 | Sigmoid | 15 |
| 6 | Linear | 8 |

Performance on the test set

- Test set size: 9×10^6 EOS

| | rho0 | esat | ksat | qsat | esym | lsym | ksym | qsym |
|-------------|----------|----------|-----------|------------|----------|----------|-----------|------------|
| mean | -0.00017 | -0.01678 | -5.35536 | 9.44117 | 0.00324 | 0.10386 | 8.03442 | 3.56895 |
| std | 0.00117 | 0.11622 | 16.67010 | 171.85117 | 0.25254 | 1.26632 | 23.22310 | 207.55388 |
| 10% | -0.00086 | -0.10795 | -25.20679 | -185.83968 | -0.18173 | -0.59669 | -16.39285 | -221.66793 |
| 90% | 0.00052 | 0.07520 | 11.41779 | 197.97955 | 0.19610 | 0.76707 | 35.22026 | 234.07942 |



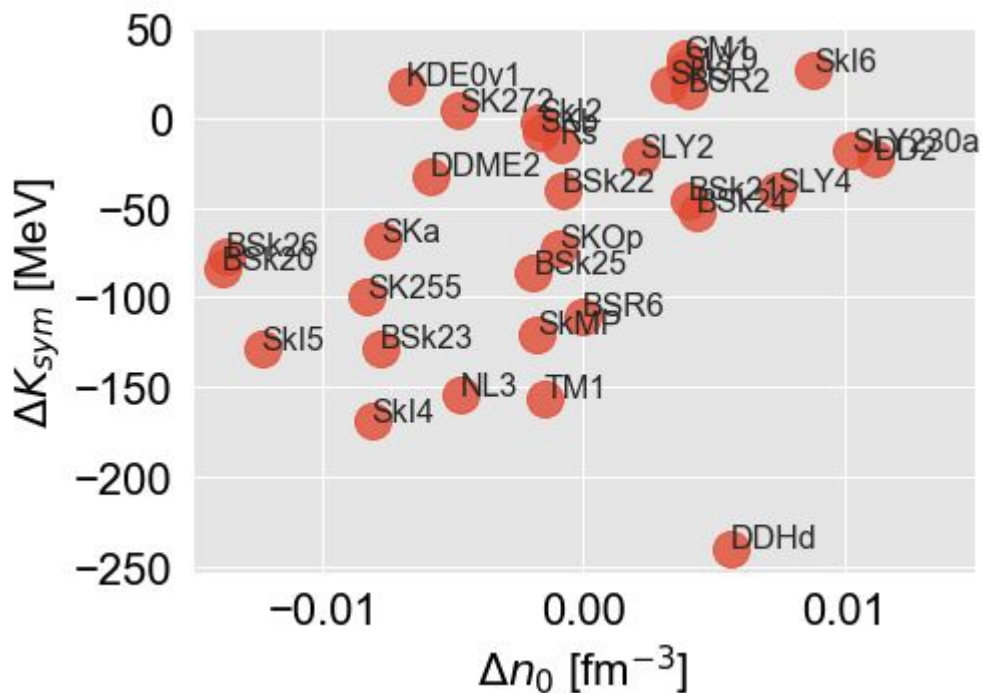
Applying the DNN to set of nuclear models

- Set of 33 EOS built from a relativistic mean field approach and non-relativistic Skyrme interactions

M. Fortin, et. al., Phys.Rev.C 94 (2016) 3, 035804

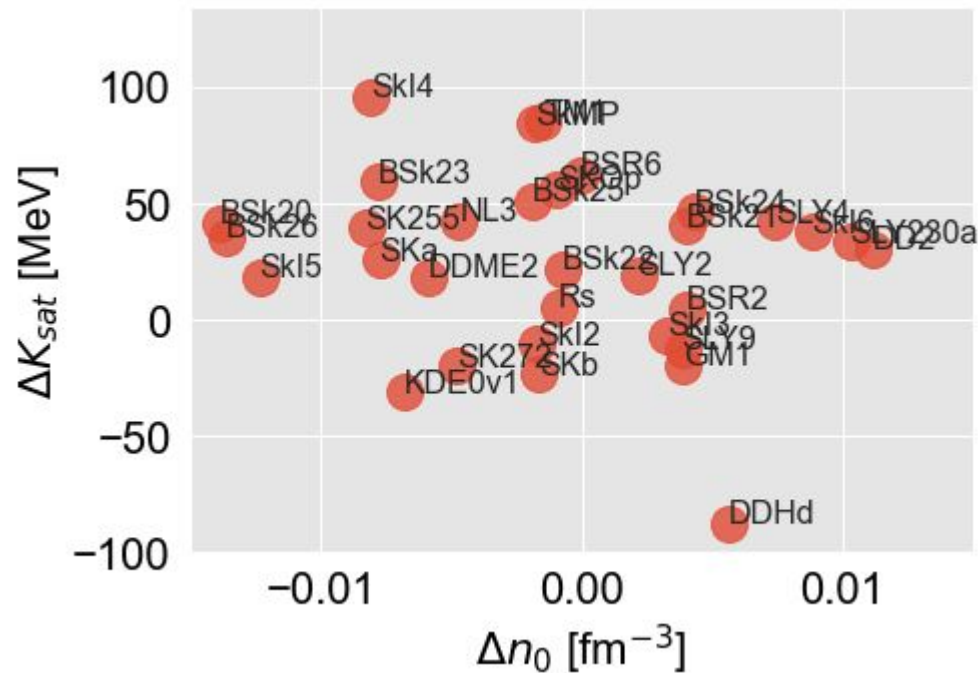
| Model | n_s (fm^{-3}) | E_s (MeV) | K (MeV) | J (MeV) | L (MeV) | K_{sym} (MeV) |
|------------------|-------------------------------|----------------|--------------|--------------|--------------|---------------------------|
| NL3 | 0.149 | -16.2 | 271.6 | 37.4 | 118.9 | 101.6 |
| NL3 $\omega\rho$ | 0.148 | -16.2 | 271.6 | 31.7 | 55.5 | -7.6 |
| DDME2 | 0.152 | -16.1 | 250.9 | 32.3 | 51.2 | -87.1 |
| GM1 | 0.154 | -16.3 | 300.7 | 32.5 | 94.4 | 18.1 |
| TM1 | 0.146 | -16.3 | 281.2 | 36.9 | 111.2 | 33.8 |
| DDH δ | 0.153 | -16.3 | 240.3 | 25.6 | 48.6 | 91.4 |
| DD2 | 0.149 | -16.0 | 242.6 | 31.7 | 55.0 | -93.2 |
| BSR2 | 0.149 | -16.0 | 239.9 | 31.5 | 62.0 | -3.1 |
| BSR6 | 0.149 | -16.1 | 235.8 | 35.6 | 85.7 | -49.6 |

Results: DNN residuals



$$\sigma(\Delta K_{sym}) = 85.4 \text{ MeV}$$

Results: DNN residuals



$$\sigma(\Delta K_{sat}) = 40.19 \text{ MeV}$$

Conclusions

- DNN model maps the NS EOS to the nuclear matter properties
- Fast model to extract the nuclear model parameters from neutron star EOS
- Low order parameters show good accuracy (with real nuclear models)
- Future work: Increase the accuracy by mapping the isoscalar/isovector parts separately.

Uncertainties on nuclear parameters

J. Margueron, et. al. *Phys.Rev.C* 97 (2018) 2, 025805

| Model | Ref. | E_{sat} MeV | n_{sat} fm ⁻³ | K_{sat} MeV | E_{sym} MeV |
|-----------------------|-------------------|-------------------|-------------------------------|------------------|------------------|
| El. scatt. | Wang-99 [56] | | 0.1607 | 235 ±15 | |
| LDM | Myers-66 [57] | -15.677 | 0.136 [†] | 295 | 28.06 |
| LDM | Royer-08 [58] | -15.5704 | 0.133 [†] | | 23.45 |
| LSD | Pomorski-03 [59] | -15.492 | 0.142 [†] | | 28.82 |
| DM | Myers-77 [60] | -15.96 | 0.145 [†] | 240 | 36.8 |
| FRDM | Buchinger-01 [61] | | 0.157 ±0.004 | | |
| INM | Satpathy-99 [62] | -16.108 | 0.1620 | 288 ±20 | |
| DF-Skyrme | Tondeur-86 [63] | | 0.158 | | |
| DF-Skyrme | Klupfel-09 [64] | -15.91 ±0.06 | 0.1610 ±0.0013 | 222 ±8 | 30.7 ±1.4 |
| DF-BSK2 | Goriely-02 [65] | -15.79 | 0.1575 | 234 | 28.0 |
| DF-BSK24, 28,29 | Goriely-15 [66] | -16.045 ±0.005 | 0.1575 ±0.0004 | 245 | 30.0 |
| DF-Skyrme | McDonnell-15 [67] | -15.75 ±0.25 | 0.160 ±0.005 | 220 ±20 | 29 ±1 |
| DF-NLRMF | NL3* [68] | -16.3 | 0.15 | 258 | 38.7 |
| DF-NLRMF | PK [69] | -16.27 | 0.148 | 283 | 37.7 |
| DF-DDRMF | DDME1,2 [70, 71] | -16.17 ±0.03 | 0.152 ±0.00 | 247 ±3 | 32.7 ±0.4 |
| DF-DDRMF | PK [69] | -16.27 | 0.150 | 262 | 36.8 |
| present estimation | | -15.8 ±0.3 | 0.155 ±0.005 | 230 ±20 | 32 ±2 |

| Model | Ref. | Q_{sat} MeV | L_{sym} MeV | K_{sym} MeV | K_{τ} MeV |
|-----------------------|----------------------|------------------|------------------|------------------|-------------------|
| DF-Skyrme | Berdichevsky-88 [94] | 30 | 0 | | |
| DF-Skyrme | Farine-97 [95] | -700 ±500 | | | |
| DF-Skyrme | Alam-14 [31] | -344 ±46 | 65 ±14 | -23 ±73 | -322 ±34 |
| DF-Skyrme | McDonnell-15 [67] | | 40 ±20 | | |
| DF-NLRMF | NL3* [68] | 124 | 123 | 106 | -690 |
| DF-NLRMF | PK [69] | -25 | 116 | 55 | -630 |
| DF-DDRMF | DDME1,2 [70, 71] | 400 ±80 | 53 ±3 | -94 ±7 | -500 ±7 |
| DF-DDRMF | PK [69] | -119 | 79.5 | -50 | -491 |
| Correlation | Centelles-09 [96] | | 70 ±40 | | -425 ±175 |
| DF-RPA | Carbone-10 [85] | | 60 ±30 | | |
| Correlation | Danielewicz-14 [87] | | 53 ±20 | | |
| Correlation | Newton-14 [97] | | 70 ±40 | | |
| Correlation | Lattimer-14 [98] | | 53 ±20 | | |
| GMR | Sagawa-07 [99] | | | | -500 ±50 |
| GMR | Patel-14 [100] | | | | -550 ±100 |
| present estimation | | 300 ±400 | 60 ±15 | -100 ±100 | -400 ±100 |