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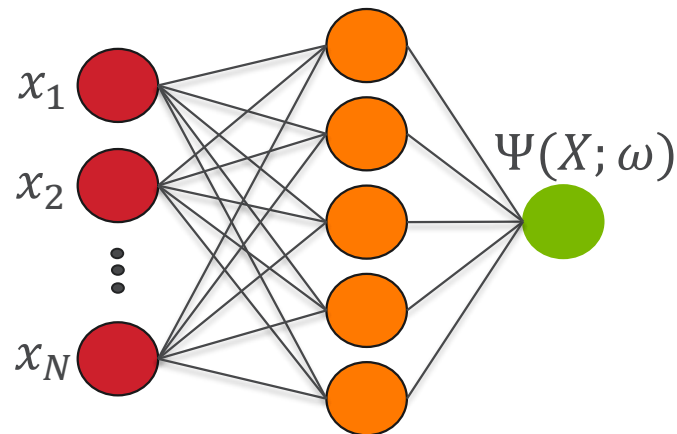
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References: This code of conduct is based heavily on that of the [INT](#) and the [APS](#). We are also grateful to Roxanne Springer for valuable discussion and guidance.

MODELING LOW-DENSITY NUCLEAR MATTER WITH NEURAL-NETWORK QUANTUM STATES

BRYCE FORE



OUTLINE

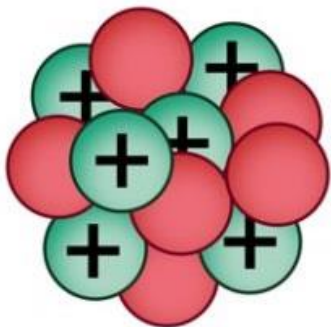
- Nuclear many body problem
 - Many body methods
 - Curse of dimensionality
- Overview of variational Monte Carlo (VMC)
 - Metropolis-Hastings Sampling
 - Parameter optimization
 - Fermionic trial wavefunctions
- Results
 - Nuclei
 - Pure neutron matter
 - Symmetric nuclear matter
 - Self emergence of clustering

THE NUCLEAR MANY-BODY PROBLEM

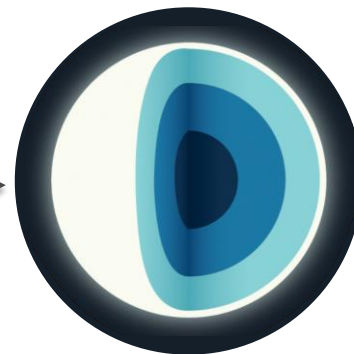
Many-body Schrödinger equation

$$\left(-\sum_i \frac{\nabla_i^2}{2m_N} + V \right) |\Psi_0\rangle = E_0 |\Psi_0\rangle$$

Nuclei



Neutron stars

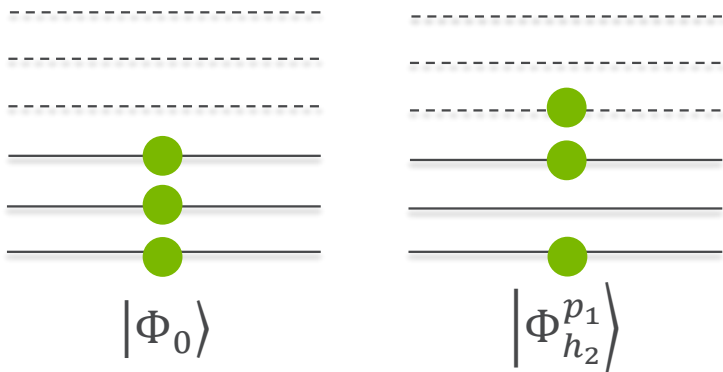


THE NUCLEAR MANY BODY METHODS

Configuration-interaction

$$|\Psi_0\rangle = \sum_{h_1, \dots, p_1, \dots} c_{h_1 \dots}^{p_1 \dots} |\Phi_{h_1 \dots}^{p_1 \dots}\rangle$$

$$|\Phi_{h_1 \dots}^{p_1 \dots}\rangle = a_{p_1}^\dagger \dots a_{h_1} \dots |\Phi_0\rangle$$



Quantum Monte Carlo

$$|\Psi_T\rangle = \sum_n c_n |\Psi_n\rangle$$

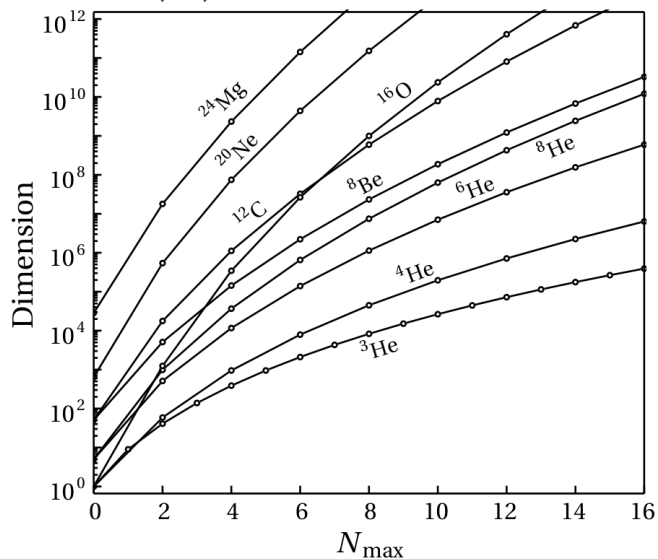
$$H|\Psi_n\rangle = E_n |\Psi_n\rangle$$

$$\lim_{\tau \rightarrow \infty} e^{-(H-E_0)\tau} |\Psi_T\rangle = c_0 |\Psi_0\rangle$$

CURSE OF DIMENSIONALITY

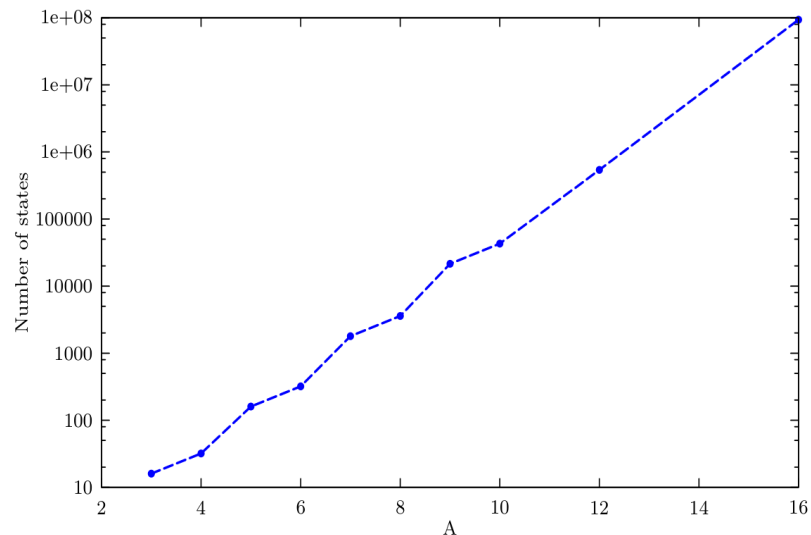
Configuration-interaction

$$\binom{N}{A} = \frac{N!}{(N-A)! A!}$$



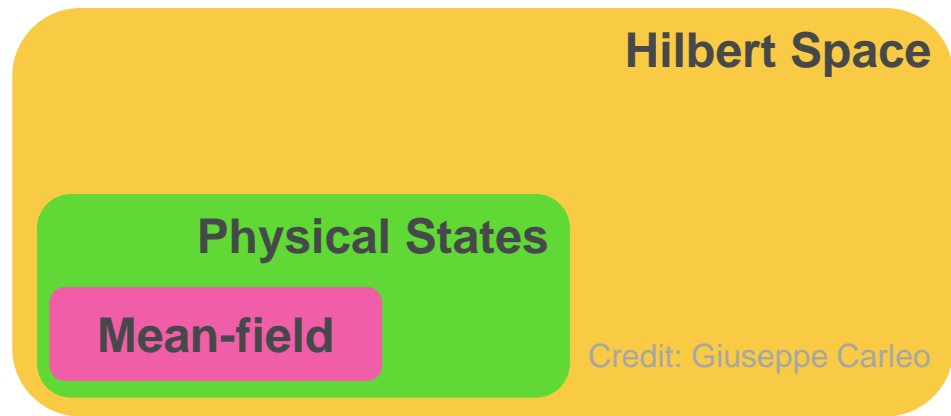
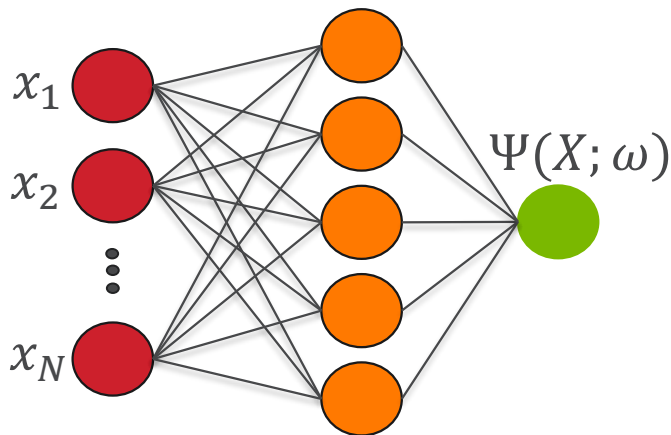
Green's function Monte Carlo

$$\lim_{\tau \rightarrow \infty} e^{-(H-E_0)\tau} |\Psi_T\rangle = c_0 |\Psi_0\rangle$$



NEURAL NETWORK QUANTUM STATES

- Artificial neural networks compactly represent complex high-dimensional functions
- Most quantum states of interest have distinctive features and intrinsic structures



Credit: Giuseppe Carleo

SCALING AND COMPUTATIONAL PERFORMANCE

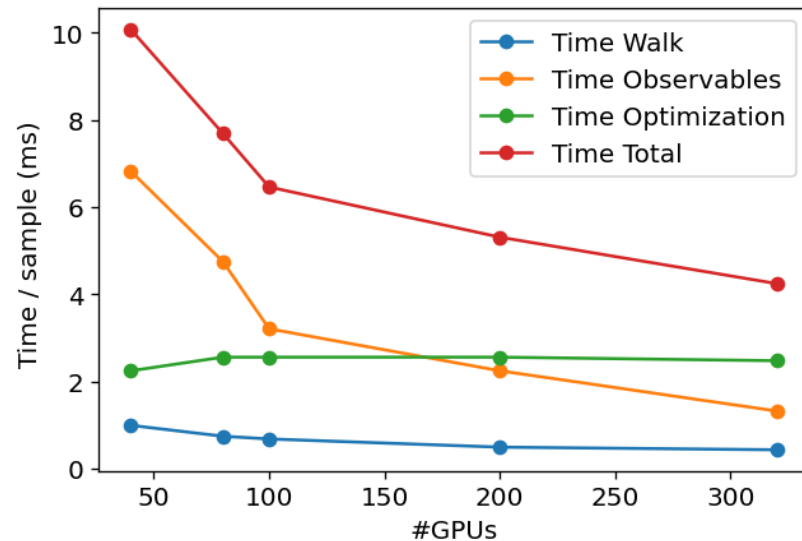
Scaling with system size

Conventional QMC: $O(2^A)$

Neural quantum states: $O(A^5)$

A = Number of particles in system

Scaling with resources



PIONLESS EFT HAMILTONIAN

- Pionless-EFT Hamiltonian

$$H_{LO} = - \sum_i \frac{\vec{\nabla}_i^2}{2m_N} + \sum_{i<j} v_{ij} + \sum_{i<j<k} V_{ijk}$$

- Two body operators including spin and isospin dependence

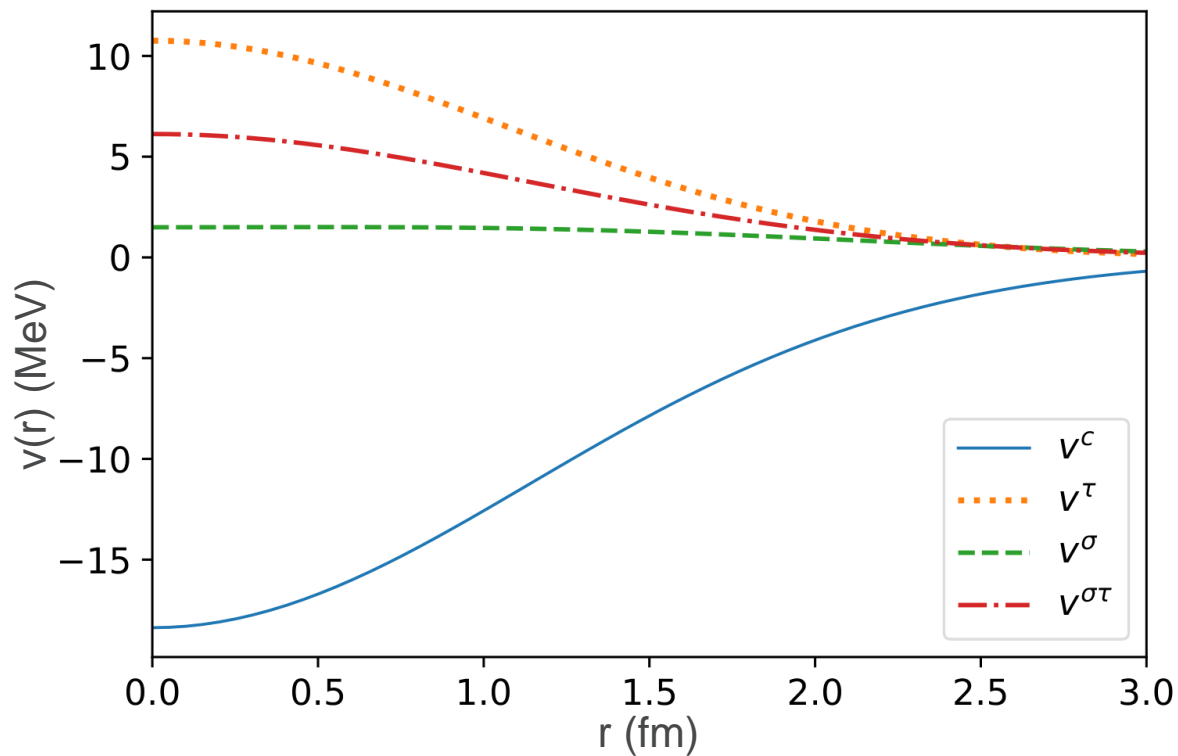
$$v_{ij}^{\text{CI}} = \sum_{p=1}^4 v^p(r_{ij}) O_{ij}^p$$

$$O_{ij}^{p=1,4} = (1, \tau_{ij}, \sigma_{ij}, \sigma_{ij} \tau_{ij})$$

$$\sigma_{ij} = \boldsymbol{\sigma}_i \cdot \boldsymbol{\sigma}_j, \quad \tau_{ij} = \boldsymbol{\tau}_i \cdot \boldsymbol{\tau}_j$$

R. Schiavilla, PRC 103, 054003(2021)

PIONLESS EFT HAMILTONIAN



VARIATIONAL MONTE CARLO (VMC)

1. Specify a parameterized function to act as the trial wavefunction

$$\Psi_T(R, S; \omega) = e^{U(R, S; \omega)} \Phi(R, S; \omega)$$

2. Use Metropolis-Hastings algorithm to sample trial wavefunction

$$\frac{\langle \Psi_T | O | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle} \approx \frac{1}{N_{conf}} \sum_{\{R, S\}} O_L(R, S)$$

3. Optimize parameters of trial wavefunction to obtain lower energy

$$E_0 \leq E_T = \frac{\langle \Psi_T | \hat{H} | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle}$$

METROPOLIS-HASTINGS SAMPLING

Sampling algorithm:

- Randomly sample coordinates, R' , and spins, S'

$$P_R = \frac{|\Psi_T(R', S)|^2}{|\Psi_T(R, S)|^2} \quad P_S = \frac{|\Psi_T(R, S')|^2}{|\Psi_T(R, S)|^2}$$

- If P is greater than uniform random variable from 0 to 1, accept new values
- Observables are estimated by taking averages over sampled configurations

$$\frac{\langle \Psi_T | O | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle} = \frac{\sum_S \int dR |\Psi_T(R, S)|^2 O_L(R, S)}{\sum_S \int dR |\Psi_T(R, S)|^2} \approx \frac{1}{N_{conf}} \sum_{\{R,S\}} O_L(R, S)$$

$$O_L = \frac{\langle RS | O | \Psi_T \rangle}{\langle RS | \Psi_T \rangle}$$

STOCHASTIC RECONFIGURATION

Improve trial wavefunction by minimizing energy expectation value

$$E_0 \leq E_T = \frac{\langle \Psi_T | \hat{H} | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle}$$

Gradient of energy ($G_i = \frac{dE_T}{d\omega_i}$), supplemented by Quantum Fisher Information S_{ij}

$$G_i = 2 \left(\frac{\langle \partial_i \Psi_T | \hat{H} | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle} - E_T \frac{\langle \partial_i \Psi_T | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle} \right); \quad S_{ij} = \frac{\langle \partial_i \Psi_T | \partial_j \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle} - \frac{\langle \partial_i \Psi_T | \Psi_T \rangle \langle \Psi_T | \partial_j \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle \langle \Psi_T | \Psi_T \rangle}$$

Parameters at step s are updated as

$$\omega^{s+1} = \omega^s - \eta(S + \Lambda)^{-1}G$$

FERMIONIC TRIAL WAVEFUNCTIONS

$$\Psi_T(X) = e^{U(X)} \Phi(X)$$

$$\Psi(\dots, x_i, \dots x_j \dots) = -\Psi(\dots, x_j, \dots x_i \dots)$$

$$U(\dots, x_i, \dots x_j \dots) = U(\dots, x_j, \dots x_i \dots)$$

$$\Phi(\dots, x_i, \dots x_j \dots) = -\Phi(\dots, x_j, \dots x_i \dots)$$

- Build in fermion antisymmetry for network compactness
- Permutation-invariant Jastrow function improves ansatz flexibility
- Build U and Φ functions from fully connected, deep neural networks

DEEP SET ARCHITECTURE

- Generic function independent of particle ordering

$$U(\dots, x_i, \dots x_j \dots) = U(\dots, x_j, \dots x_i \dots)$$

- Map configuration for each particle to a latent space, sum results, map to real numbers

$$U(X) = \rho \left(\sum_i \vec{\phi}(x_i) \right) \quad \begin{array}{l} \vec{\phi}: \mathbb{R}^5 \rightarrow \mathbb{R}^{latent} \\ \rho: \mathbb{R}^{latent} \rightarrow \mathbb{R} \end{array}$$

- $\vec{\phi}$ and ρ are represented by neural networks

NEURAL SLATER-JASTROW ANSATZ

- Use Slater determinant to enforce antisymmetry
- Single particle wavefunctions represented by neural networks

$$\Psi_T(X) = e^{U(X)} \Phi(X)$$

$$\Phi(X) = \begin{vmatrix} \phi_1(x_1) & \phi_1(x_2) & \dots & \phi_1(x_n) \\ \phi_2(x_1) & & & \vdots \\ \vdots & & & \\ \phi_n(x_1) & \dots & & \phi_n(x_n) \end{vmatrix}$$

NEURAL PFAFFIAN ANSATZ

$$\Psi_T(X) = e^{U(X^*)} \Phi_{pf}(X^*)$$

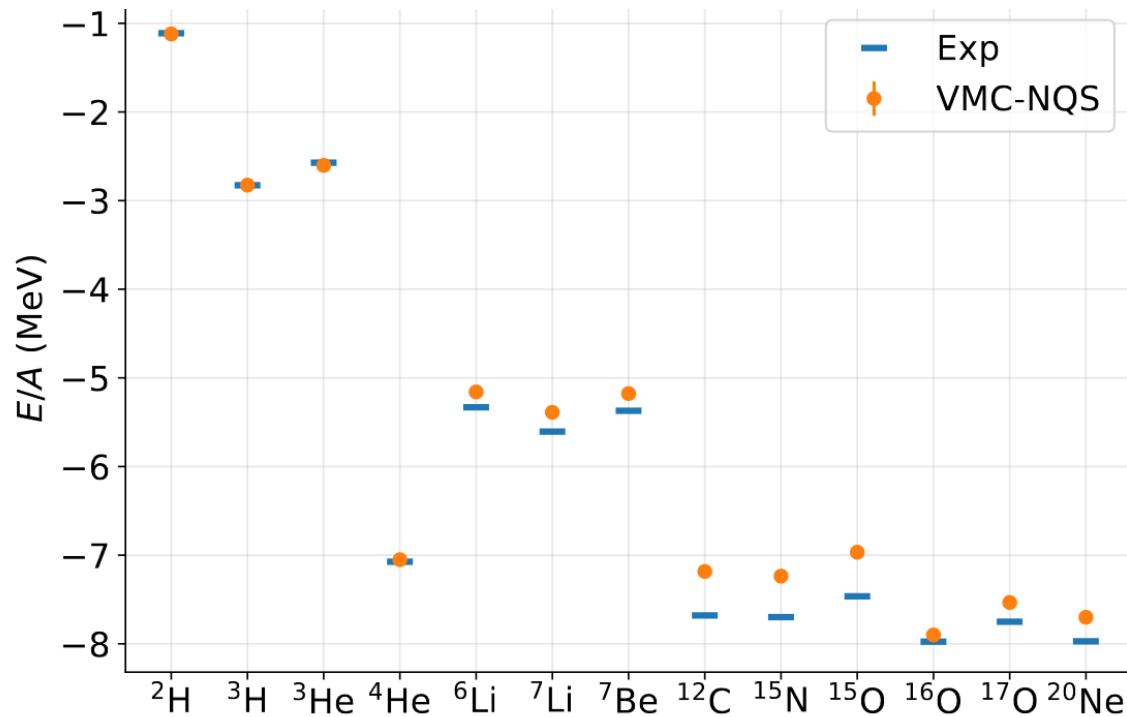
$$\Phi_{pf}(X) = pf[M]$$

$$M_{ij} = \phi(x_i, x_j) - \phi(x_j, x_i)$$

- Input, X , with backflow preprocessing gives X^*
- Slater determinant \rightarrow Pfaffian
- M must be skew symmetric, $A = -A^T$, and square with even size
- Built in pairwise structure
- Pfaffian requires only one neural net, ϕ , so uses far fewer parameters

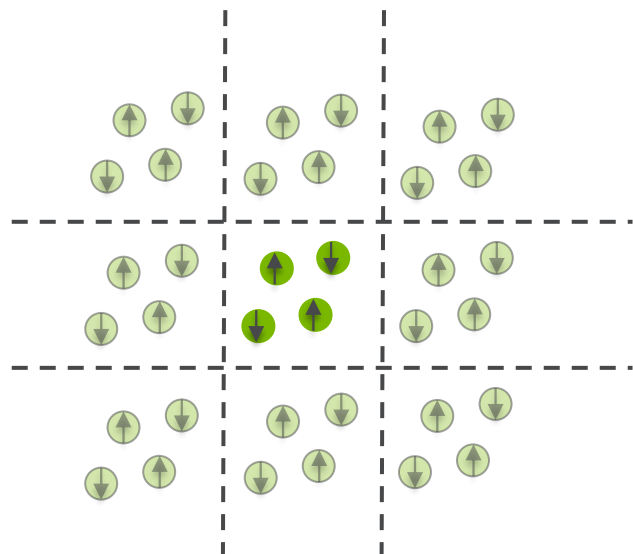
J. Kim, *Commun Phys* **7**, 148 (2024)

NEURAL QUANTUM STATE RESULTS IN NUCLEI



A. Gnech, Phys. Rev. Lett. 133, 142501

BULK NUCLEAR MATTER SETUP

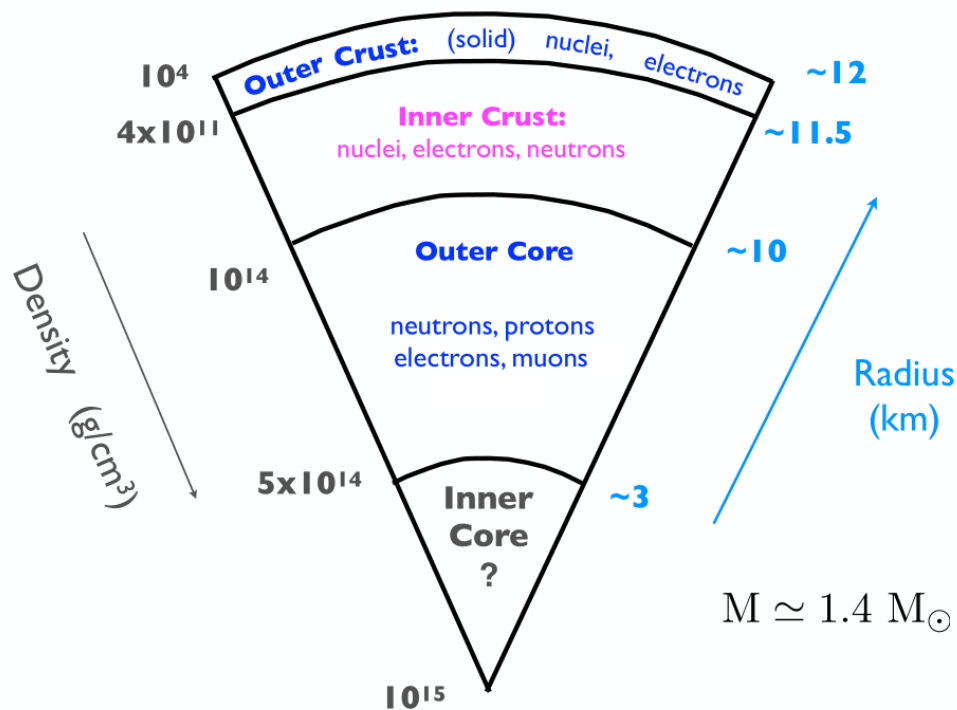


- Periodic boundary conditions and coordinate system

$$\mathbf{r}_i \longrightarrow \tilde{\mathbf{r}}_i = \left\{ \sin \left(\frac{2\pi}{L} \mathbf{r}_i \right), \cos \left(\frac{2\pi}{L} \mathbf{r}_i \right) \right\}$$

- Potential energy contribution from particle images
- Remove Coulomb potential

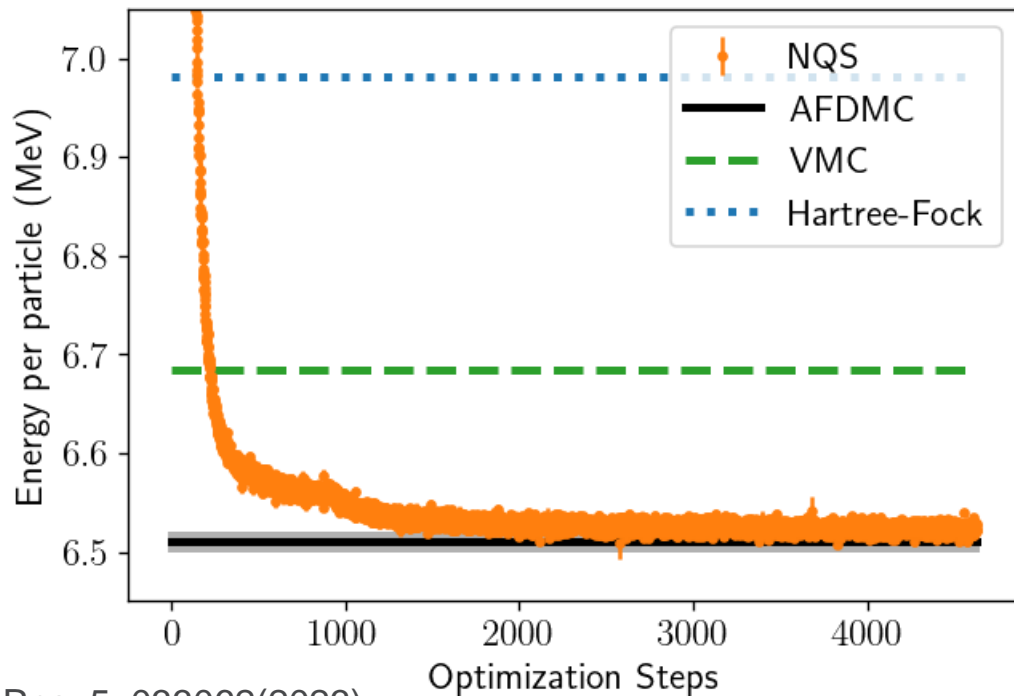
NEUTRON STAR STRUCTURE



- Mostly neutrons but composition varies with density
- Nuclei in crust are squeezed into uniform matter in core
- Likely neutron superfluid in inner crust and outer core
- Calculations currently focus on inner crust

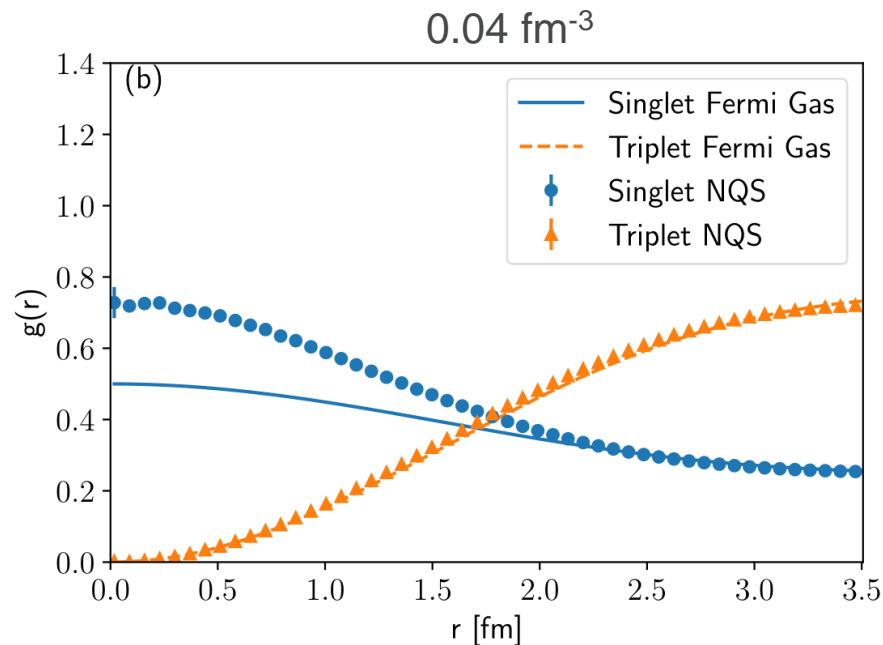
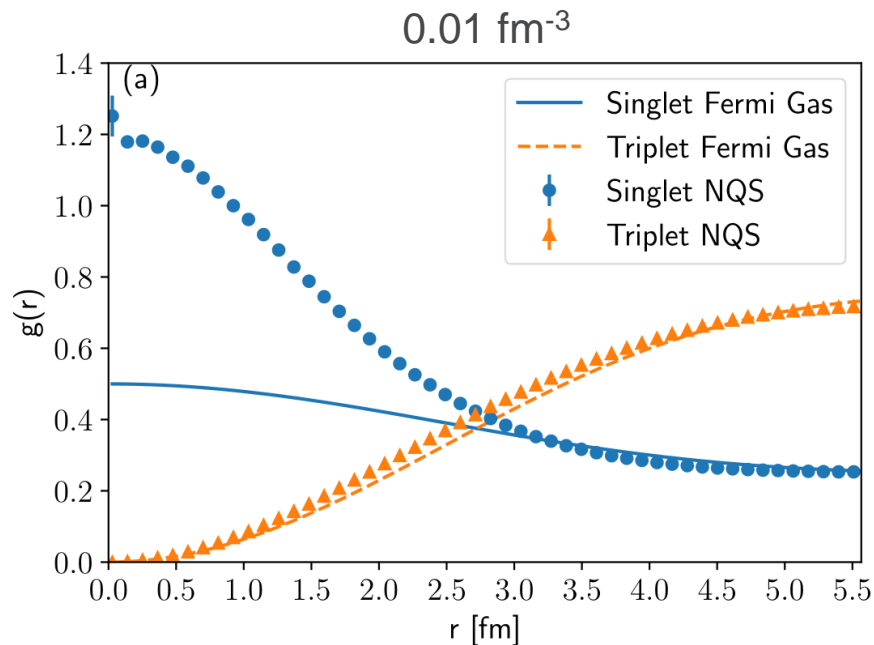
PURE NEUTRON MATTER

Neutron Matter (14 particles, 0.04 fm^{-3})



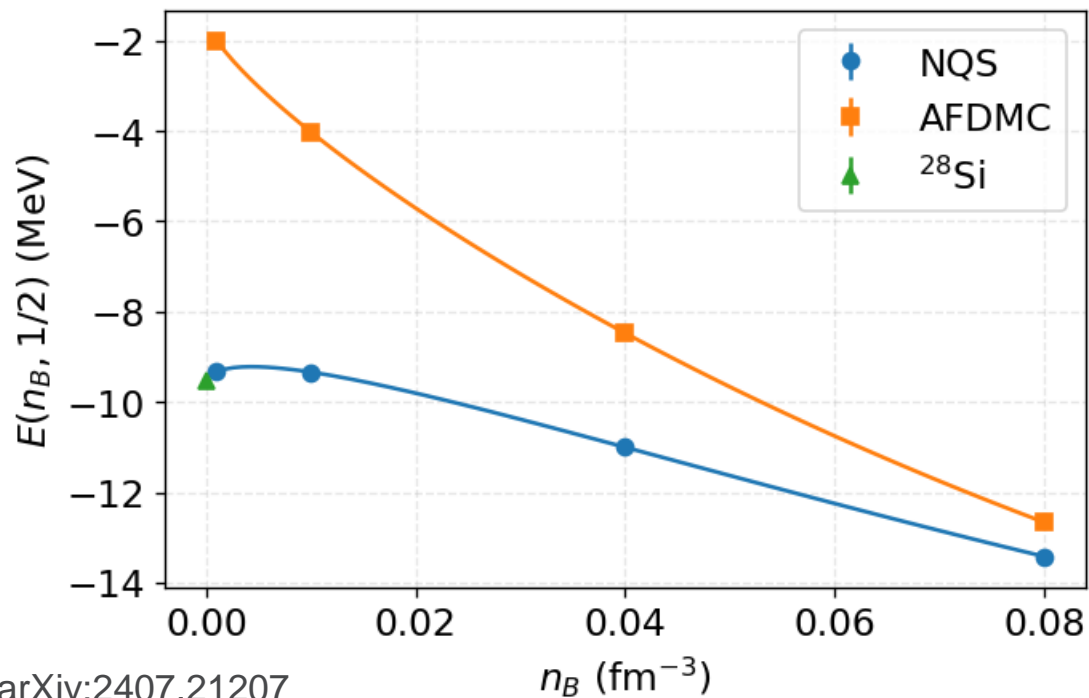
B. Fore, Phys. Rev. Res. 5, 033062(2023)

TWO-BODY PAIR DISTRIBUTIONS



B. Fore, Phys. Rev. Res. 5, 033062(2023)

SYMMETRIC NUCLEAR MATTER

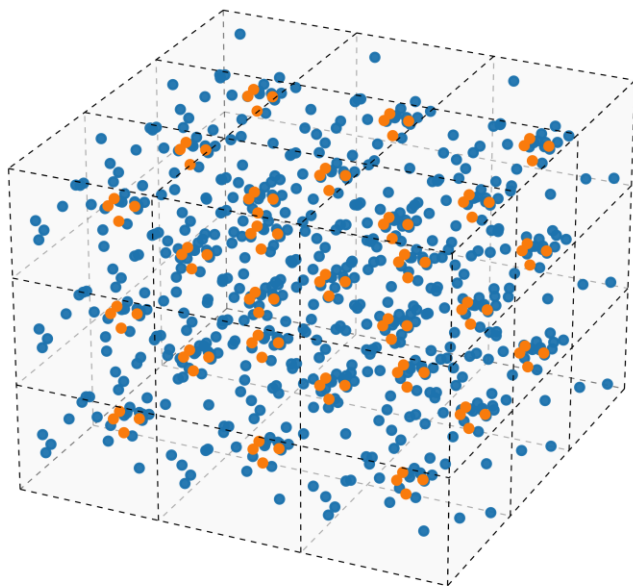


B. Fore, arXiv:2407.21207

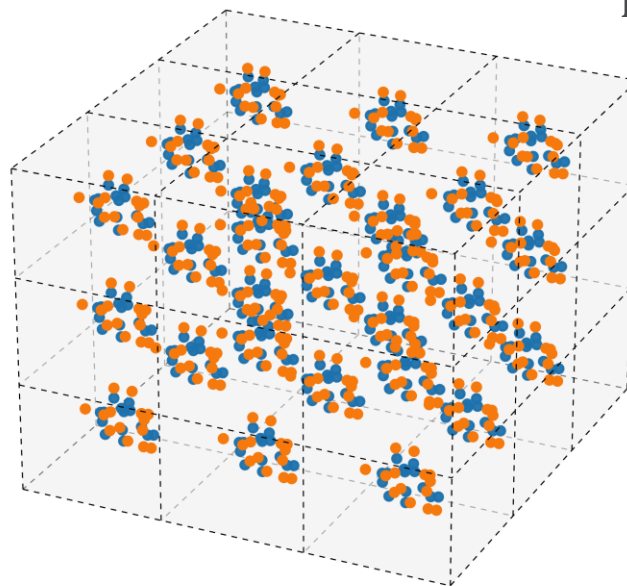
SELF EMERGING CLUSTERING

Density: 0.01 fm^{-3}

● Proton
● Neutron



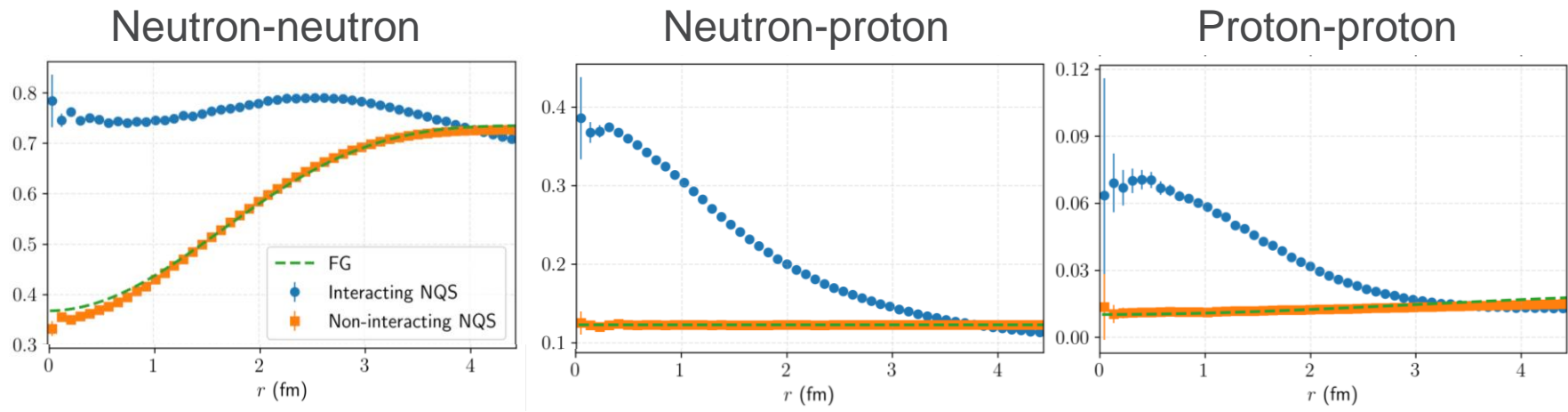
4 protons; 24 neutrons



14 protons; 14 neutrons

B. Fore, arXiv:2407.21207

CLUSTERING: TWO-BODY PAIR DISTRIBUTIONS



B. Fore, arXiv:2407.21207

ASYMMETRIC MATTER ENERGY

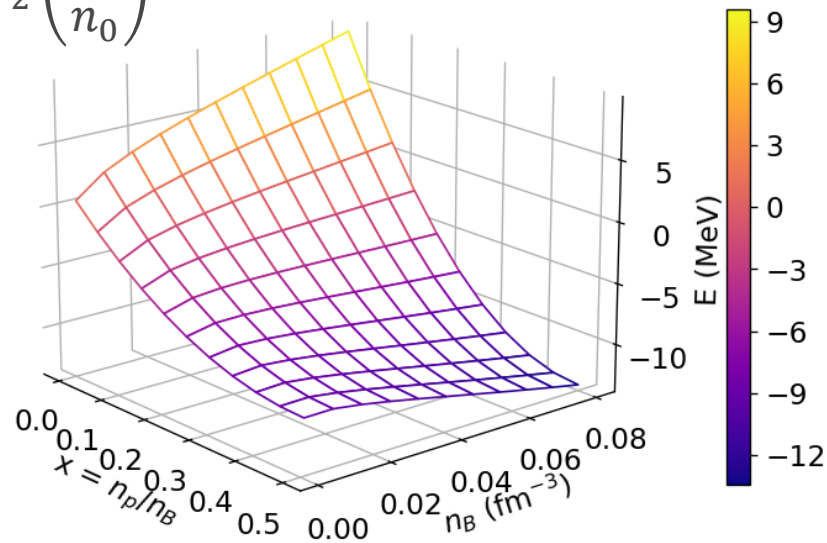
- Fit $E(n_B, 0)$ and $E(n_B, 1/2)$

$$E(n_B, x) = a_0 + a_{2/3} \left(\frac{n_B}{n_0} \right)^{2/3} + a_1 \left(\frac{n_B}{n_0} \right) + a_2 \left(\frac{n_B}{n_0} \right)^2$$

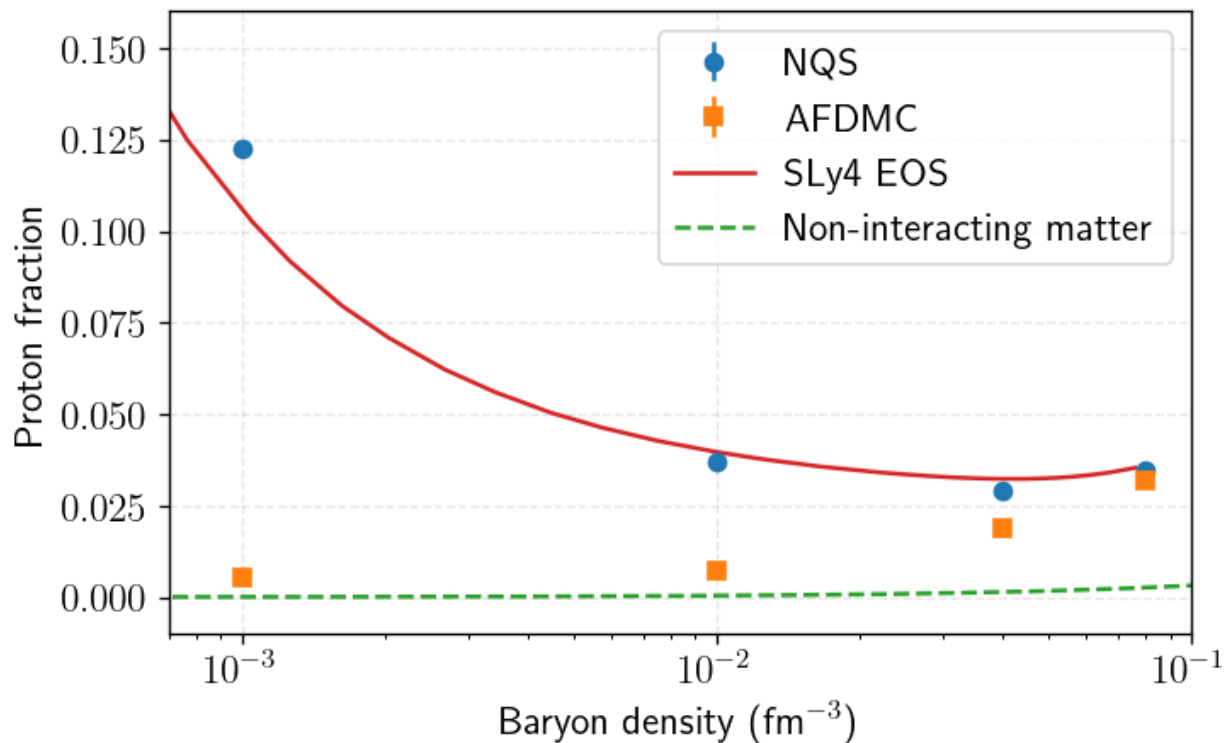
- Model energy by symmetry energy expansion

$$S(n_B) = E(n_B, 0) - E(n_B, 1/2)$$

$$E(n_B, x) = E(n_B, 1/2) + (1 - 2x)^2 S(n_B)$$



NUCLEAR MATTER PROTON FRACTION



Assumptions:

- Charge neutrality

$$n_p = n_e$$

- Beta equilibrium

$$\mu_e = \mu_n - \mu_p$$

B. Fore, arXiv:2407.21207

CONCLUSIONS AND NEXT STEPS

- Conclusions:
 - Favorable scaling with number of fermions
 - Universal and accurate approximations for fermion wavefunctions
 - Scaling to leadership-class computers
 - NQS can model a variety of phases of nuclear matter material
- Next steps:
 - Improved nuclear potential including tensor term
 - Expanding to larger nuclei and larger periodic systems to avoid finite size effects

COLLABORATORS



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THANK YOU