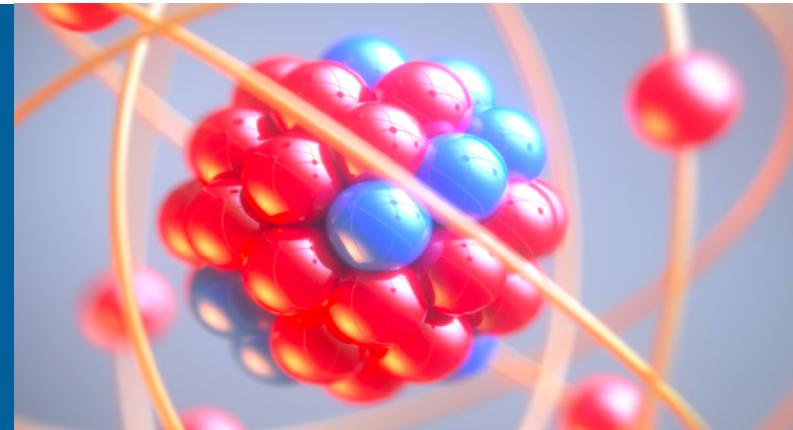


# VARIATIONAL LEARNING QUANTUM WAVE FUNCTIONS



ALESSANDRO LOVATO

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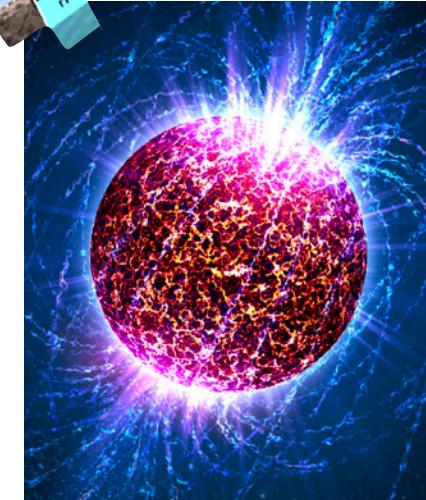
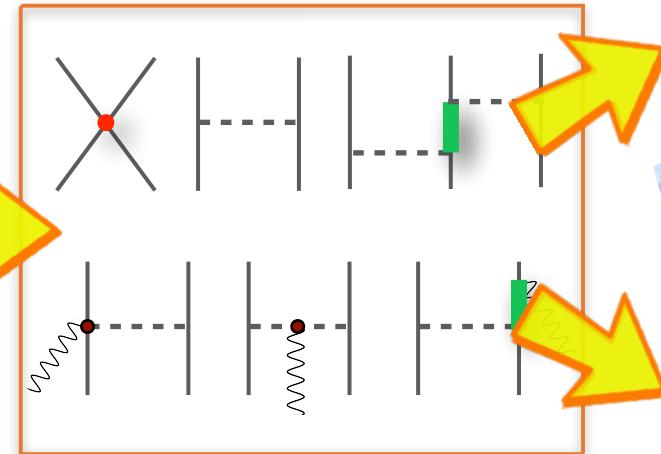


M. Rigo



A. Di Donna,  
F. Pederiva

# “AB-INITIO” NUCLEAR THEORY



# CONFIGURATION-INTERACTION METHODS

$$\Psi_0(x_1, \dots, x_A) = \sum_n c_n \Phi_n(x_1, \dots, x_A)$$

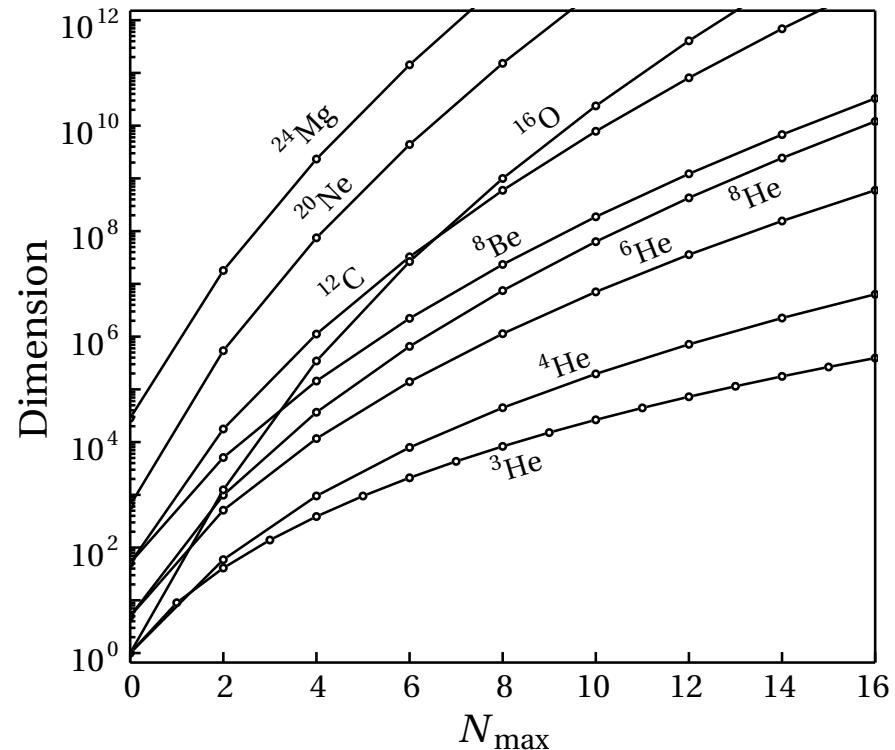
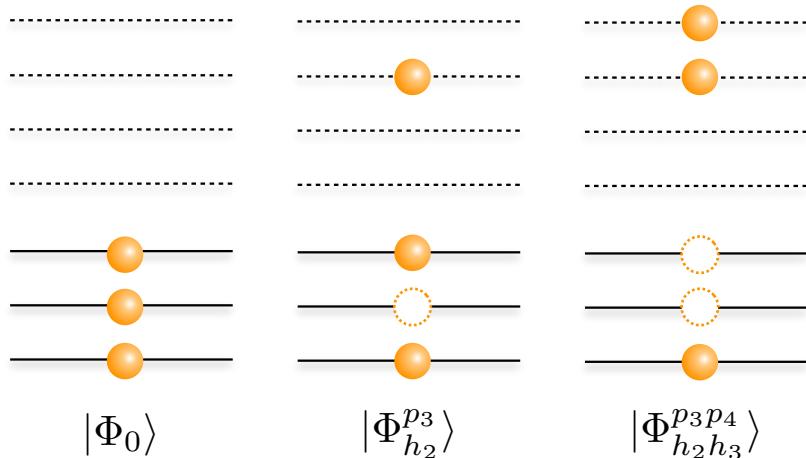
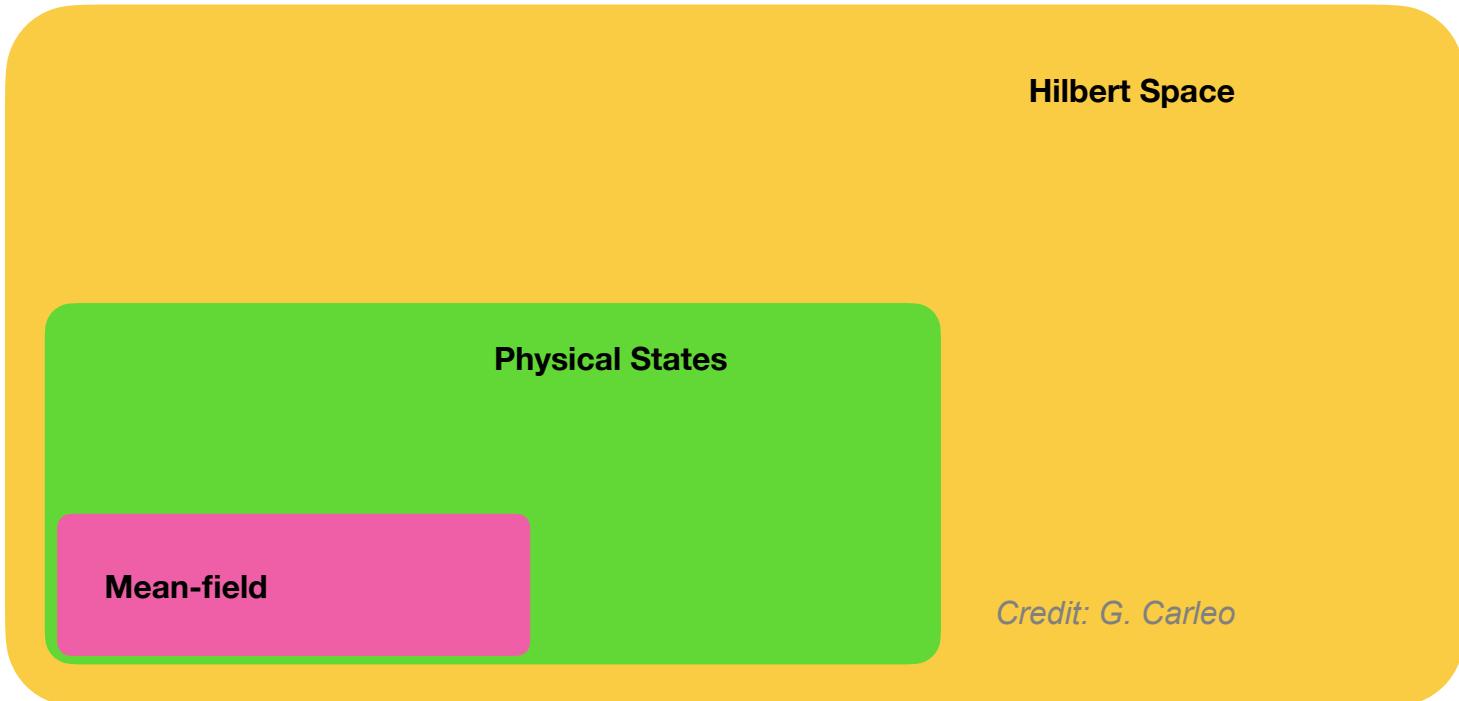


Image courtesy of Patrick Fasano

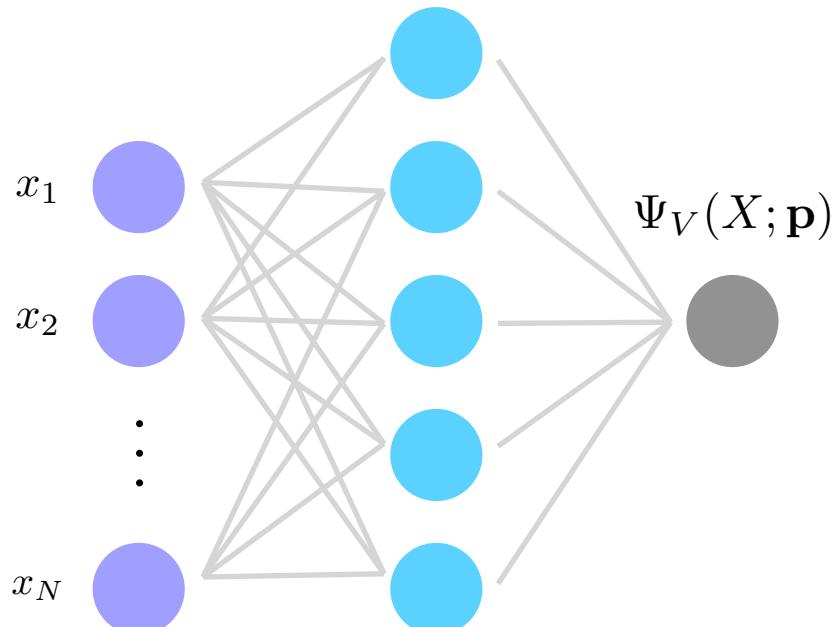
# NEURAL-NETWORK QUANTUM STATES

Quantum states of physical interest have distinctive features and intrinsic structures



# NEURAL-NETWORK QUANTUM STATES

NQS are now widely and successfully applied to study condensed-matter systems



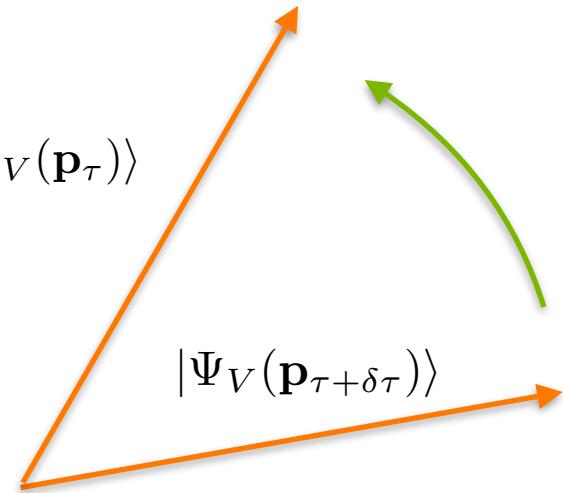
$$E_V \equiv \frac{\langle \Psi_V | H | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle} > E_0$$

$$E_V \simeq \frac{1}{N} \sum_{X \in |\Psi_V(X)|^2} \frac{\langle X | H | \Psi_V \rangle}{\langle X | \Psi_V \rangle}$$

# WAVE FUNCTION OPTIMIZATION

ANN trained by performing an imaginary-time evolution in the variational manifold

$$\left\{ \begin{array}{l} |\bar{\Psi}_V(\mathbf{p}_\tau)\rangle \equiv (1 - H\delta\tau)|\Psi_V(\mathbf{p}_\tau)\rangle \\ \mathbf{p}_{\tau+\delta\tau} = \arg \max_{\mathbf{p} \in R^d} \left( |\langle \bar{\Psi}_V(\mathbf{p}_\tau) | \Psi_V(\mathbf{p}_{\tau+\delta\tau}) \rangle|^2 \right) \end{array} \right.$$



The parameters are updated as

$$\mathbf{p}_{\tau+\delta\tau} = \mathbf{p}_\tau - \delta\tau S^{-1} \mathbf{g}_\tau$$

J. Stokes, *et al.*, *Quantum* 4, 269 (2020).

S. Sorella, *Phys. Rev. B* 64, 024512 (2001)

# NEURAL-NETWORK QUANTUM STATES

Nucleons are fermions

$$\Psi_V(x_1, \dots, x_i, \dots, x_j, \dots, x_A) = -\Psi_V(x_1, \dots, x_j, \dots, x_i, \dots, x_A)$$

Slater-Jastrow ansatz

$$\Psi_V(X) = e^{J(X)} \Phi(X) \quad ; \quad \Phi(X) = \det \begin{bmatrix} \phi_1(\mathbf{x}_1) & \phi_1(\mathbf{x}_2) & \cdots & \phi_1(\mathbf{x}_N) \\ \phi_2(\mathbf{x}_1) & \phi_2(\mathbf{x}_2) & \cdots & \phi_2(\mathbf{x}_N) \\ \vdots & \vdots & \ddots & \vdots \\ \phi_N(\mathbf{x}_1) & \phi_N(\mathbf{x}_2) & \cdots & \phi_N(\mathbf{x}_N) \end{bmatrix}$$

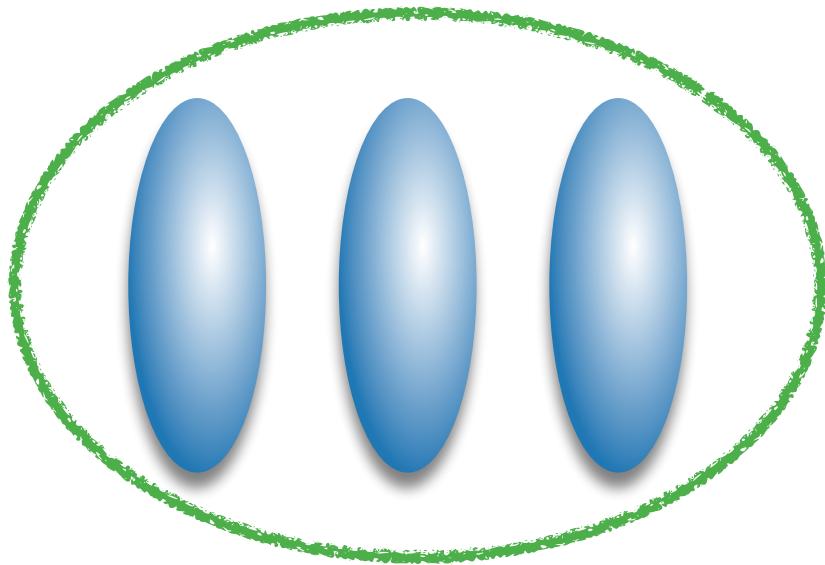
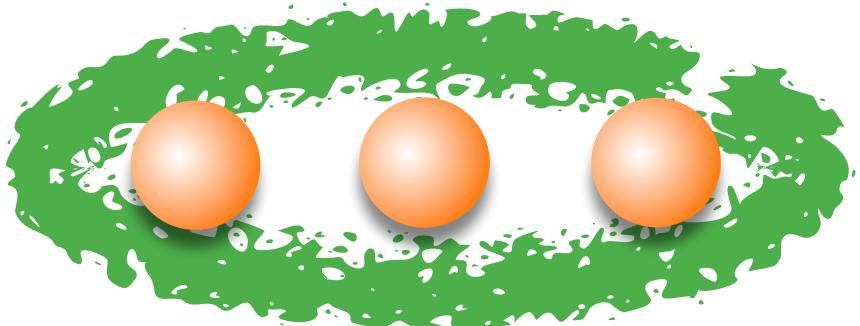
J. Stokes et al., PLB, 102, 205122 (2020)

Pfau et al., PRR 2, 033429 (2020)

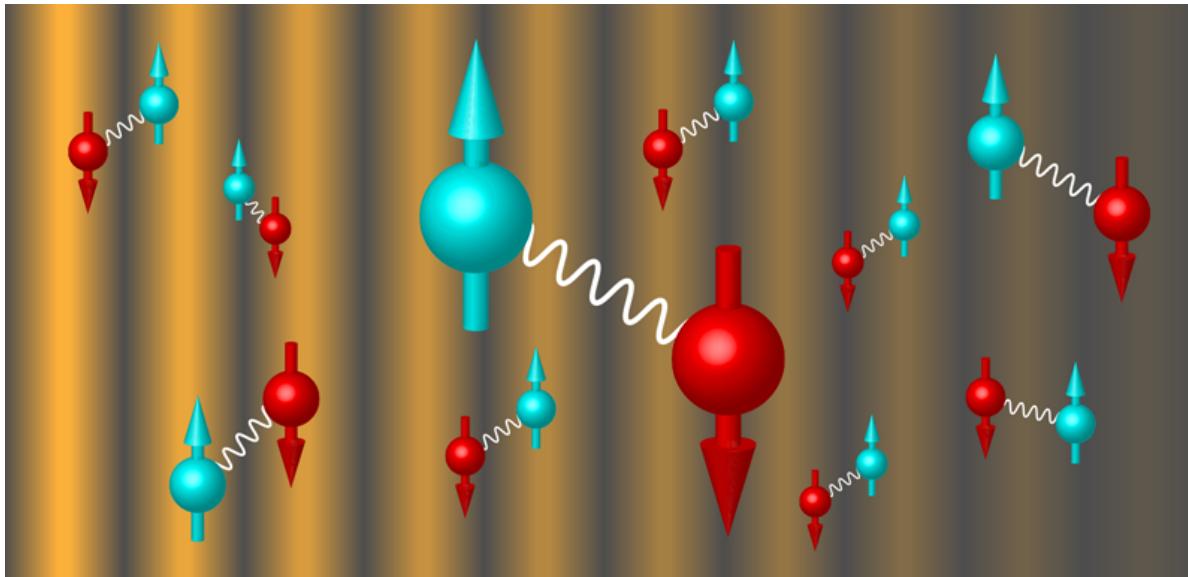
Hermann et al., Nature Chemistry, 12, 891 (2020)

# NEURAL BACKFLOW CORRELATIONS

The nodal structure is improved with neural back-flow transformations  $\mathbf{x}_i \xrightarrow{\text{orange arrow}} \phi(\mathbf{x}_i; \mathbf{x}_{j \neq i})$



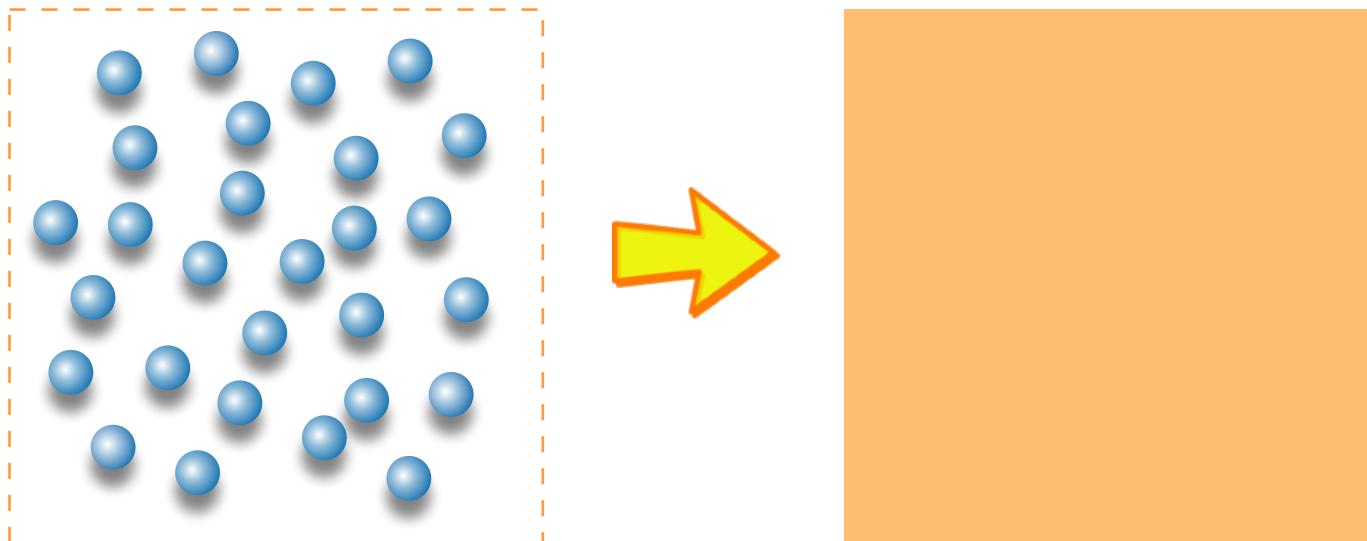
# CONDENSED-MATTER DETOUR



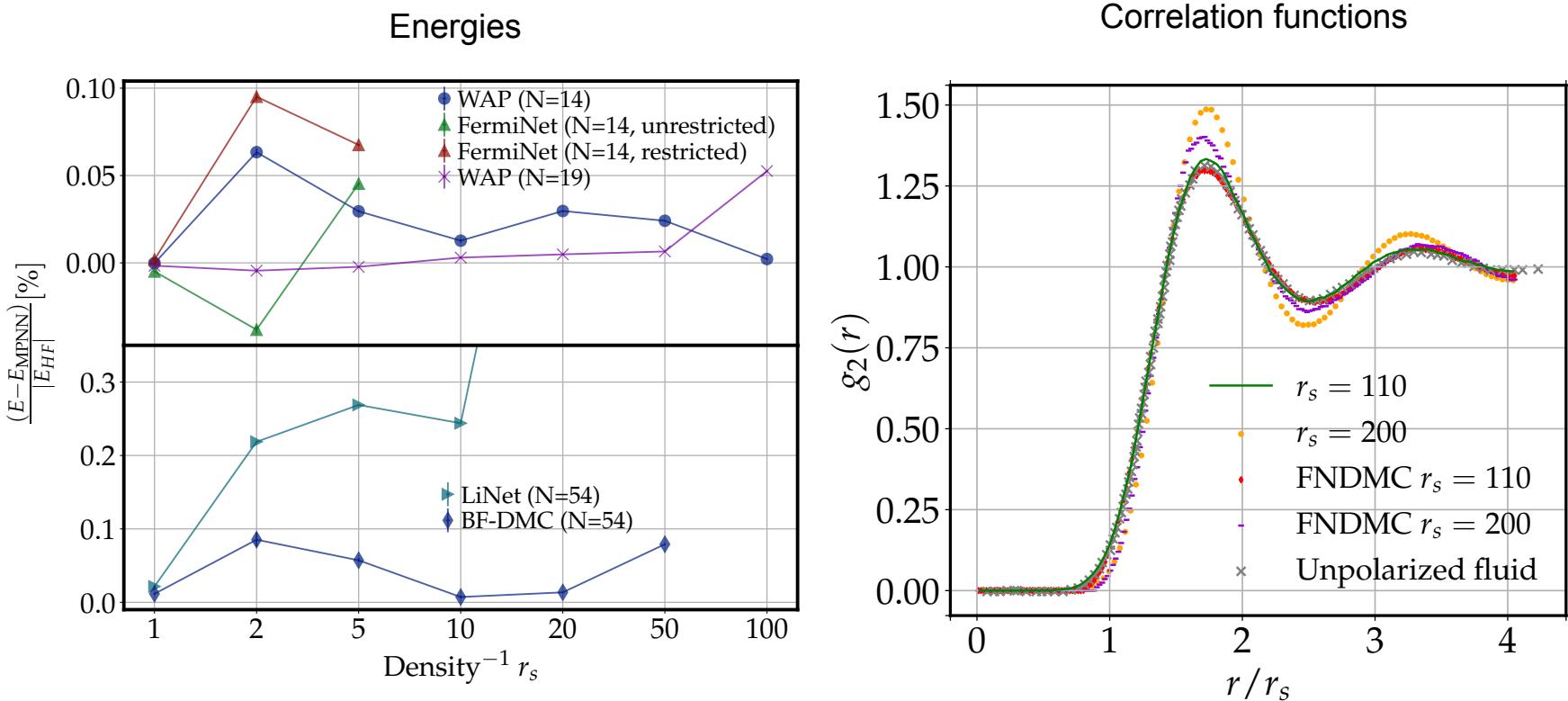
# HOMOGENEOUS ELECTRON GAS

We develop translation invariant NQS to study the Homogeneous Electron Gas.

$$H = -\frac{1}{2r_s^2} \sum_i^N \nabla_{\vec{r}_i}^2 + \frac{1}{r_s} \sum_{i < j}^N \frac{1}{\|\vec{r}_i - \vec{r}_j\|} + \text{const.}$$

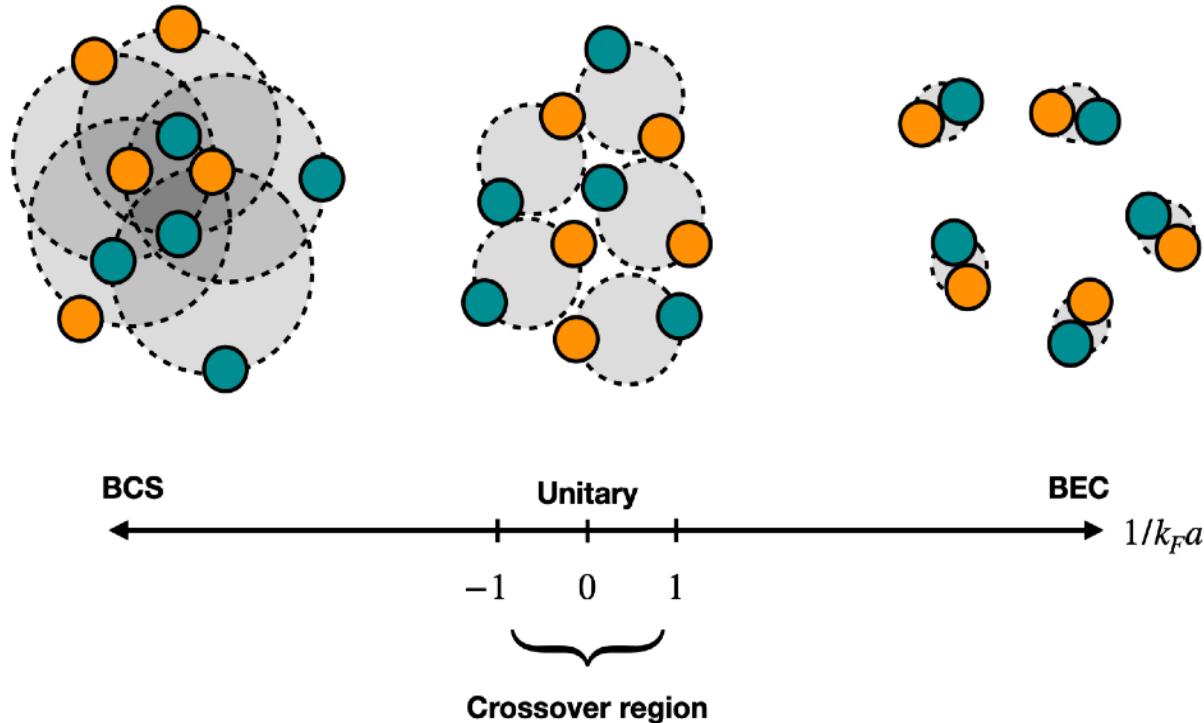


# HOMOGENEOUS ELECTRON GAS



# COLD FERMI GASES

Periodic-NQS to solve the two-components Fermi gas in the BCS- BEC crossover region



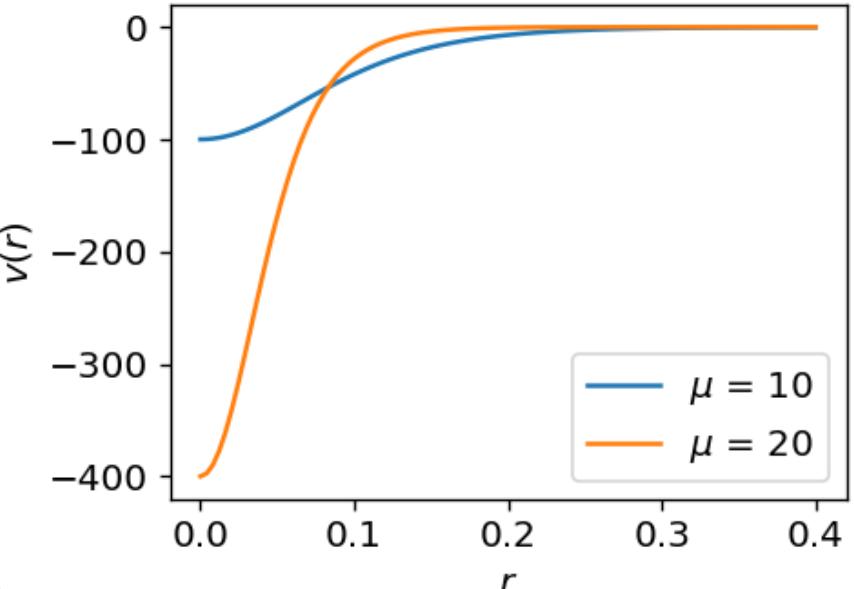
# COLD FERMI GASES

We model the 3D unpolarized gas of fermions with the Hamiltonian

$$H = \sum_i \frac{p_i^2}{2m} + \sum_{i < j} v_{ij}.$$

- Modified Pöschl-Teller potential between opposite-spin particles

$$v_{ij} = (\delta_{s_i, s_j} - 1) v_0 \frac{2\hbar^2}{m} \frac{\mu^2}{\cosh^2(\mu r_{ij})}$$



# COLD FERMI GASES

We introduce a Pfaffian-Jastrow ansatz

$$\Phi_{PJ}(X) = \text{pf} \begin{bmatrix} 0 & \phi(x_1, x_2) & \cdots & \phi(x_1, x_N) \\ \phi(x_2, x_1) & 0 & \cdots & \phi(x_2, x_N) \\ \vdots & \vdots & \ddots & \vdots \\ \phi(x_N, x_1) & \phi(x_N, x_2) & \cdots & 0 \end{bmatrix}$$

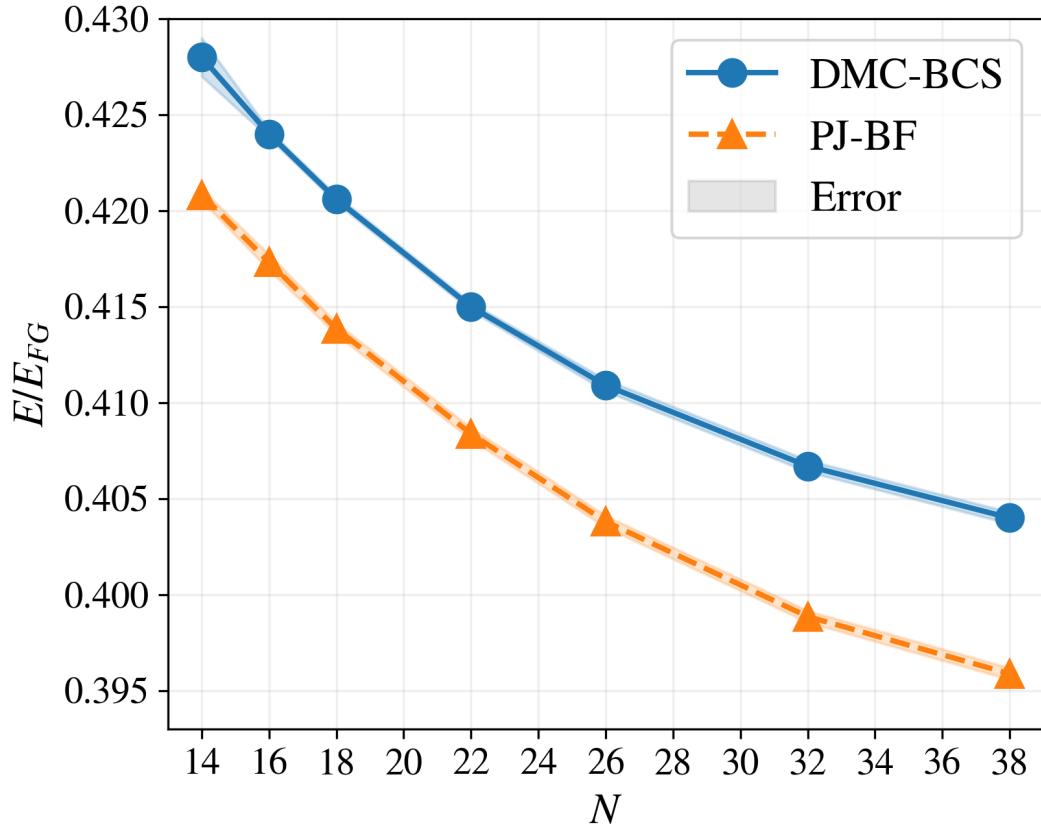
In order for the above matrix to be skew-symmetric, the neural pairing orbitals are taken to be

$$\phi(x_i, x_j) = \eta(x_i, x_j) - \eta(x_j, x_i)$$

Example:

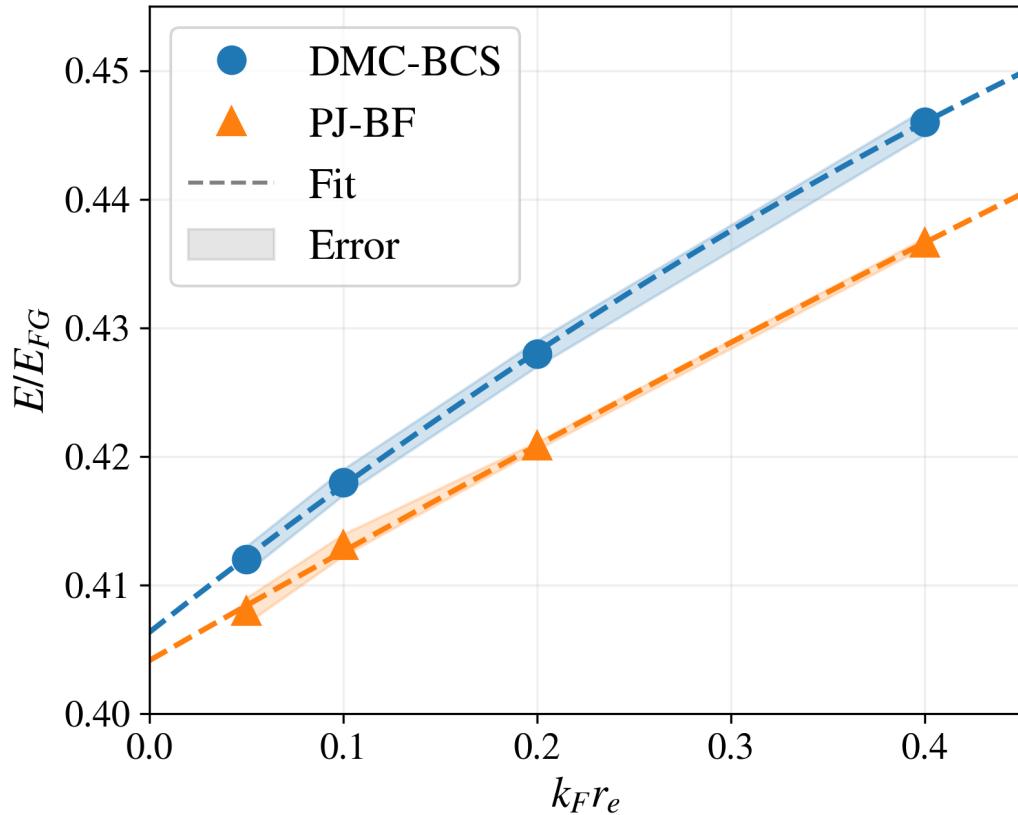
$$\text{pf} \begin{bmatrix} 0 & \phi_{12} & \phi_{13} & \phi_{14} \\ -\phi_{12} & 0 & \phi_{23} & \phi_{24} \\ -\phi_{13} & -\phi_{23} & 0 & \phi_{34} \\ -\phi_{14} & -\phi_{24} & -\phi_{34} & 0 \end{bmatrix} = \phi_{12}\phi_{34} - \phi_{13}\phi_{24} + \phi_{14}\phi_{23}$$

# COLD FERMI GASES



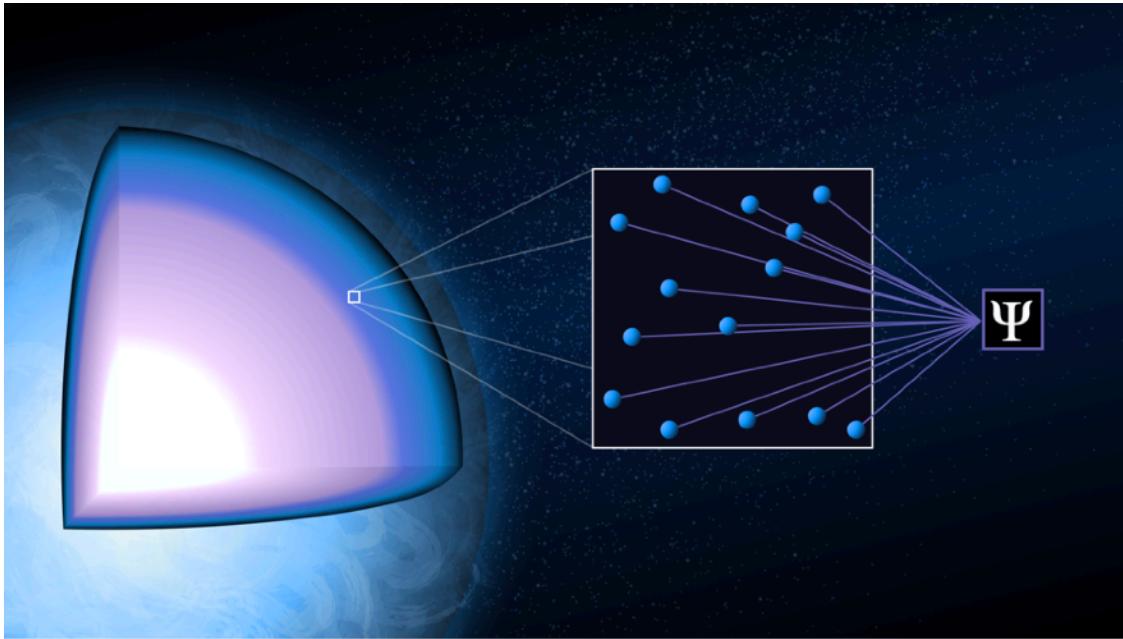
$$\left( \frac{E}{E_{FG}} \right)_{\text{exp}} = \xi = 0.376(5)$$

# COLD FERMI GASES

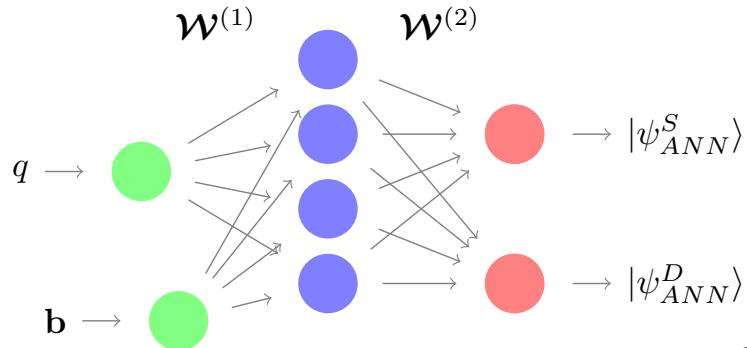


$$\left( \frac{E}{E_{FG}} \right)_{\text{exp}} = \xi = 0.376(5)$$

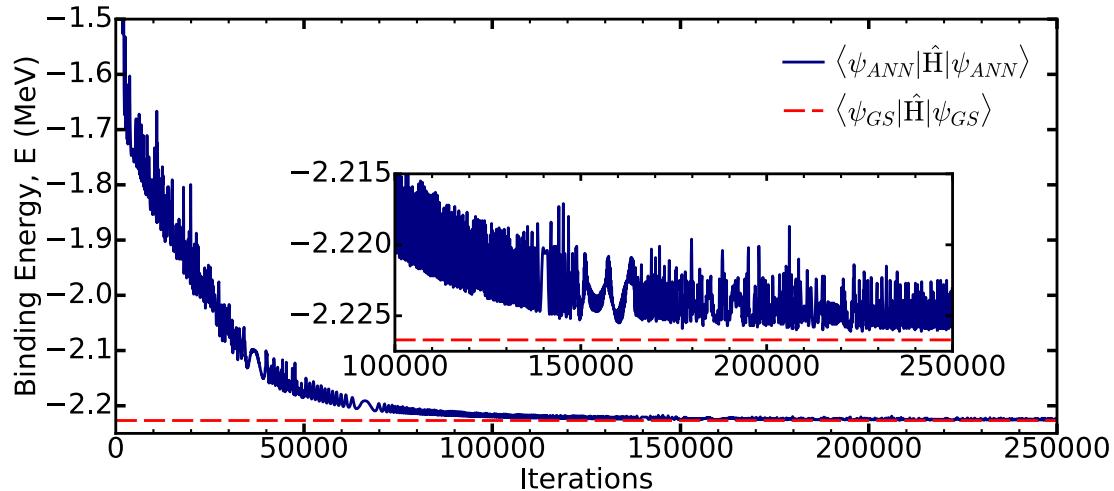
# BACK TO NUCLEAR PHYSICS



# NEURAL-NETWORK QUANTUM STATES



$$E^{\mathcal{W}} = \frac{\langle \Psi_{\text{ANN}}^{\mathcal{W}} | \hat{H} | \Psi_{\text{ANN}}^{\mathcal{W}} \rangle}{\langle \Psi_{\text{ANN}}^{\mathcal{W}} | \Psi_{\text{ANN}}^{\mathcal{W}} \rangle}$$



Keeble, Rios, PLB 809, 135743 (2020)

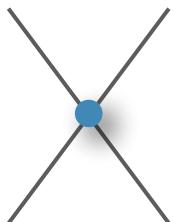
Sarmiento, et al., EPJ 139 (2024) 2, 189

# “ESSENTIAL” HAMILTONIAN

Input: Hamiltonian inspired by a LO pionless-EFT expansion

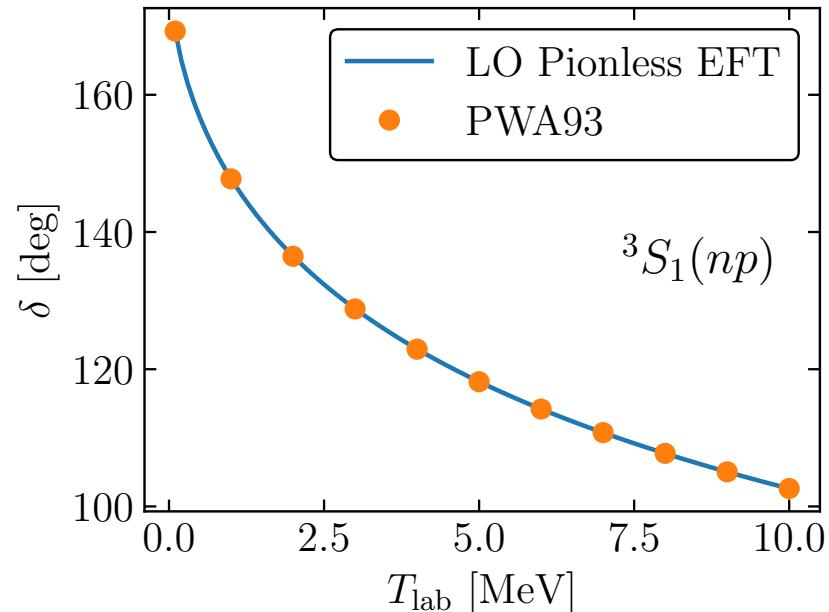
$$H_{LO} = - \sum_i \frac{\vec{\nabla}_i^2}{2m_N} + \sum_{i < j} v_{ij} + \sum_{i < j < k} V_{ijk}$$

- NN potential fit to s-wave np scattering lengths and effective ranges



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

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

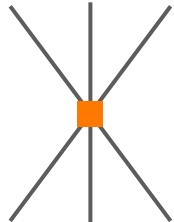


# “ESSENTIAL” HAMILTONIAN

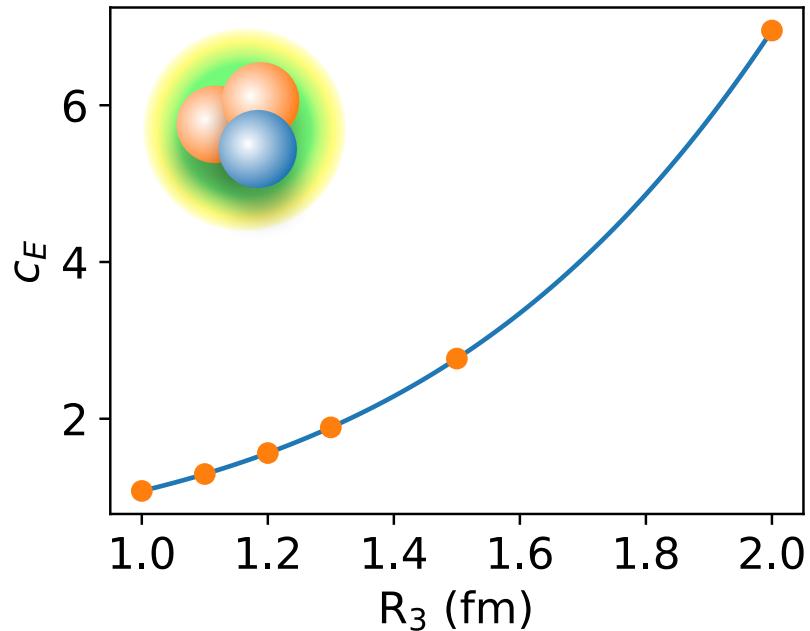
Input: Hamiltonian inspired by a LO pionless-EFT expansion

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

- 3NF adjusted to reproduce the energy of  ${}^3\text{H}$ .

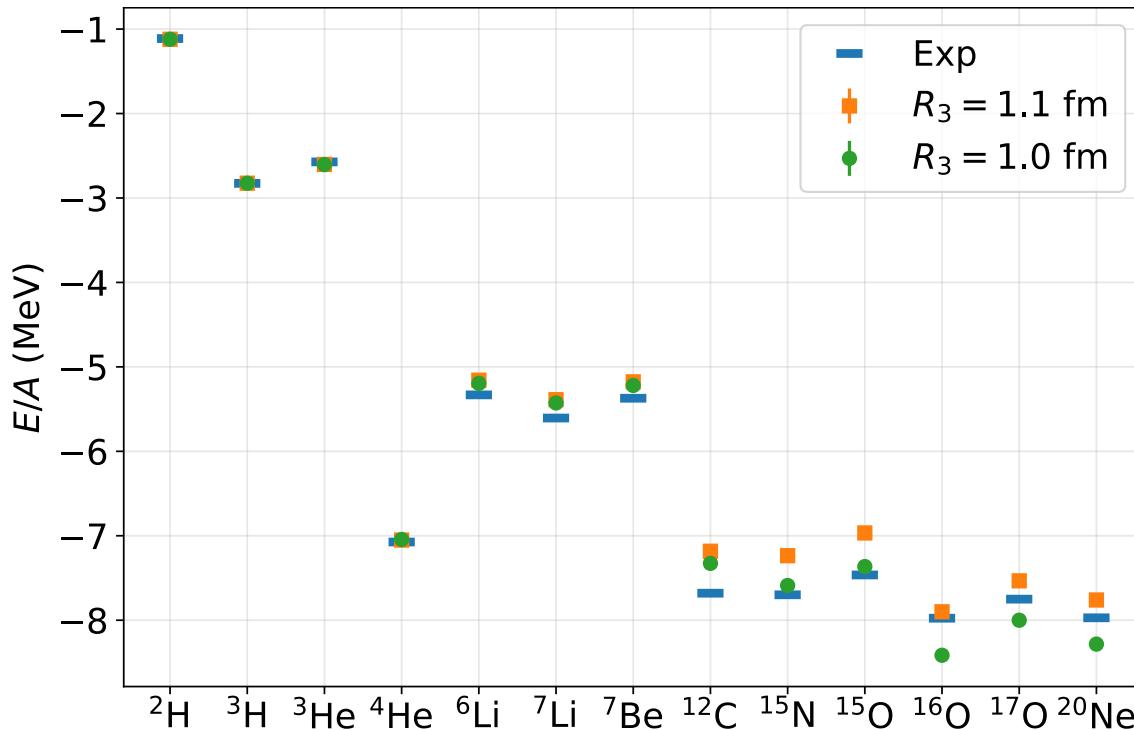


$$V_{ijk} \propto c_E \sum_{\text{cyc}} e^{-(r_{ij}^2 + r_{jk}^2)/R_3^2}$$

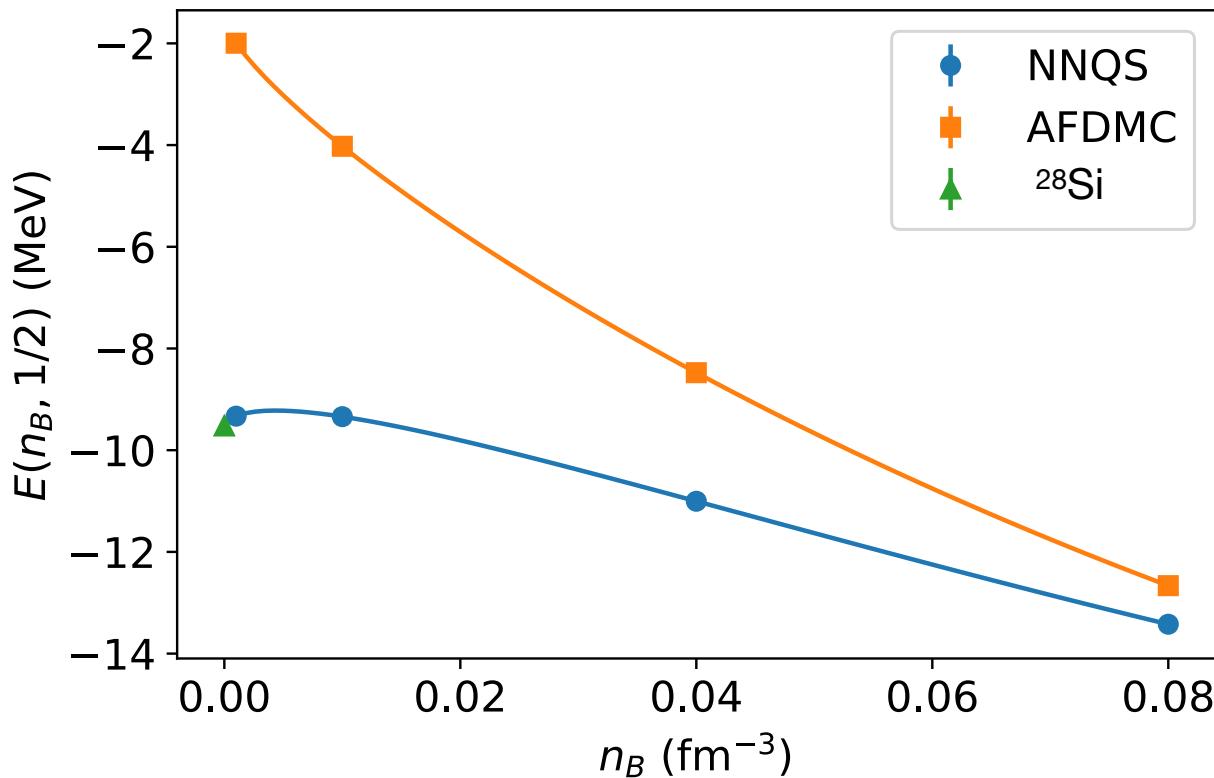


# “ESSENTIAL” HAMILTONIAN

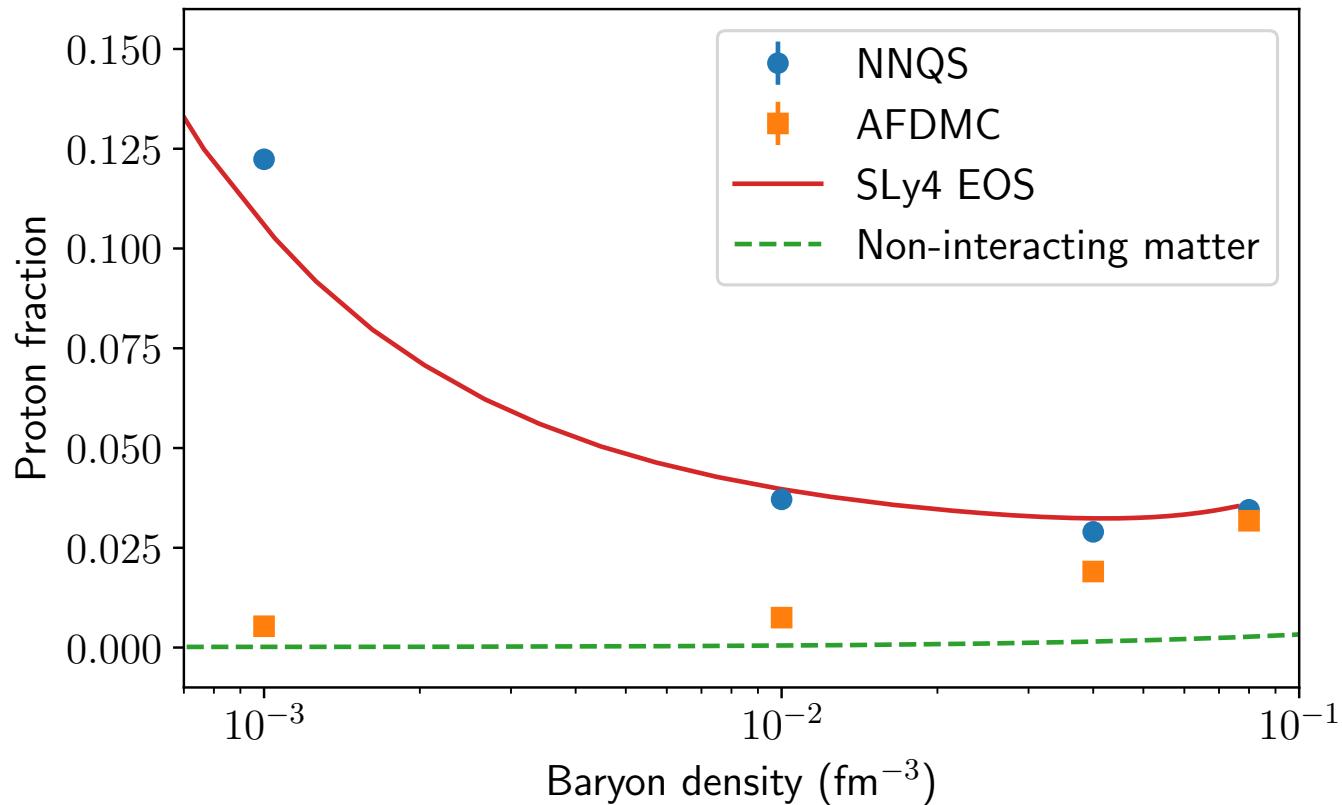
Our “essential” Hamiltonian reproduces well the spectrum of different nuclei



# DILUTE NUCLEONIC MATTER

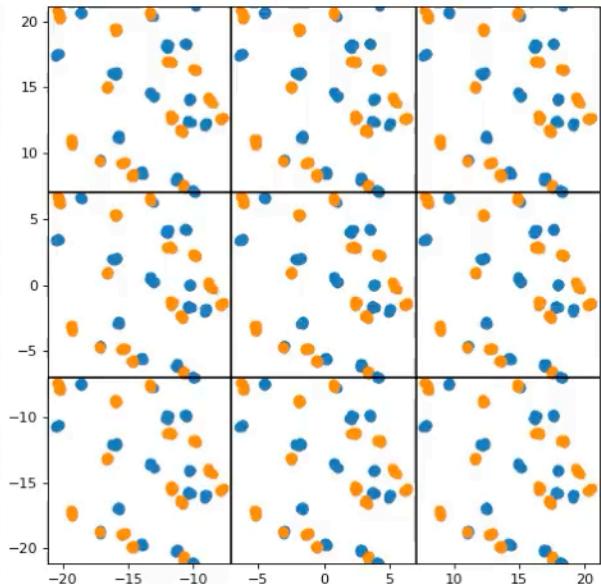
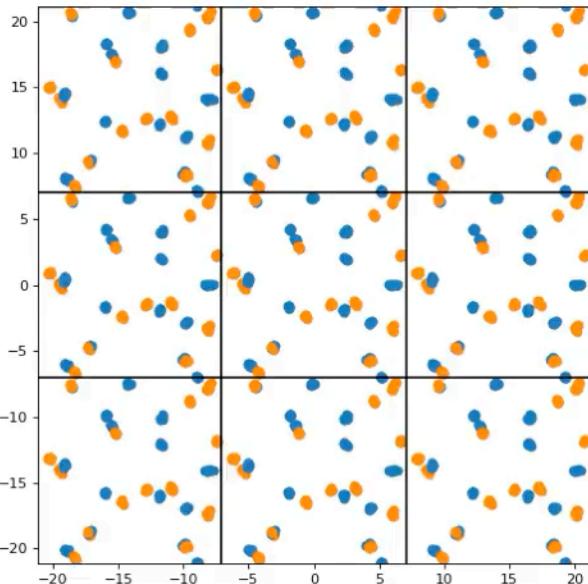
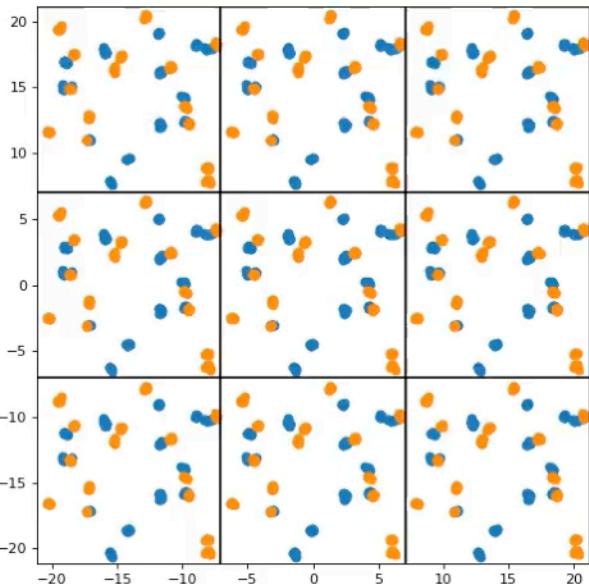


# DILUTE NUCLEONIC MATTER



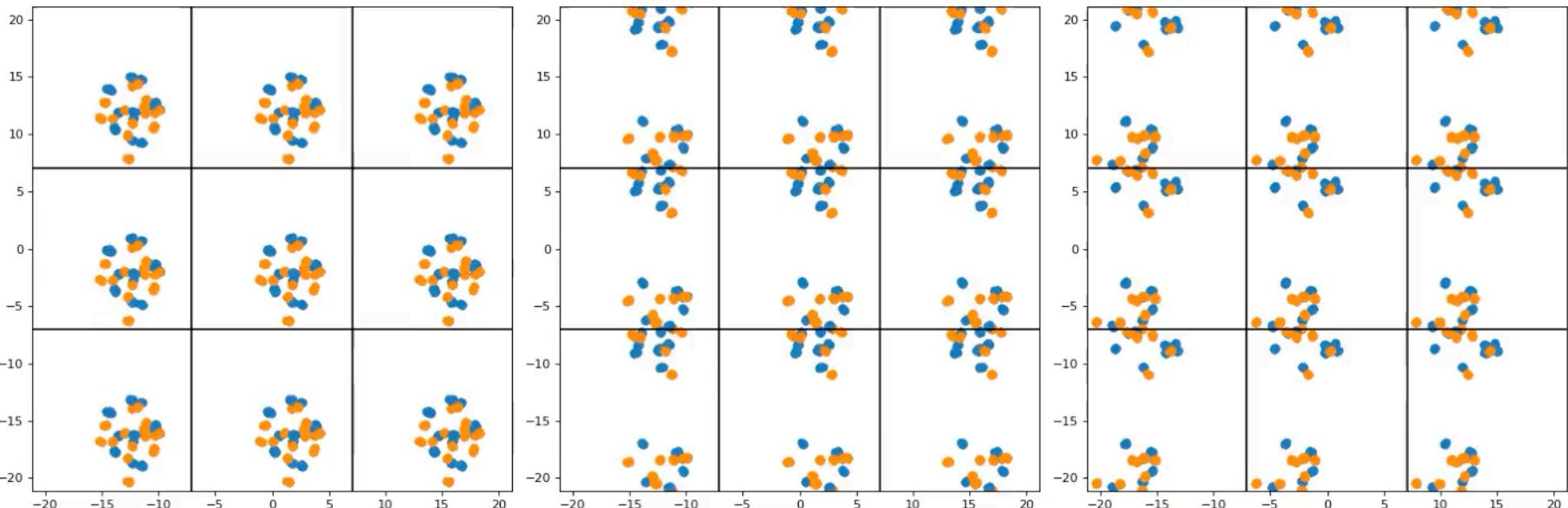
# DILUTE NUCLEONIC MATTER

14 Neutrons, 14 Protons @  $\rho=0.01 \text{ fm}^{-3}$



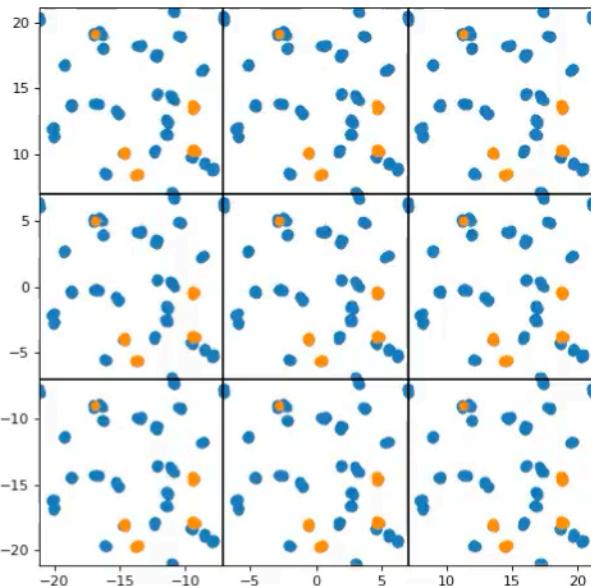
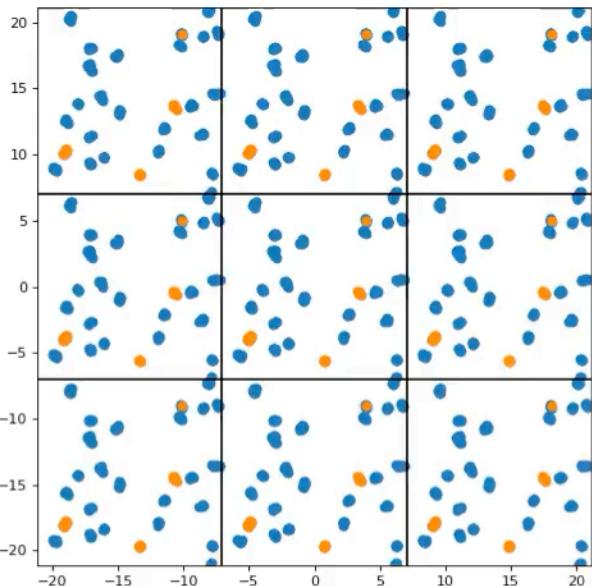
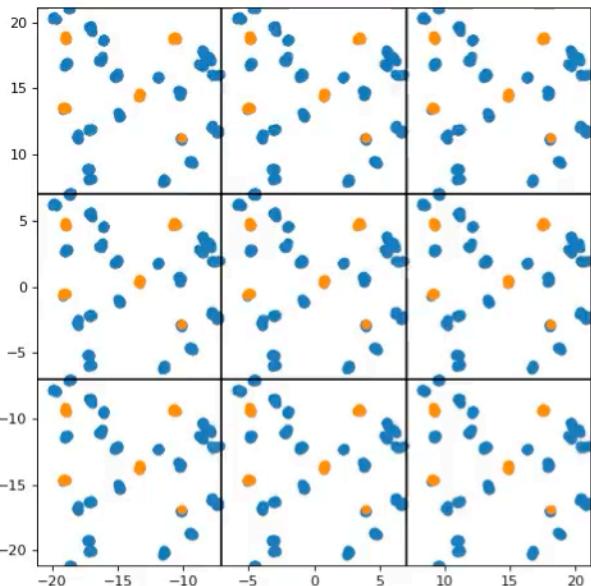
# DILUTE NUCLEONIC MATTER

14 Neutrons, 14 Protons @  $\rho=0.01 \text{ fm}^{-3}$



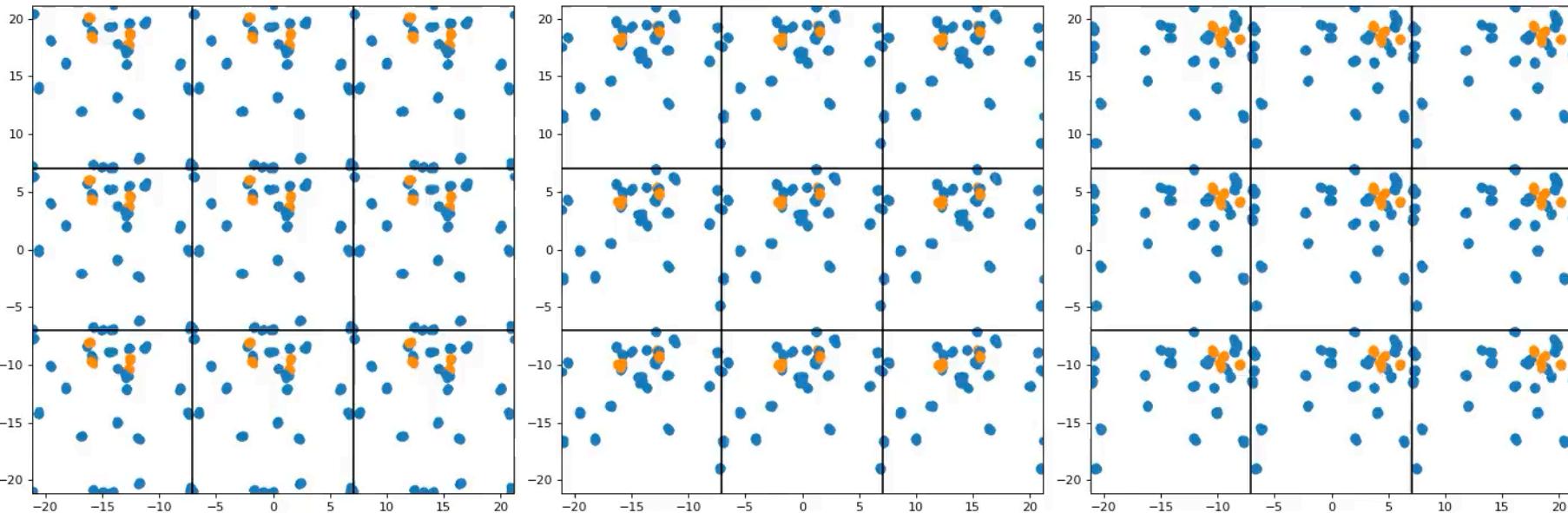
# DILUTE NUCLEONIC MATTER

24 Neutrons, 4 Protons @  $\rho=0.01 \text{ fm}^{-3}$

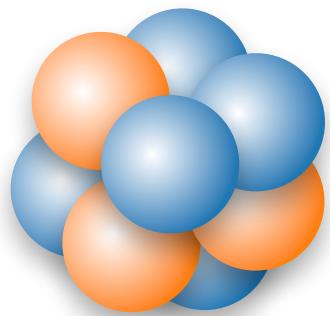


# DILUTE NUCLEONIC MATTER

24 Neutrons, 4 Protons @  $\rho=0.01 \text{ fm}^{-3}$

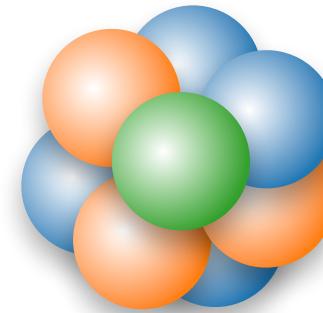


# HYPERNUCLEI



$^{12}\text{C}$

6 protons, 6 neutrons

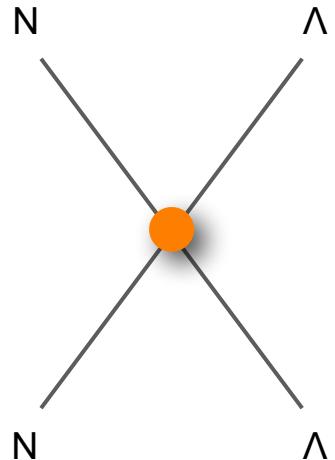


$^{12}_{\Lambda}\text{C}$

6 protons, 5 neutrons, 1 lambda

# HYPERNUCLEI

Input: Hamiltonian inspired by a LO pionless-EFT expansion



$$V_{\Lambda N} = \sum_{S,T} v_{ST}(r_{ij}) \hat{P}_{ST}$$

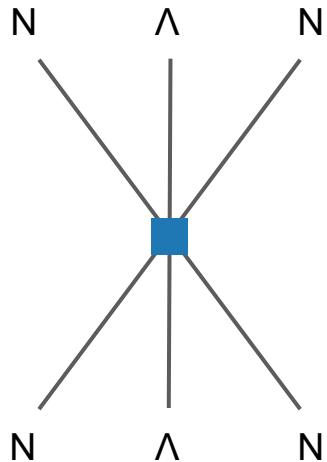
→ T=1/2, S=1

→ T=1/2, S=0

Parameters are determined by fitting proton- $\Lambda$  scattering length and effective range

# HYPERNUCLEI

Input: Hamiltonian inspired by a LO pionless-EFT expansion



$$V_{\Lambda NN} = \sum_{S,T} D_{S,T} v_{ST}(r_{i\Lambda}) v_{ST}(r_{j\Lambda}) \hat{Q}_{ST}$$

→ T=0, S=1/2

→ T=0, S=3/2

→ T=1, S=1/2

Parameters are determined by fitting  $^3_{\Lambda}\text{H}$ ,  $^4_{\Lambda}\text{H}(S=0)$ ,  $^4_{\Lambda}\text{H}(S=1)$ , and  $^5_{\Lambda}\text{He}$ .

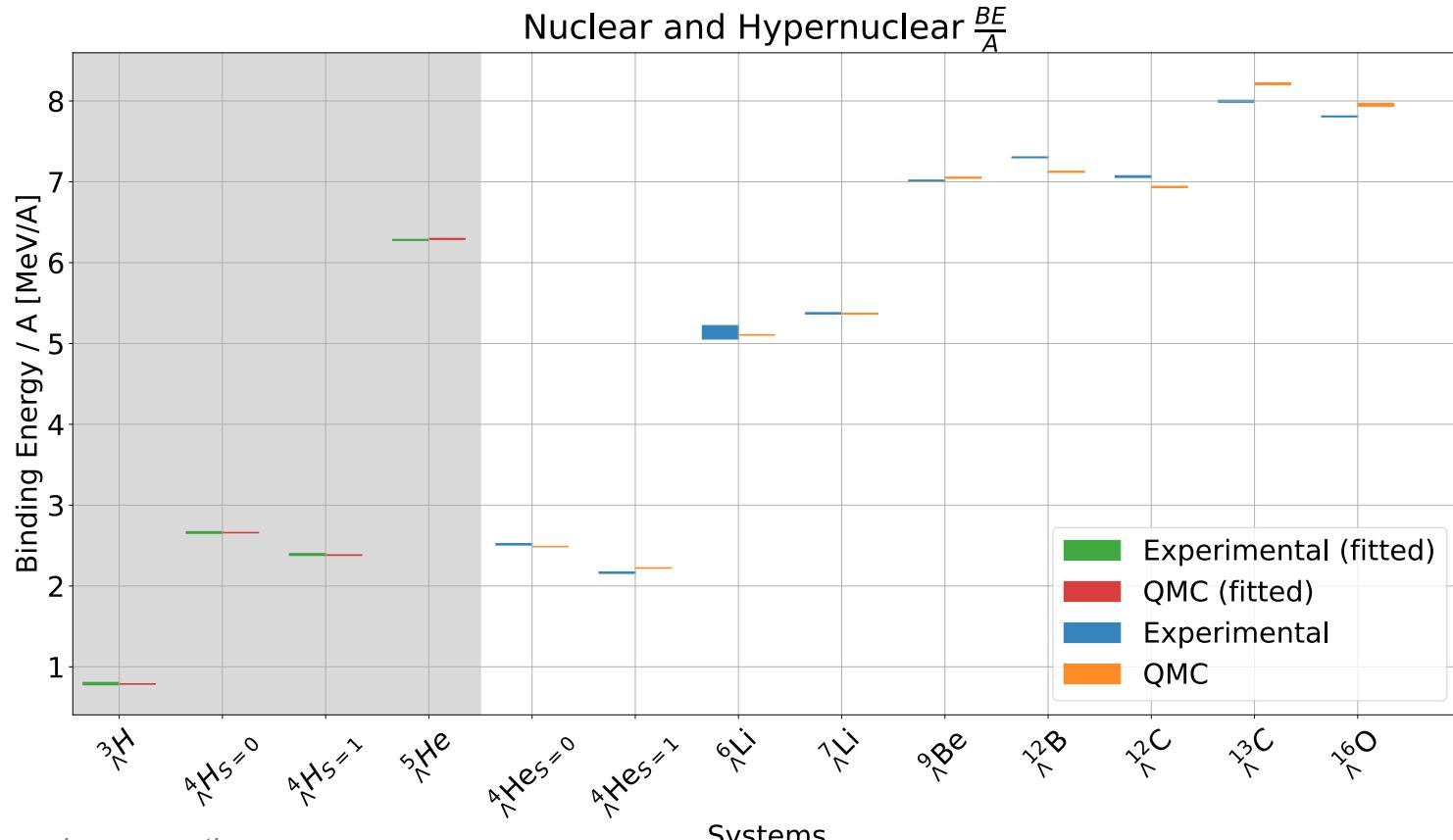
# HYPERNUCLEI

No terms in the Hamiltonian mix lambda and nucleons: **distinguishable**

**Andrea Di Donna** proposed the following ansatz

$$\Psi(x_\Lambda, x_1, \dots, x_A) = \mathcal{U}(x_\Lambda; x_1, \dots, x_A) \times \Psi_{HN}(x_1, \dots, x_A)$$

# HYPERNUCLEI



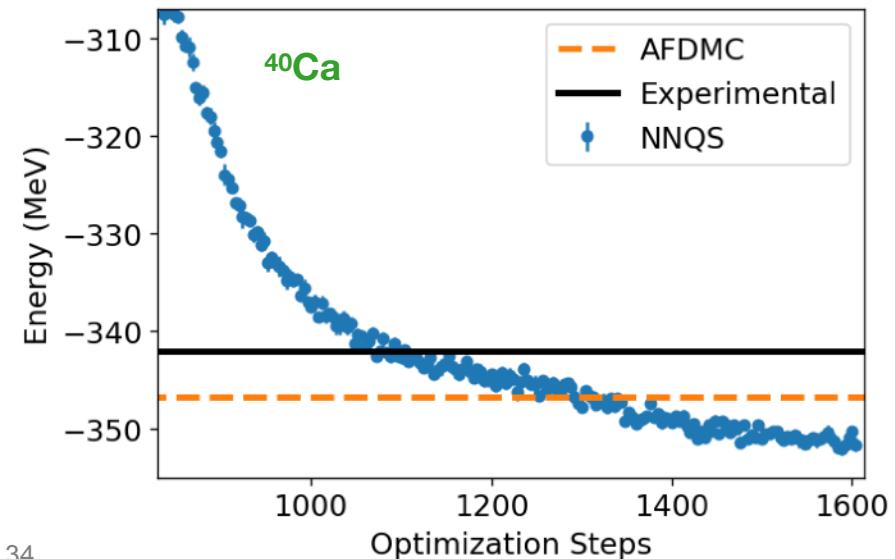
# CONCLUSIONS

NQS successfully applied to study:

- Ultra-cold Fermi gases, outperforming state-of-the-art continuum DMC;
- Dilute nucleonic matter, including the self-emergence of nuclei;
- Essential Elements of nuclear binding

Ongoing efforts:

- Medium-mass nuclei
- Excited states
- Chiral-EFT potentials
- Real-time dynamics



# THANK YOU