# Neural Network Quantum States for Atomic Nuclei

#### **ECT\*** online program

ADVANCES IN MANY-BODY THEORIES: FROM FIRST PRINCIPLE METHODS TO QUANTUM COMPUTING AND MACHINE LEARNING

November 3, 2020

Alessandro Lovato





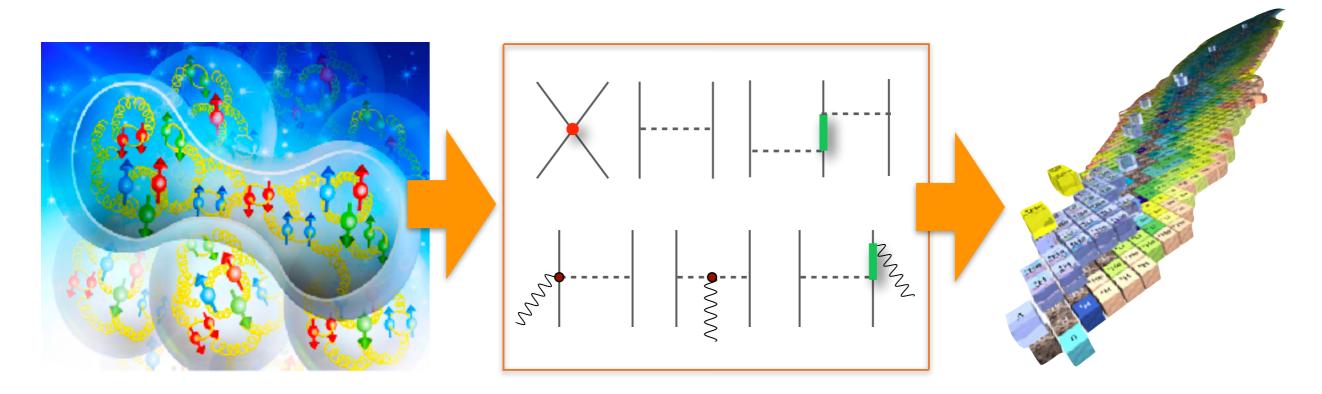
In collaboration with:

C. Adams, P. Balaprakash, G. Carleo, K. Raghavan, N. Rocco

## The microscopic model of nuclear theory

In the low-energy regime, quark and gluons are confined within hadrons and the relevant degrees of freedoms are protons, neutrons, and pions

Effective field theories are the link between QCD and nuclear observables. They exploit the separation between the "hard" (nucleon mass) and "soft" (exchanged momentum) scales



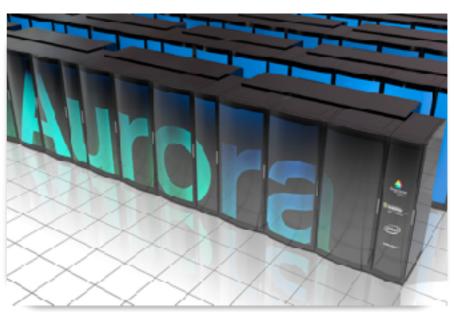
Nucleons can treated as point-like particles interacting through the Hamiltonian

$$H = \sum_{i} \frac{\mathbf{p}_{i}^{2}}{2m} + \sum_{i < j} v_{ij} + \sum_{i < j < k} V_{ijk} + \dots$$

# "Conventional" continuum nuclear quantum Monte Carlo methods



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



$$J_{mn} = \langle \Psi_m | J | \Psi_n \rangle$$

#### Variational Monte Carlo

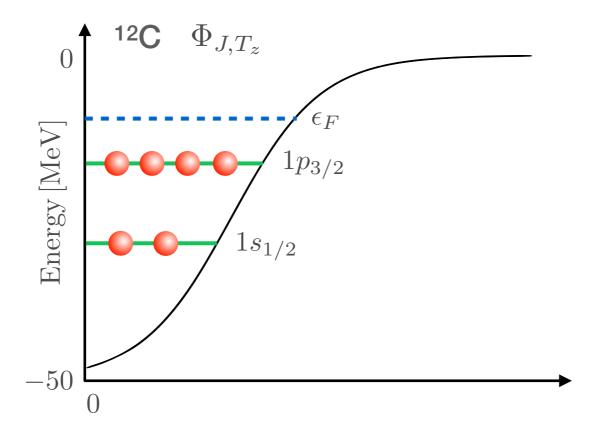
In variational Monte Carlo, one assumes a suitable form for the trial wave function

$$|\Psi_T\rangle = \left(1 + \sum_{ijk} F_{ijk}\right) \left(\mathcal{S} \prod_{i < j} F_{ij}\right) |\Phi_{J,T_z}\rangle$$

The best variational parameters are found by optimizing the variational energy

$$E_T = \langle \Psi_T | H | \Psi_T \rangle \geq E_0$$

The long-range antisymmetric is typically a Slater determinant of single-particle orbitals



#### Variational Monte Carlo

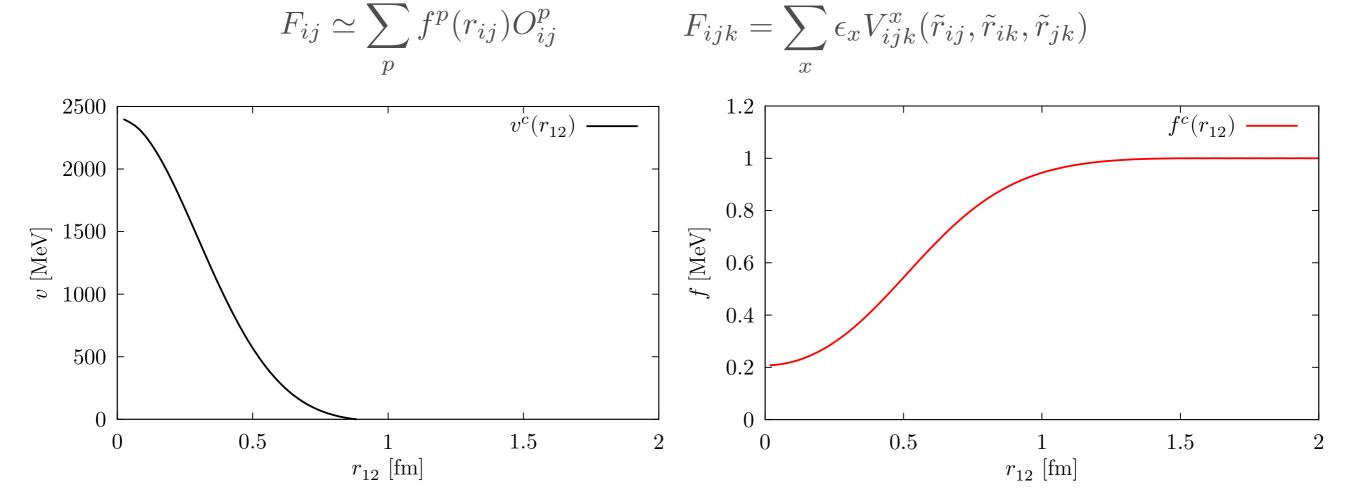
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The correlation operator reflects the spin-isospin dependence of the nuclear interaction



#### Green's function Monte Carlo

GFMC overcomes the limitations of the variational wave-function by using an imaginarytime projection technique

Any trial wave function can be expanded in the complete set of eigenstates of the the Hamiltonian according to

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

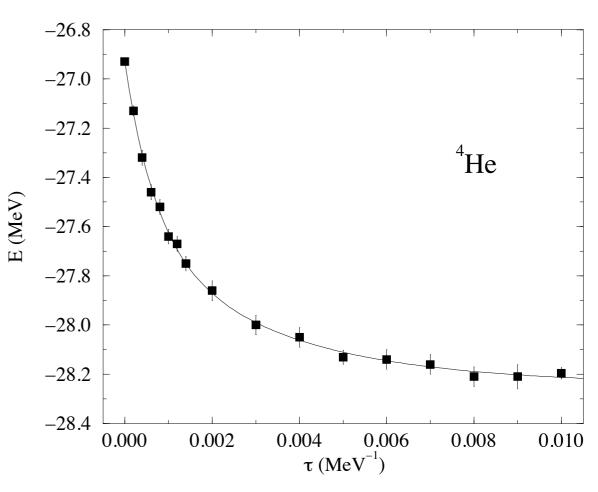
GFMC projects out the exact lowest-energy state, provided the trial wave function it is not

orthogonal to the ground state.

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

$$\lim_{\tau \to \infty} \sum_{n \to \infty} c_n e^{-(E_n - E_0)\tau} |\Psi_n\rangle = c_0 |\Psi_0\rangle$$

B. Pudliner et al., PRC 56, 1720 (1997)



#### Green's function Monte Carlo

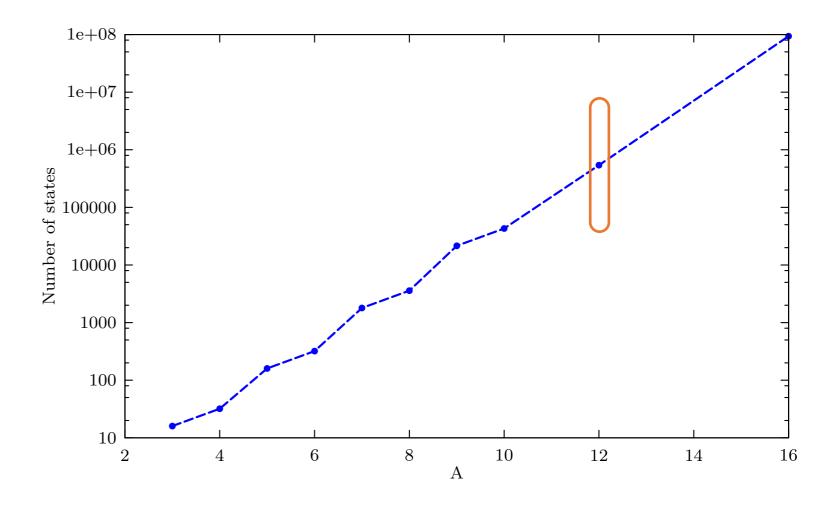
In the GFMC, a sum over all the many-body spin-isospin states is performed

$$\sum_{SS'} \langle S' | e^{-[V - E_0]\delta \tau} | S \rangle \simeq \sum_{SS'} \langle S' | \prod_{i < j} e^{-V_{ij}\delta \tau} | S \rangle e^{E_0\delta \tau}$$

Many-body spin-isospin states are utilized — unfavorable exponential scaling

The size of the spin state alone grows like 2<sup>A</sup>

$$|S\rangle = \begin{pmatrix} s \uparrow \uparrow \uparrow \\ s \uparrow \uparrow \downarrow \\ s \uparrow \downarrow \uparrow \\ s \downarrow \uparrow \uparrow \\ s \downarrow \downarrow \uparrow \\ s \downarrow \downarrow \downarrow \end{pmatrix}$$



GFMC is extremely accurate but **limited to A**≤12 nuclei and small (A ≤14) neutron systems

#### Auxiliary-field diffusion Monte Carlo

The AFDMC method uses a spin-isospin basis given by the outer product of single-nucleon spinors

$$|S\rangle = |s_1\rangle \otimes |s_2\rangle \cdots \otimes |s_A\rangle \quad \longleftrightarrow \quad |s\rangle = s_{p\uparrow}|p\uparrow\rangle + s_{p\downarrow}|p\downarrow\rangle + s_{n\uparrow}|n\uparrow\rangle + s_{n\downarrow}|n\downarrow\rangle$$

In purely neutron systems  $au_{ij}=1$  and the spin-dependent part of the potential reads

$$V_{SD} = \frac{1}{2} \sum_{i\alpha j\beta} A_{i\alpha,j\beta} \sigma_i^{\alpha} \sigma_j^{\beta} = \frac{1}{2} \sum_{n=1}^{3A} O_n^2 \lambda_n ,$$

To preserve the single-particle representation, the short-time propagator is linearized utilizing the **Hubbard-Stratonovich transformation** 

$$e^{-\lambda O^2 \delta \tau/2} = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} dx e^{-x^2/2} e^{x\sqrt{-\lambda \delta \tau} O}$$

At each imaginary-time step the single particle spinors are rotated

$$e^{-V\delta\tau}[|s_1\rangle\otimes|s_2\rangle\cdots\otimes|s_A\rangle]=|s_1'\rangle\otimes|s_2'\rangle\cdots\otimes|s_A'\rangle$$

AFDMC scales polynomially with the number of nucleons but is limited to simplified potentials

K. E. Schmidt et al., PLB **446**, 99 (1999)

An exponential in the particle number operations is required to evaluate the correlation operator

$$\left( \mathcal{S} \prod_{i < j} F_{ij} \right) |\Phi\rangle$$

Consider simple spin-pair correlation (no tensor)

$$F_{ij} = f^c(r_{ij}) + f^{\sigma}(r_{ij})\sigma_{ij} \quad \longleftrightarrow \quad \sigma_{ij} = 2P_{ij}^{\sigma} - 1$$

Let us apply the correlation operator to a three-body spin state

$$(f_{12}^c + f_{12}^\sigma \sigma_{12})(f_{13}^c + f_{13}^\sigma \sigma_{13})(f_{23}^c + f_{23}^\sigma \sigma_{23})|s_1, s_2, s_3\rangle$$

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$$= (f_{12}^c + f_{12}^\sigma \sigma_{12})(f_{13}^c + f_{13}^\sigma \sigma_{13})[(f_{23}^c - f_{23}^\sigma)|s_1, s_2, s_3\rangle + 2f_{23}^\sigma|s_1, s_3, s_2\rangle]$$

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$$= (f_{12}^{c} + f_{12}^{\sigma}\sigma_{12})(f_{13}^{c} + f_{13}^{\sigma}\sigma_{13})[(f_{23}^{c} - f_{23}^{\sigma})|s_{1}, s_{2}, s_{3}\rangle + 2f_{23}^{\sigma}|s_{1}, s_{3}, s_{2}\rangle]$$

$$= (f_{12}^{c} + f_{12}^{\sigma}\sigma_{12})[(f_{13}^{c} - f_{13}^{\sigma})(f_{23}^{c} - f_{23}^{\sigma})|s_{1}, s_{2}, s_{3}\rangle + 2(f_{13}^{c} - f_{13}^{\sigma})f_{23}^{\sigma}|s_{1}, s_{3}, s_{2}\rangle$$

$$+ 2f_{13}^{\sigma}(f_{23}^{c} - f_{23}^{\sigma})|s_{3}, s_{2}, s_{1}\rangle + 4f_{13}^{\sigma}f_{23}^{\sigma}|s_{2}, s_{3}, s_{1}\rangle]$$

Sampling the spin-isospin state and evaluating the trial wave function's amplitude for that sampled state still requires a number of operations exponential in the particle number

Recent AFDMC calculations employ a **linear approximation** to spin-isospin dependent correlations. Simple but it **violates the factorization theorem**.

S. Gandolfi, AL et al., PRC 90, 061306 (2014)

$$\left(S\prod_{i < j} F_{ij}\right)|\Phi_{J,T_z}\rangle \longrightarrow \left(\prod_{i < j} f_{ij}^c\right)\left(1 + \sum_{i < j} U_{ij}\right)|\Phi_{J,T_z}\rangle$$

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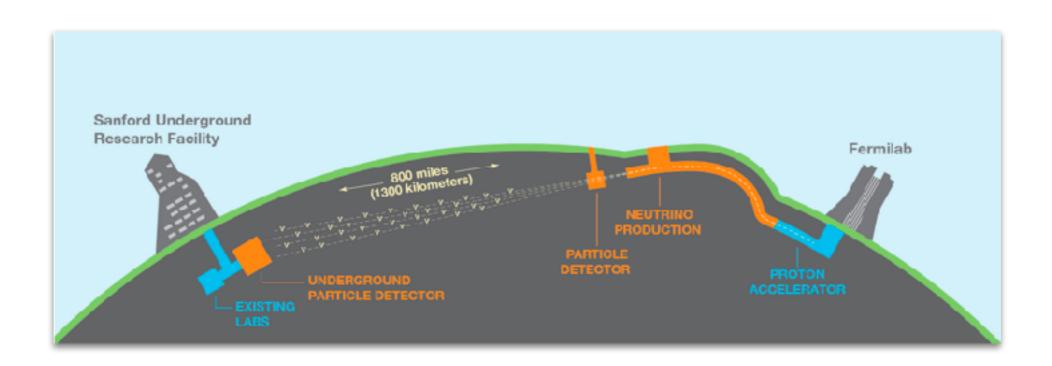
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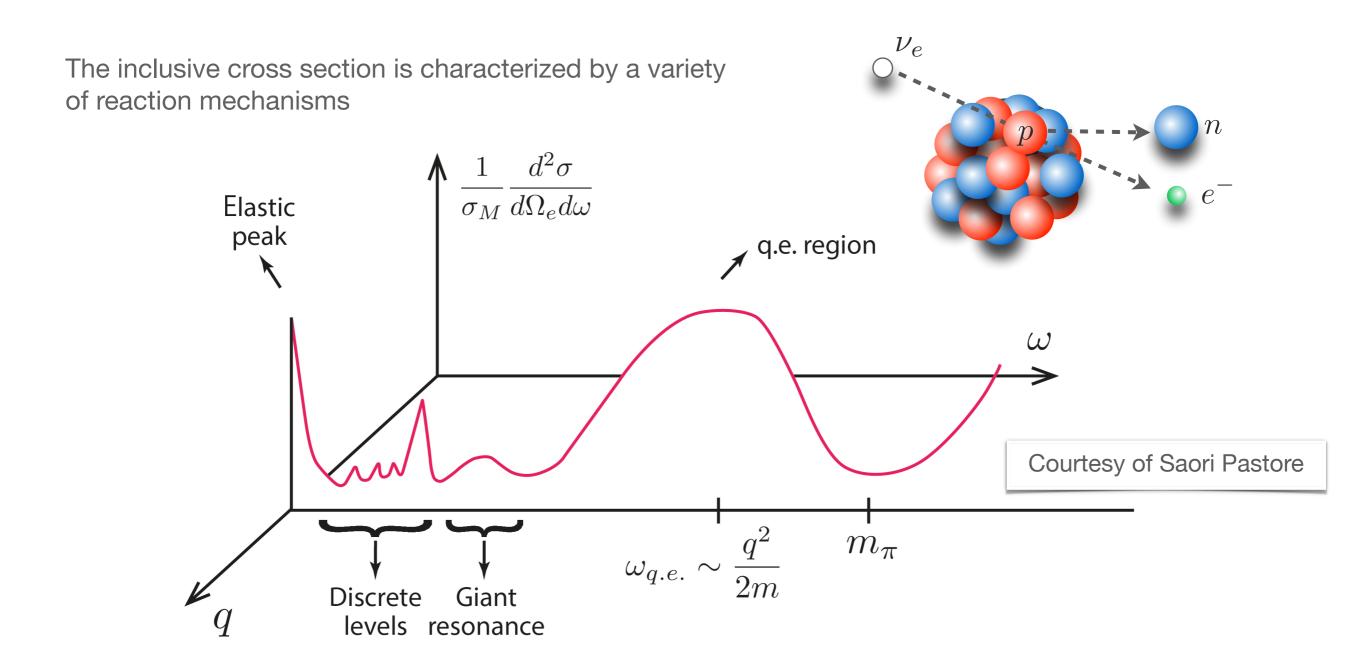
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$$\left(\prod_{i < j} f_{ij}^c\right)\left(1 + \sum$$

# Lepton-nucleus scattering with ML methods





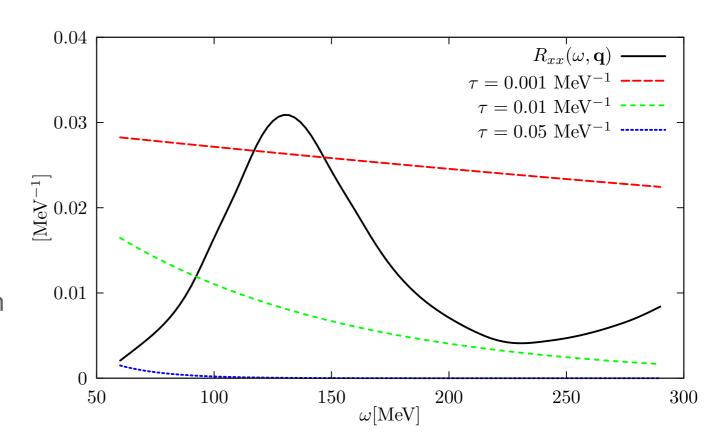
The response functions contain all information on the structure and dynamics of the target

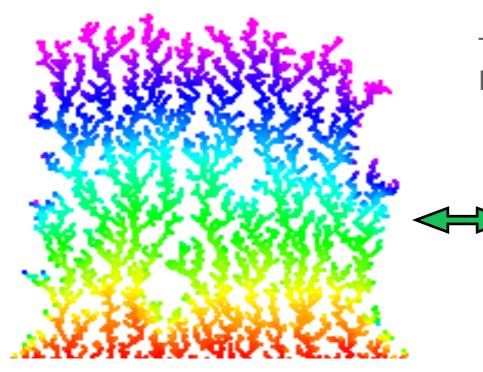
$$R_{\alpha\beta}(\omega, \mathbf{q}) = \sum_{f} \langle \Psi_0 | J_{\alpha}^{\dagger}(\mathbf{q}) | \Psi_f \rangle \langle \Psi_f | J_{\beta}(\mathbf{q}) | \Psi_0 \rangle \delta(\omega - E_f + E_0)$$

On type of integral transform is the **Laplace transform** 

$$E_{\alpha\beta}(\tau, \mathbf{q}) \equiv \int d\omega e^{-\omega\tau} R_{\alpha\beta}(\omega, \mathbf{q})$$

At finite imaginary time the contributions from large energy transfer are quickly suppressed





The system is first heated up by the transition operator. Its cooling determines the **Euclidean response** of the system

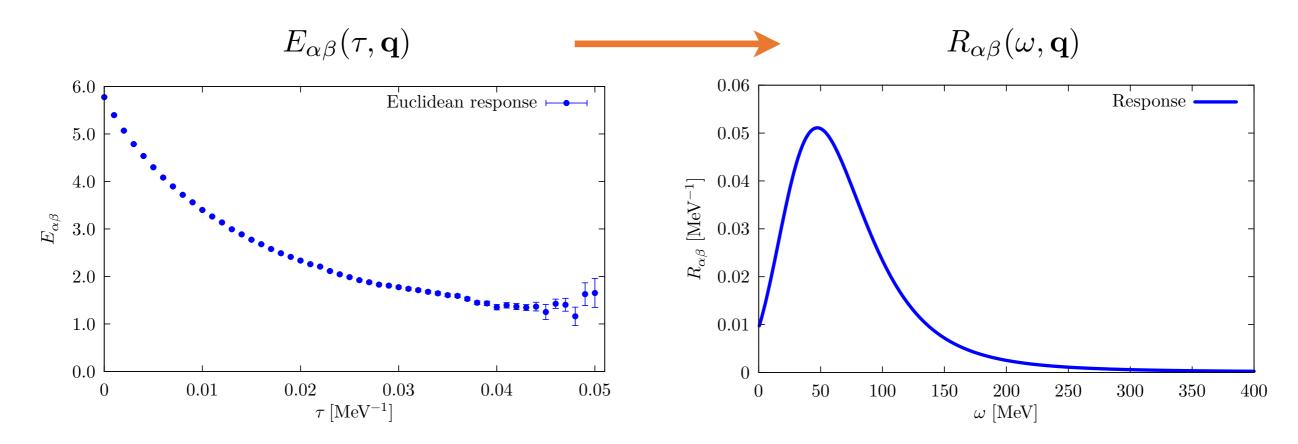
$$\sum_{f} |\Psi_{f}\rangle \langle \Psi_{f}|$$

$$E_{\alpha\beta}(\tau, \mathbf{q}) = \langle \Psi_{0}|J_{\alpha}^{\dagger}(\mathbf{q})e^{-(H-E_{0})\tau}J_{\beta}(\mathbf{q})|\Psi_{0}\rangle$$

Analogous techniques are used in Lattice QCD and condensed matter Physics

The Euclidean response formalism allows one to extract dynamical properties of the system from ground-state calculations

**Inverting the Euclidean response is an ill posed problem**: any set of observations is limited and noisy and the situation is even worse since the kernel is a smoothing operator.

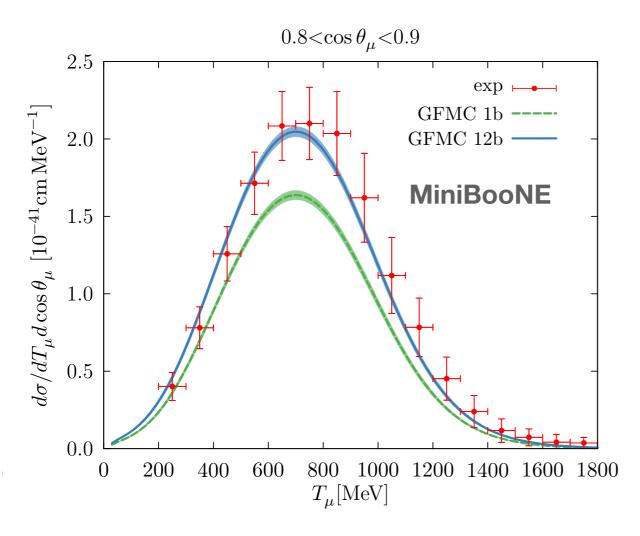


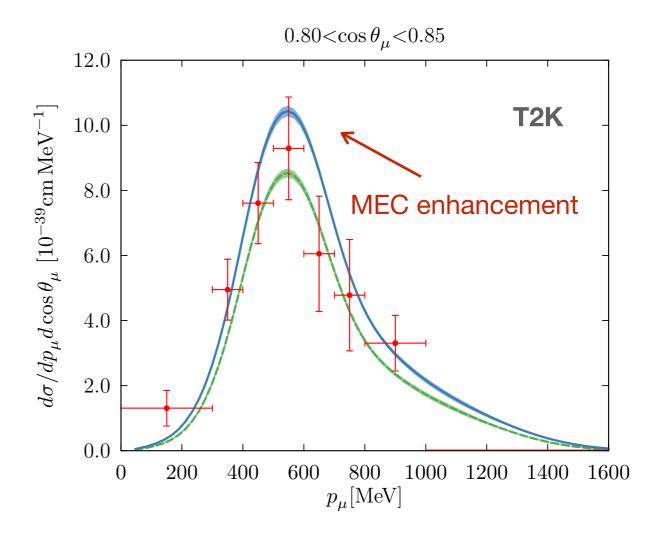
#### Maximum-entropy techniques:

- Reliable enough for quasi-elastic responses;
- Fail to reproduce the low-energy structure of the response functions;

First microscopic calculation of neutrino-12C cross section

$$\left\langle \frac{d\sigma}{dT_{\mu} d\cos\theta_{\mu}} \right\rangle = \int dE_{\nu} \,\phi(E_{\nu}) \,\frac{d\sigma(E_{\nu})}{dT_{\mu} d\cos\theta_{\mu}}$$





Treating larger nuclei, including <sup>16</sup>O and <sup>40</sup>Ar:

- Higher noise level in the Euclidean responses;
- Non-negligible low-energy transitions;

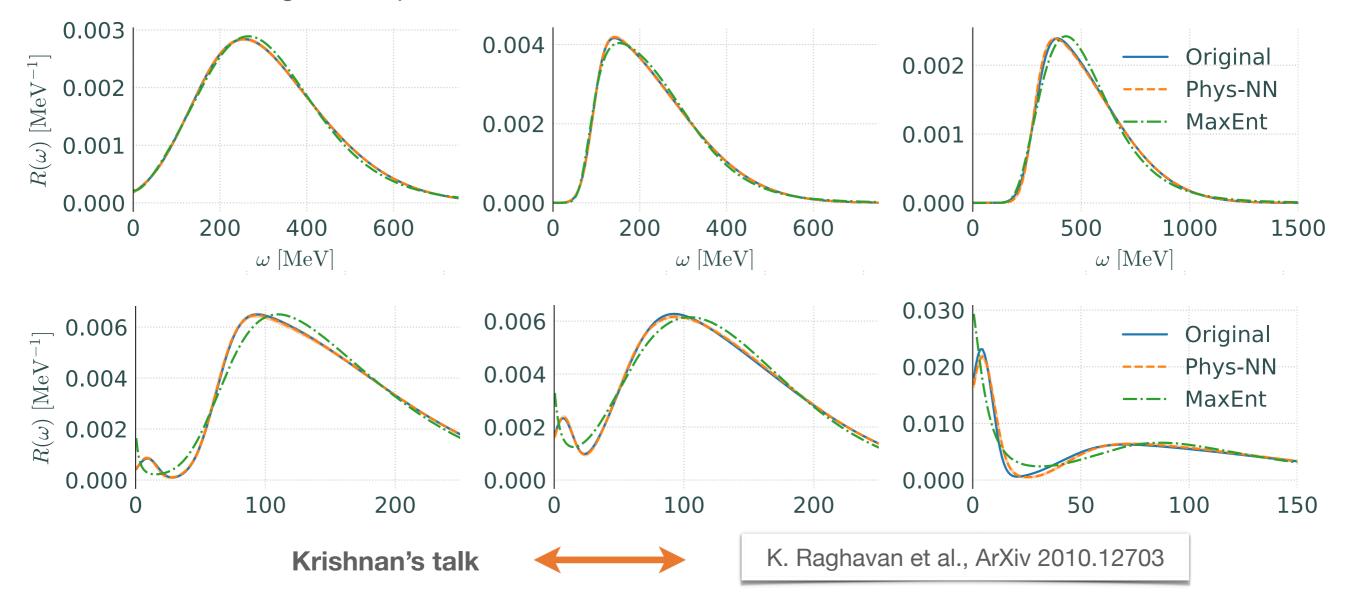
Need to go beyond Maximum Entropy

Several works have demonstrated that ML approaches are suitable for tackling inverse problems in condensed-matter physics

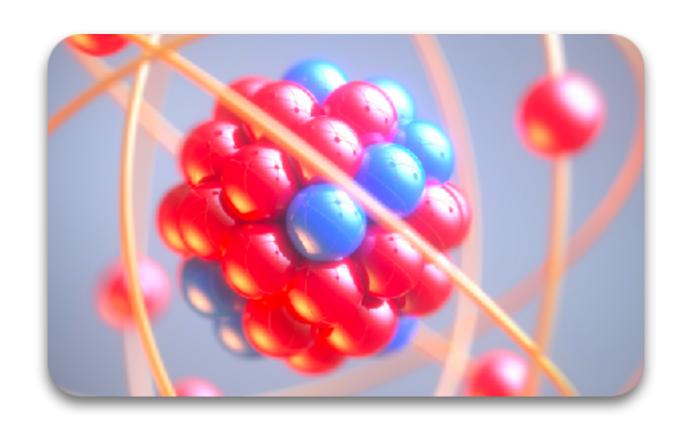
H. Yoo et al., Phys. Rev. B 98, 245101 (2018)

R. Fournier et al., Phys. Rev. Lett. 124, 056401 (2020)

We propose a **Physics-inspired ANN** to approximate the inverse of the Laplace transform of nuclear electromagnetic response functions



# Neural Network Quantum States for Atomic Nuclei

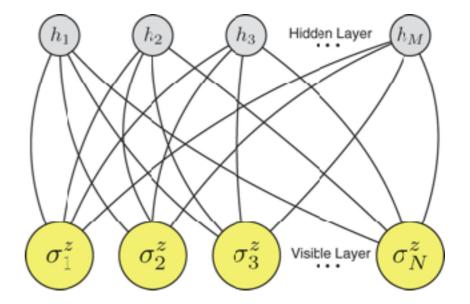


#### Neural-network quantum states

- Artificial neural networks (ANNs) can compactly represent complex high-dimensional functions;
- Variational representations of spin-systems quantum states based on ANNs have been found to outperform conventional variational ansatz;

G. Carleo et al. Science 355, 602 (2017)

G. Carleo et al. Nat. Commun. 9, 532 (2018)

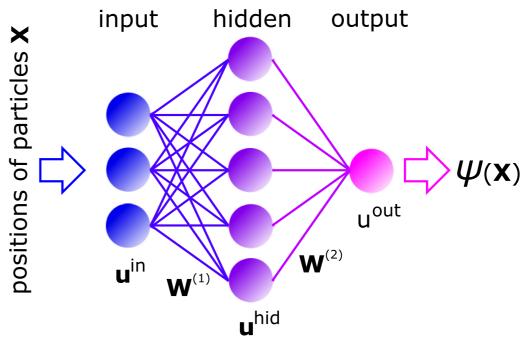


 Applications to the continuum to few-body systems and quantum chemistry problems have followed shortly thereafter;

H. Saito, J. Phys. Soc. Jpn. 87, 074002 (2018)

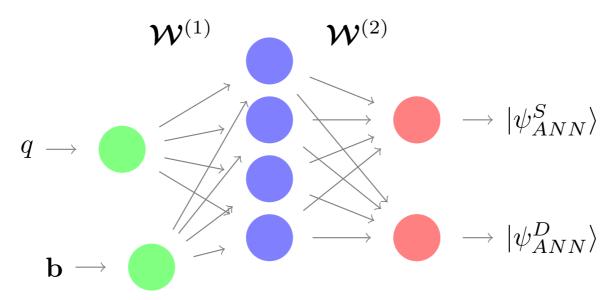
Pfau et al., arXiv:1909.02487 (2019)

Hermann et al., arXiv:1909.08423 (2019)

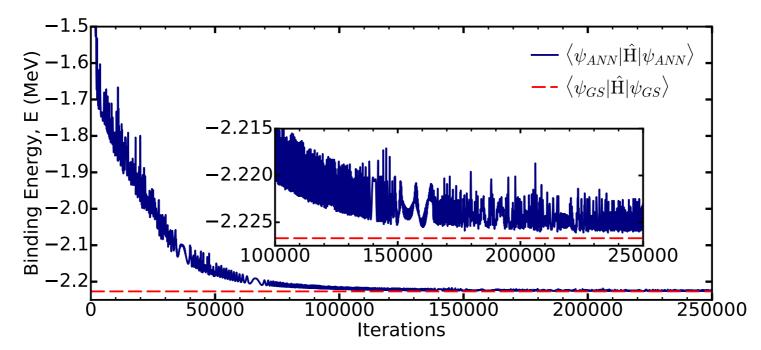


#### Machine-learning the deuteron

 ANN were recently applied to solve the deuteron in momentum space using the sophisticated N3LO Entem-Machleidt chiral-EFT nucleon-nucleon force



• The parameters of the ANN are optimized minimizing the variational energy using RMSprop



James Keeble talk



Keeble, Rios, PLB 809, 135743 (2020)

We consider a the leading-order pionless-EFT Hamiltonian, which includes a three-body force

$$H_{LO} = -\sum_{i} \frac{\vec{\nabla}_{i}^{2}}{2m_{N}} + \sum_{i < j} (C_{1} + C_{2} \vec{\sigma_{i}} \cdot \vec{\sigma_{j}}) e^{-r_{ij}^{2} \Lambda^{2}/4} + D_{0} \sum_{i < j < k \text{ cyc}} \sum_{cyc} e^{-(r_{ik}^{2} + r_{ij}^{2})\Lambda^{2}/4}$$

We introduce an ANN representation of the variational wave function of the form

$$|\Psi_V^{\text{ANN}}\rangle = e^{U(\mathbf{r}_1, \dots, \mathbf{r}_A)} |\Phi\rangle$$

More general Ansatz than standard product of two- and three-body Jastrow terms

$$|\Psi_V^J\rangle = \prod_{i < j < k} \left(1 - \sum_{\text{cyc}} u(r_{ij})u(r_{jk})\right) \prod_{i < j} f(r_{ij})|\Phi\rangle,$$

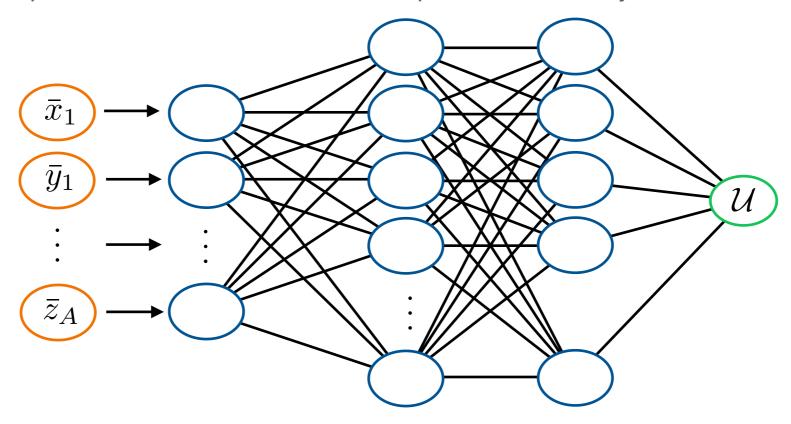
• For the s-shell nuclei that we consider, we assume the mean-field part to only depend upon the spin and isospin degrees of freedom

$$|\Phi_{^{2}H}\rangle = \mathcal{A}|\uparrow p, \uparrow n\rangle \qquad \longrightarrow \qquad \sigma_{12}|\Phi_{^{2}H}\rangle = |\Phi_{^{2}H}\rangle$$

$$|\Phi_{^{3}He}\rangle = \mathcal{A}|\uparrow p, \downarrow p, \uparrow n\rangle \qquad \longrightarrow \qquad \sigma_{12}|\Phi_{^{3}He}\rangle = -|\Phi_{^{3}He}\rangle$$

$$|\Phi_{^{4}He}\rangle = \mathcal{A}|\uparrow p, \downarrow p, \uparrow n, \downarrow n, \rangle \qquad \longrightarrow \qquad \sigma_{12}|\Phi_{^{4}He}\rangle = -|\Phi_{^{4}He}\rangle$$

The correlation factor is parameterized with an ANN comprised of four fully connected layers



- The center of mass contributions to the kinetic energy are removed by  $\, ar{{f r}}_i = {f r}_i {f R}_{
  m CM} \,$
- The kinetic energy requires the derivatives of  $\mathcal{U}$ . We use differentiable softplus activation functions.
- A Gaussian function is added to effectively confine the nucleons within a finite volume

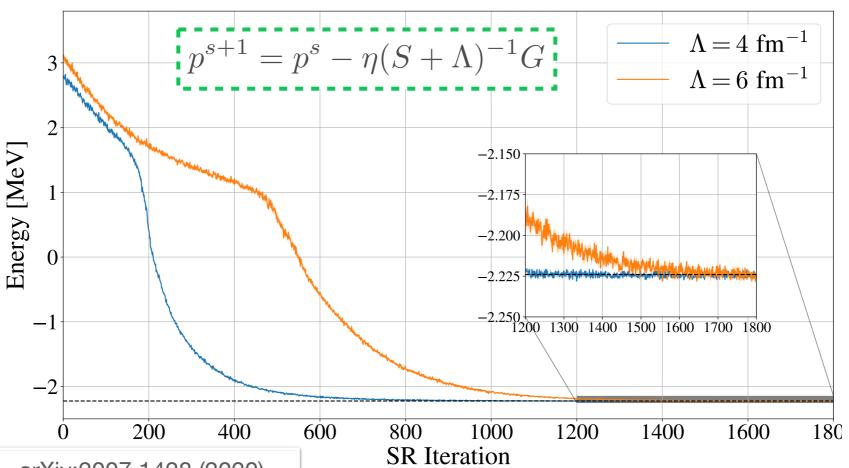
$$\mathcal{U}(\mathbf{r}_1,\ldots,\mathbf{r}_A) \longrightarrow \mathcal{U}(\mathbf{r}_1,\ldots,\mathbf{r}_A) - \alpha \sum_i \bar{\mathbf{r}}_i^2$$

The optimal values of the 18304 trainable parameters are found minimizing

$$E_V = \frac{\langle \Psi_V | H | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle}$$

Using a pre-conditioner based on the quantum Fisher information (analogous to the **stochastic reconfiguration method**) is more efficient than stochastic gradient descent

$$G_{i} = 2\left(\frac{\langle \partial_{i}\Psi_{V}|H|\Psi_{V}\rangle}{\langle \Psi_{V}|\Psi_{V}\rangle} - E_{V}\frac{\langle \partial_{i}\Psi_{V}|\Psi_{V}\rangle}{\langle \Psi_{V}|\Psi_{V}\rangle}\right) \quad S_{ij} = \frac{\langle \partial_{i}\Psi_{V}|\partial_{j}\Psi_{V}\rangle}{\langle \Psi_{V}|\Psi_{V}\rangle} - \frac{\langle \partial_{i}\Psi_{V}|\Psi_{V}\rangle\langle \Psi_{V}|\partial_{j}\Psi_{V}\rangle}{\langle \Psi_{V}|\Psi_{V}\rangle\langle \Psi_{V}|\Psi_{V}\rangle}$$

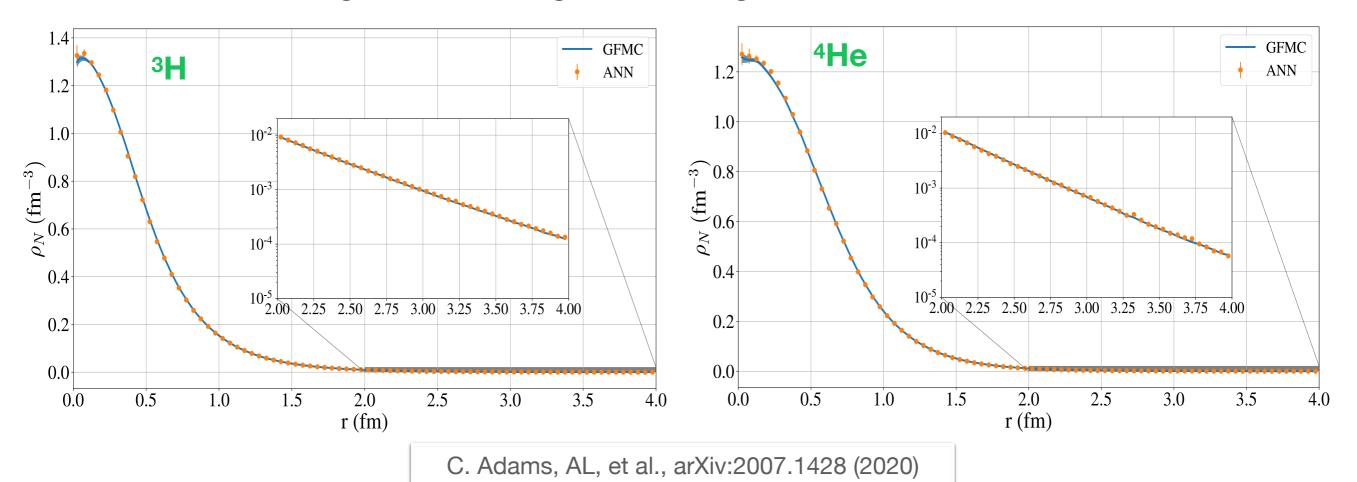


C. Adams, AL, et al., arXiv:2007.1428 (2020)

•The ANN ansatz outperforms standard Jastrow correlations, but lacks spin-isospin correlations

	Λ	VMC-ANN	VMC-JS	GFMC	$\overline{\mathrm{GFMC}_c}$
	$4 \text{ fm}^{-1}$	$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	-2.223(1)	-2.224(1)	-
					-
	$4~\mathrm{fm}^{-1}$	-7.81(1) $-7.79(3)$	-7.80(1)	-8.38(2)	-7.82(1)
<sup>4</sup> He	$4~\mathrm{fm}^{-1}$	$ \begin{array}{r} -22.76(3) \\ -23.99(6) \end{array} $	-22.54(1)	-23.62(3)	-22.77(2)
	$6 \text{ fm}^{-1}$	-23.99(6)	-23.44(2)	-25.06(3)	-24.10(2)

• Excellent results for ground-state energies and for single-nucleon densities

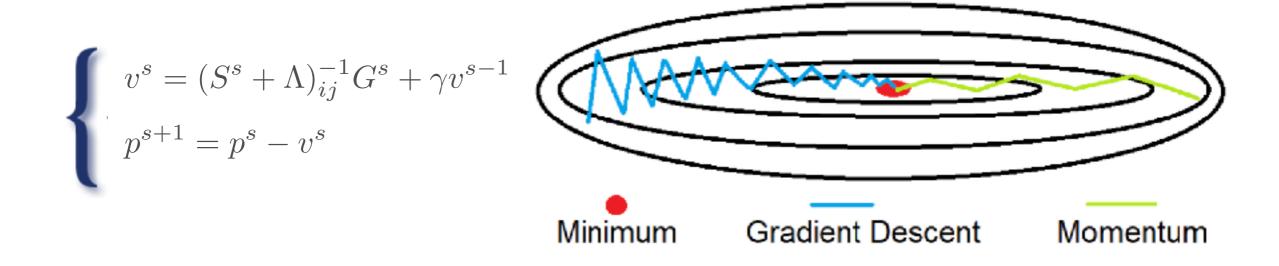


### Some perspectives

We are improving the convergence of the training combining SR with state-of-the-art optimization methods developed in ML applications;

$$G_i \longrightarrow (S+\Lambda)_{ij}^{-1}G_i$$

As a first example, we considered the "Momentum" method, which helps accelerate SGD in the relevant direction and dampens oscillations;



With momentum we already achieved a tenfold speedup in the convergence. We plan to test **RMSprop** and **Adam**, two of the most popular optimization methods.

### Some perspectives

Including spin-dependent correlations requires encoding the anti-symmetry of the wave function. We are investigating the use of a **generalized back-flow ansatz** 

$$\hat{\psi}(X) = \det \begin{pmatrix} \phi_1(x_1; x_{j \neq 1}) & \phi_1(x_2; \{x_{j \neq 2}\}) & \dots & \phi_1(x_A; \{x_{j \neq A}\}) \\ \phi_2(x_1; x_{j \neq 1}) & \phi_2(x_2; \{x_{j \neq 2}\}) & \dots & \phi_2(x_A; \{x_{j \neq A}\}) \\ \vdots & \vdots & & \vdots \\ \phi_A(x_1; x_{j \neq 1}) & \phi_A(x_2; \{x_{j \neq 2}\}) & \dots & \phi_A(x_A; \{x_{j \neq A}\}) \end{pmatrix},$$

Hermann et al., arXiv:1909.08423 (2019)

Pfau et al., arXiv:1909.02487 (2019)

- We introduced the generalized coordinate  $x_i = \{\mathbf{r}_i, s_i^z, t_i^z\}$
- The functions  $\phi_i(x_i; x_{j \neq i})$  must be invariant under the exchange of the order of  $x_{j \neq i}$ : need to Generalization of the **Deep Set architecture** to the complex case

$$X \in \mathbb{R}^{M} \longrightarrow \mathbb{R}^{NxM} \longrightarrow \mathbb{R}^{N} \longrightarrow \mathbb{R}$$

$$Wagstaff et al., arXiv:1901.09006 (2019)$$

## Some perspectives

Our primary goal is to devise accurate wave functions for medium-mass nuclei otherwise inaccessible by conventional nuclear QMC methods;



Compact representations useful to both theory and experimental communities

ANN allows study **properties of excited states**, a central task and nontrivial challenge for several nuclear many-body quantum approaches;

K. Choo et al. Phys. Rev. Lett. 121, 167204 (2018)

ANN can be extended to solve the time-dependent Schrödinger equation, similarly to the time-dependent variational Monte Carlo method. The distance to be minimized is

$$\min_{\mathbf{p}} \left\| i \frac{\partial \Psi_{\mathbf{V}}^{\mathbf{ANN}}}{\partial t} - H \Psi_{\mathbf{V}}^{\mathbf{ANN}} \right\|$$



Response function calculations; interplay with quantum-computing methods

## Summary and outlook

- We have developed a noise-resilient ML-based protocol for inverting the Laplace transform
  - More accurate than Maximum Entropy, especially in the low energy-transfer region;
  - Resilient to noise in the input Euclidean response, see Krishnan's talk;
  - Applicable to other integral-transform, including the Lorentz and the Gauss;
- Encouraging results with ANN representation of nuclear variational states
  - The ANN correlation outperforms existing two- and three-body Jastrow ansatz;
  - Using the SR method enables relatively fast training, further improved when used in combination with state-of-the-art ML optimization methods;
  - Correct behavior of single-particle densities, including the slowly decaying tail;
  - Extension to spin-isospin dependent correlation (and to larger nuclei) requires encoding permutation-invariant symmetry;
  - Access to excited states and nuclear response functions;