

# A hybrid Gorkov-Dyson SCGF for infinite matter and Neural Network based methods in nuclei

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# Rooting Nuclear Density Functional into Ab Initio Theory

PHYSICAL REVIEW C **104**, 024315 (2021)

## Nuclear energy density functionals grounded in *ab initio* calculations

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F. Marino  
(EPS PhD thesis  
prize, 2025)

DFT is in principle exact – but the energy density functional (EDF) is not known

For nuclear physics this is even more demanding: need to link the EDF to theories rooted in QCD!

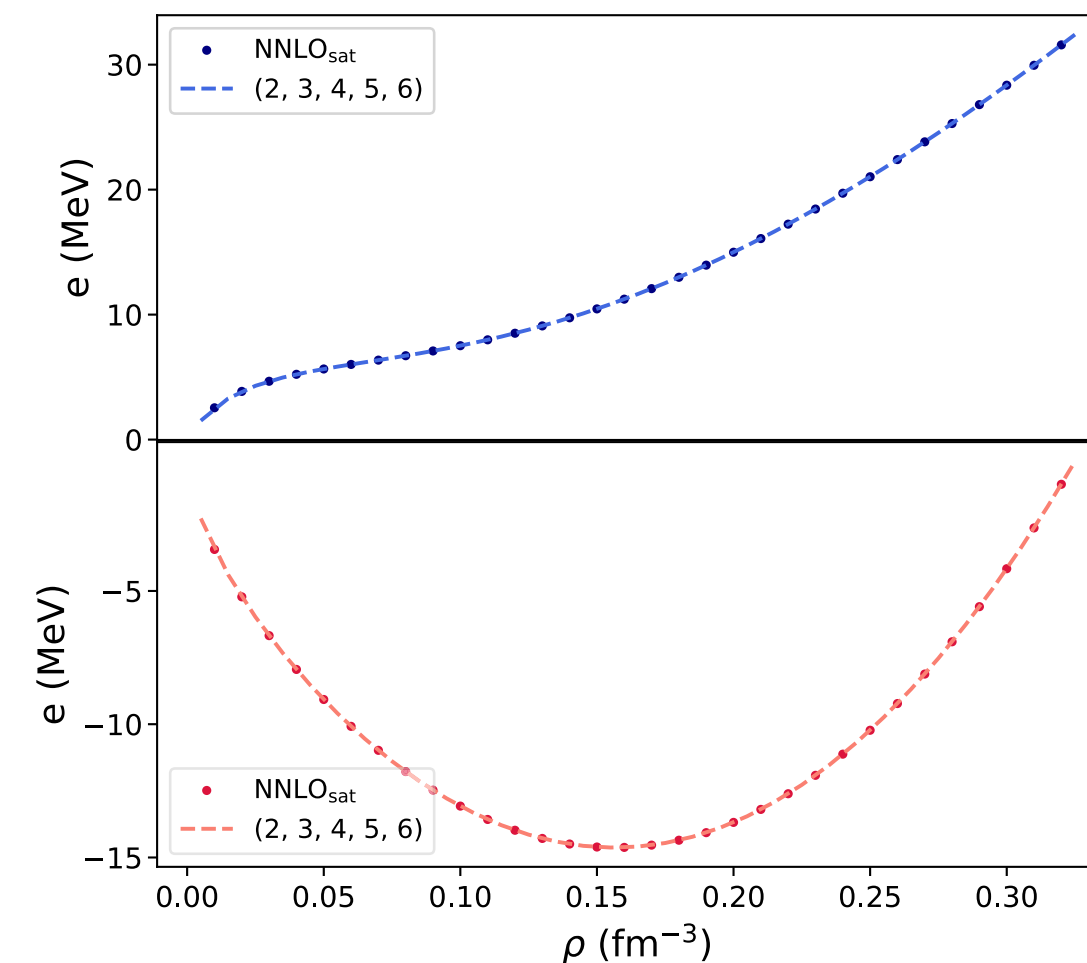
Machine-learn DFT functional on the nuclear equation of state

Jacob's ladder



+ approximate GA

Benchmark in finite systems



$$E = \int d\mathbf{r} \mathcal{E}(\mathbf{r}) = E_{\text{kin}} + E_{\text{pot}} + E_{\text{Coul}}$$

$$E_{\text{GA}} = E_{\text{LDA}} + E_{\text{surf}}$$

$$E_{\text{surf}} = \int d\mathbf{r} \left[ \sum_{t=0,1} C_t^\Delta \rho_t \Delta \rho_t - \frac{W_0}{2} \left( \rho \nabla \cdot \mathbf{J} + \sum_q \rho_q \nabla \cdot \mathbf{J}_q \right) \right]$$



# Benchmark on finite systems

Machine-learn DFT functional  
on the nuclear equation of state

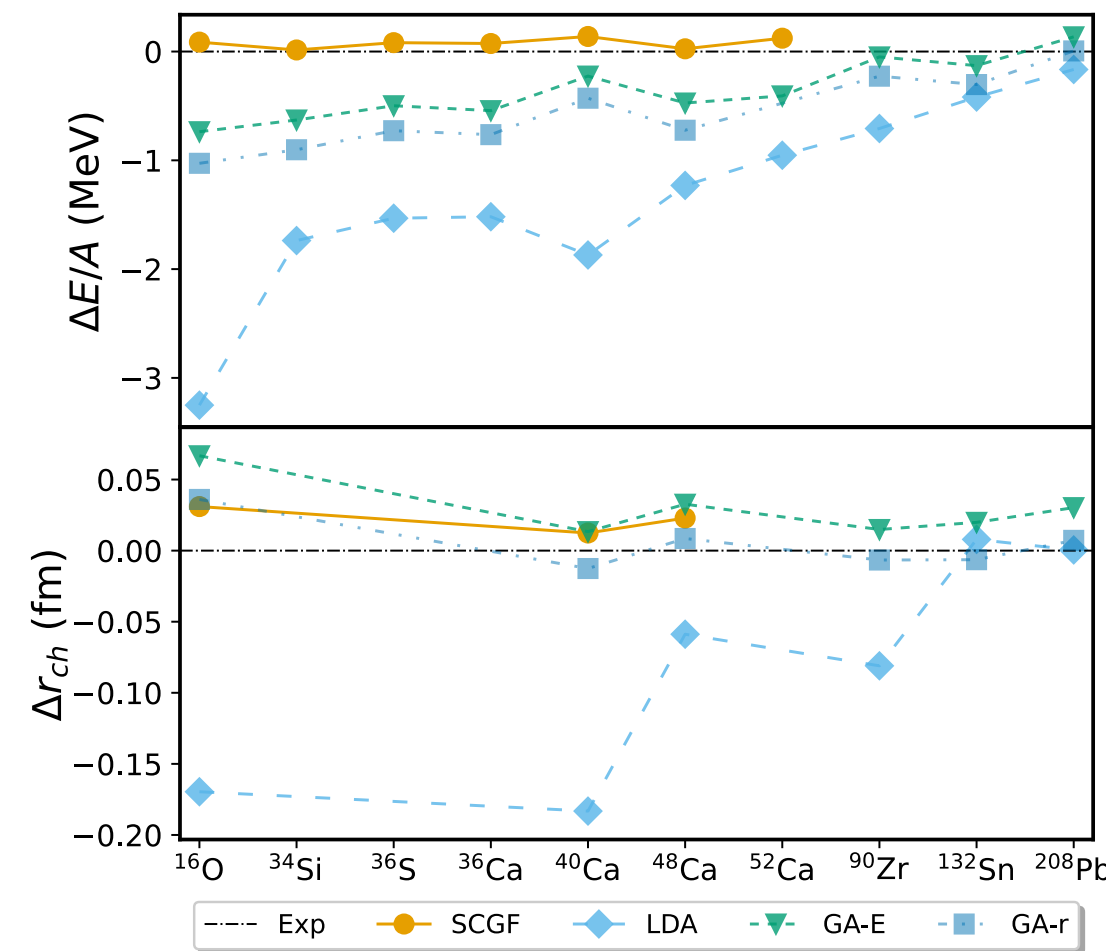
Jacob's ladder



+ approximate GA

Benchmark in finite systems

Gradient terms are important!



Problem: gradients here were added ad hoc to the EDF, so far.

→

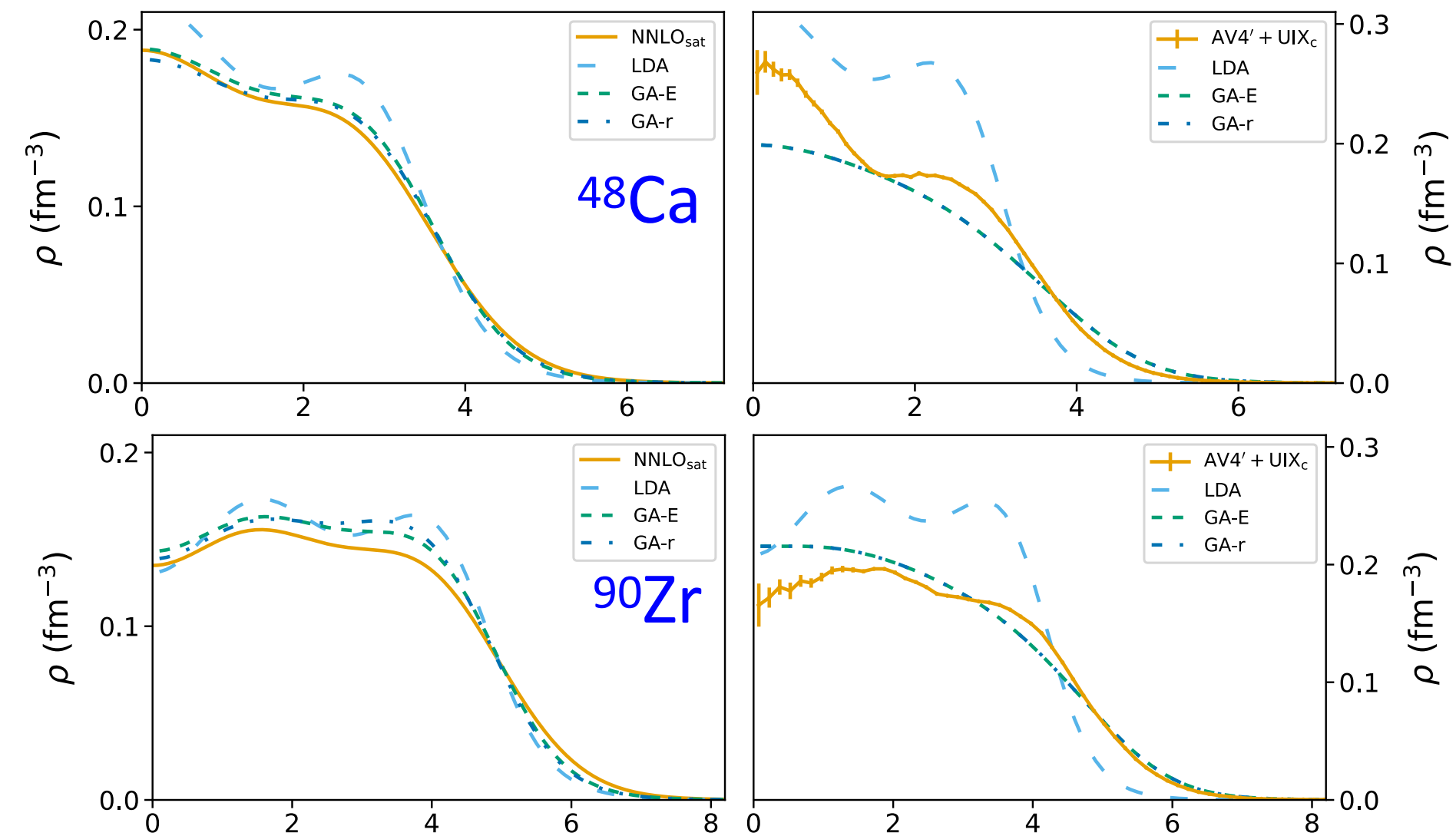
Need to extract gradient information from non-uniform matter. E.g., external (monochromatic) perturbations:

$$v(\mathbf{x}) = v_q e^{i\mathbf{q}\cdot\mathbf{x}} + c.c. = 2v_q \cos(\mathbf{q}\cdot\mathbf{x})$$

$$\delta\rho(\mathbf{x}) = 2\rho_q \cos(\mathbf{q}\cdot\mathbf{x})$$

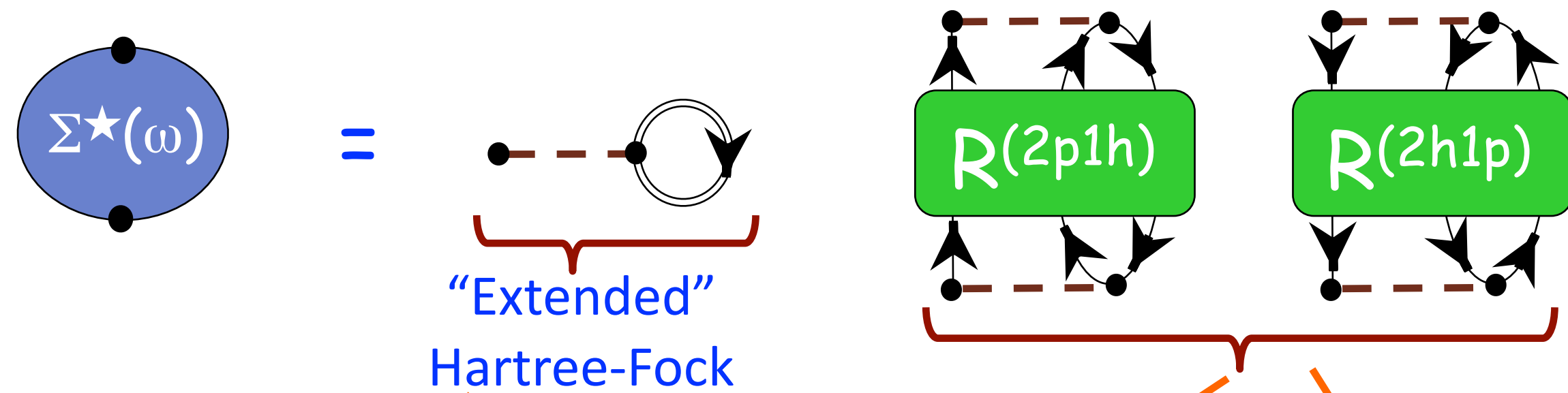
SCGF/NNLO-sat :

AFDMC/AV4' :



# Algebraic Diagrammatic Construction at 3<sup>rd</sup> order [ADC(3)] in a few words

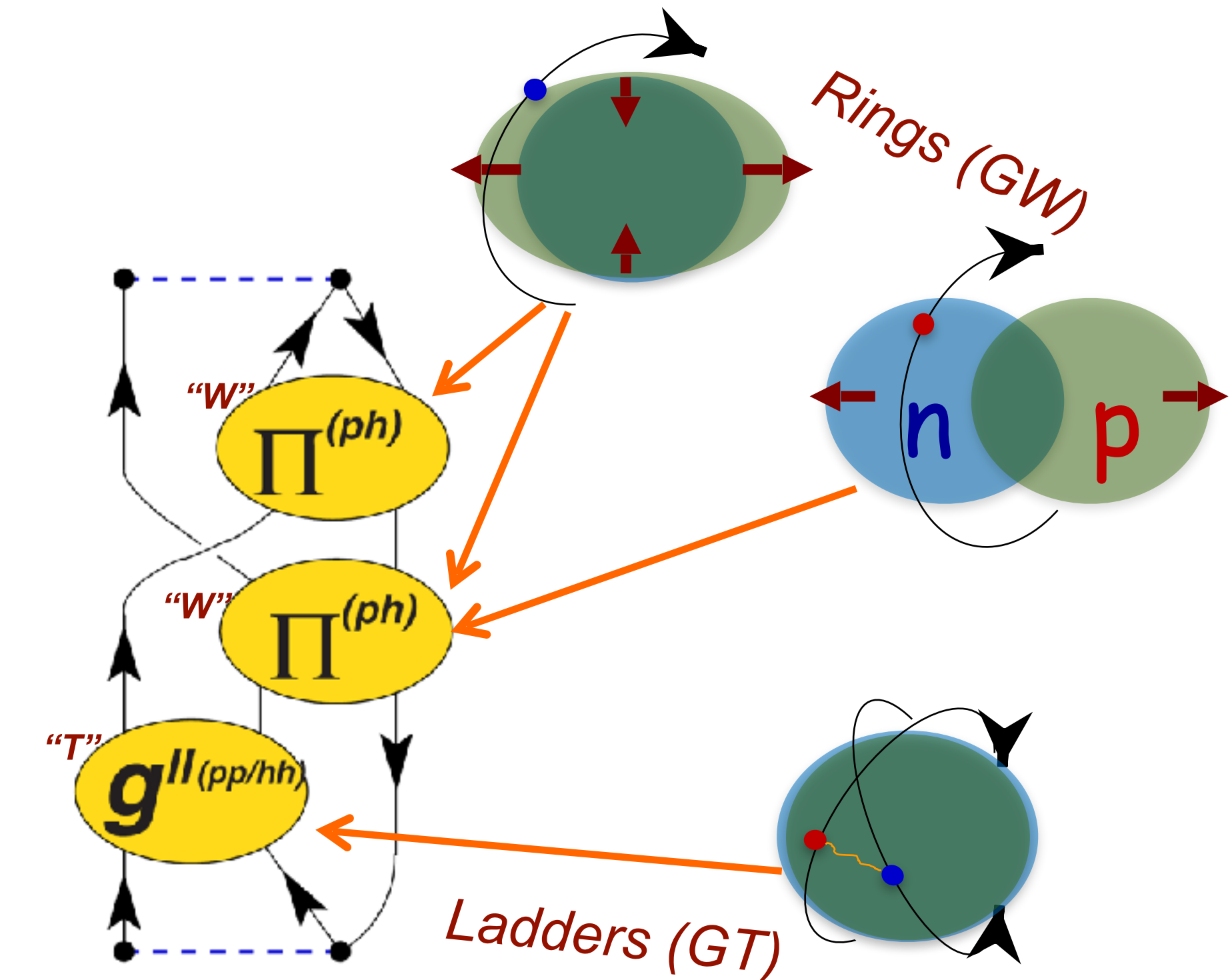
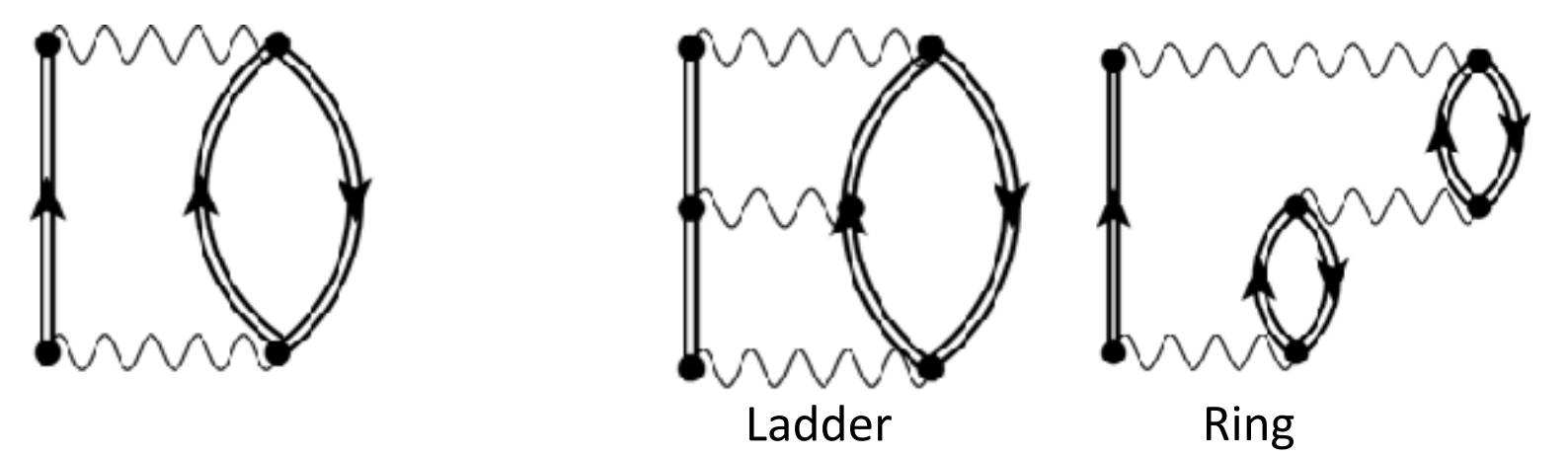
The nuclear self-energy includes both *mean field* and *dynamical correlations* (leading to optical potentials and spectroscopy)



ADC(3): Lect. Notes in Phys 936 (2017) — Chapter 11.

$$\Sigma_{\alpha\beta}^{(\star)}(\omega) = -U_{\alpha\beta} + \Sigma_{\alpha\beta}^{(\infty)} + M_{\alpha,r}^\dagger \left[ \frac{1}{\omega - [E^> + C]_{r,r'} + i\eta} \right]_{r,r'} M_{r',\beta} + N_{\alpha,s} \left[ \frac{1}{\omega - (E^< + D) - i\eta} \right]_{s,s'} N_{s',\beta}^\dagger$$

The ADC(3) truncation level uses diagrams up to 3<sup>rd</sup> order as 'seeds' for all-orders summations:



Self-consistent GF:  $\Sigma^* = \Sigma^* [g(\omega)]$



# Gorkov ansatz... for atomic nuclei

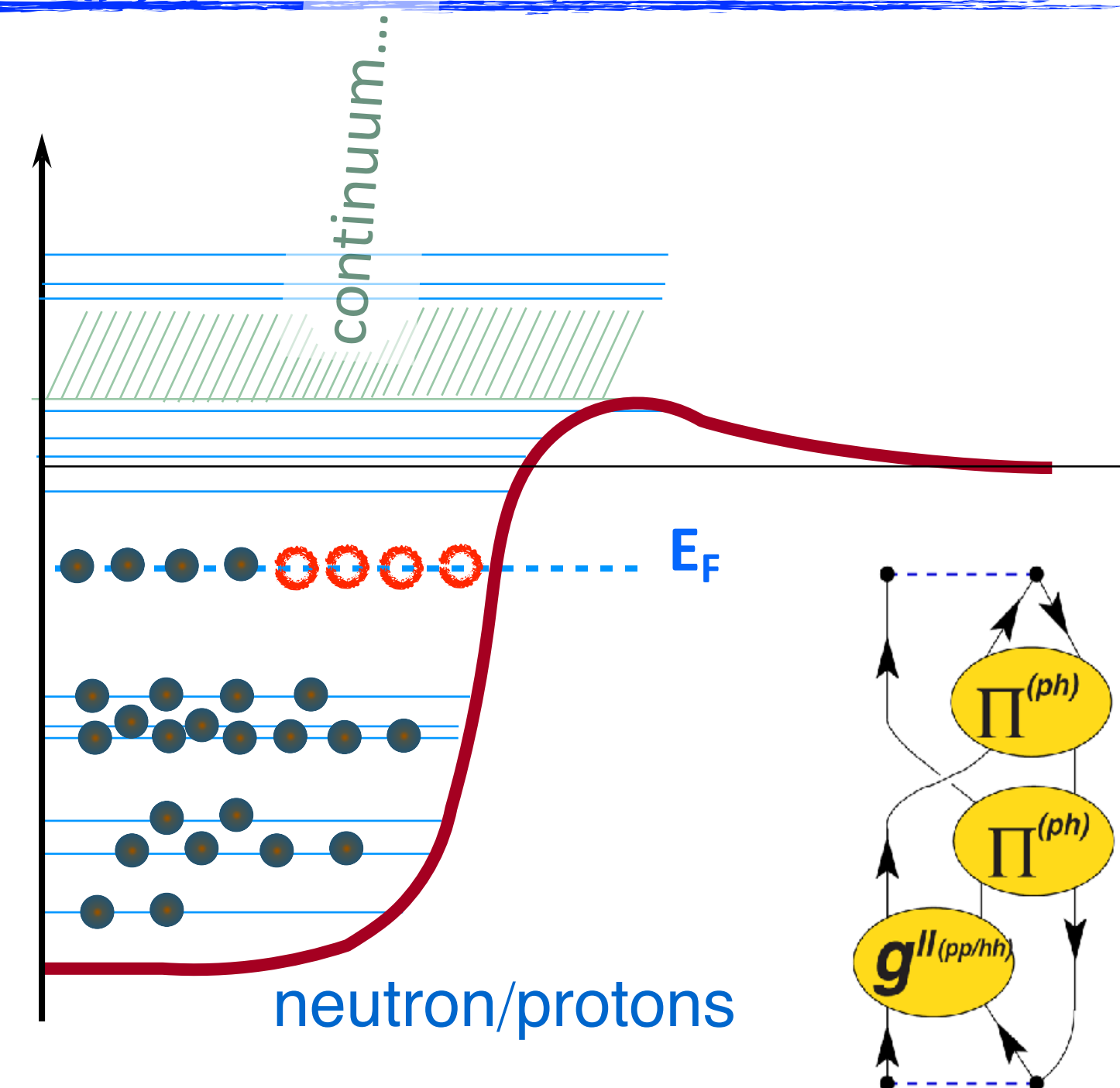
- In the presence of degeneracies (vanishing ph-gaps), enforce (two nucleon) pairing to mitigate instabilities:

- Ansatz many-body state:  $|\Psi_0\rangle = \sum_{n=0}^{\infty} c_{2n} |\psi^{2n}\rangle$

➤ Introduce a “grand-canonical” potential  $\Omega \equiv H - \mu N$

➤  $|\Psi_0\rangle$  minimizes  $\Omega_0 = \min_{|\Psi_0\rangle} \{\langle \Psi_0 | \Omega | \Psi_0 \rangle\}$  under the constraint  $N = \langle \Psi_0 | N | \Psi_0 \rangle$

- Generates a set of two normal and two anomalous self-energies:



$$\Sigma_{\alpha\beta}^*(\omega) = \Sigma_{\alpha\beta}^{\infty} + \tilde{\Sigma}_{\alpha\beta}(\omega) = \begin{pmatrix} \Sigma_{\alpha\beta}^{11,\infty} & \Sigma_{\alpha\beta}^{12,\infty} \\ \left(\Sigma_{\beta\alpha}^{12,\infty}\right)^* & -\Sigma_{\beta\alpha}^{11,\infty} \end{pmatrix} + \begin{pmatrix} \tilde{\Sigma}_{\alpha\beta}^{11}(\omega) & \tilde{\Sigma}_{\alpha\beta}^{12}(\omega) \\ -\left(\tilde{\Sigma}_{\beta\alpha}^{12}(-\omega)\right)^* & -\tilde{\Sigma}_{\beta\alpha}^{11}(-\omega) \end{pmatrix}$$

[Gorkov 1958]

In the limit of particle number restoration:

V. Somà, T. Duguet, CB, Phys. Rev. C **84**, 064317 (2011)  
 V. Somà, CB, T. Duguet, Phys. Rev. C **89**, 024323 (2022)  
 CB, T. Duguet, V. Somà, Phys. Rev. C **105**, 044330 (2022)



# Algebraic diagrammatic construction [ADC(3)] for infinite matter

Finite size box (of length L) with periodic Boundary conditions:

$$\rho = \frac{A}{L} \quad p_F = \sqrt[3]{\frac{6\pi^2\rho}{v_d}} \quad \mathbf{k} = \frac{1}{L} \left( 2\pi\mathbf{n} + \boldsymbol{\theta} \right), \quad |\mathbf{k}|^2 = k_x^2 + k_y^2 + k_z^2 \leq \frac{4\pi^2}{L^2} N_{max}$$

## Gorkov superfluid formulation

$$\mathbf{g}(\omega) = \mathbf{g}^{(0)}(\omega) + \mathbf{g}^{(0)}(\omega) \mathbf{\Sigma}^*(\omega) \mathbf{g}(\omega)$$

Normal self-energy      Pairing field

$$\mathbf{\Sigma}^*(\omega) = \begin{pmatrix} \Sigma^{11}(\omega) & \Delta \\ -\Delta^* & \Sigma^{22}(\omega) \end{pmatrix}$$



$$\left( \begin{array}{ccc|ccc} T+\Lambda-1\mu_q & M^\dagger & N & & & \\ M & E^>+C-1\mu_q & & \Delta & & \\ N^\dagger & & E^<+D-1\mu_q & & & \\ \hline \Delta^\dagger & & & -(T+\Lambda)^*+1\mu_q & -M^T & -N^* \\ & & & -M^* & -(E^>+C)^*+1\mu_q & \\ & & & -N^T & & -(E^<+D)^*+1\mu_q \end{array} \right) \begin{pmatrix} \mathcal{U}^q \\ \mathcal{W}^q \\ \mathcal{Z}^q \\ \mathcal{V}^q \\ \mathcal{R}^q \\ \mathcal{S}^q \end{pmatrix} = \hbar\omega_q \begin{pmatrix} \mathcal{U}^q \\ \mathcal{W}^q \\ \mathcal{Z}^q \\ \mathcal{V}^q \\ \mathcal{R}^q \\ \mathcal{S}^q \end{pmatrix}$$

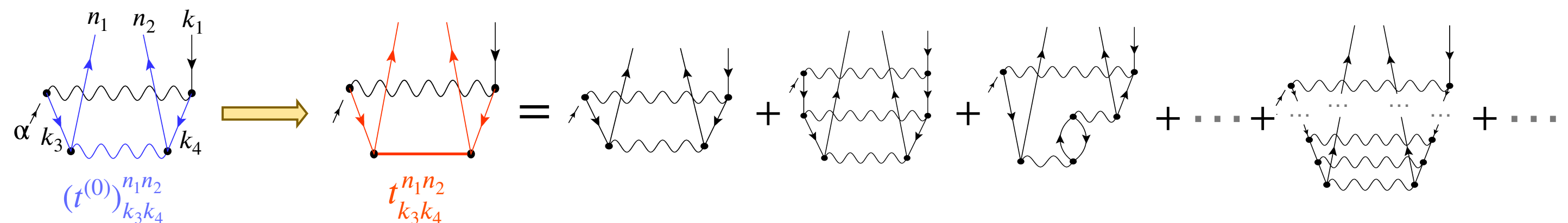
Normal self-energy in full ADC(3) but use pairing only at 1st order

Extension of ADC(3)



ADC(3)-D

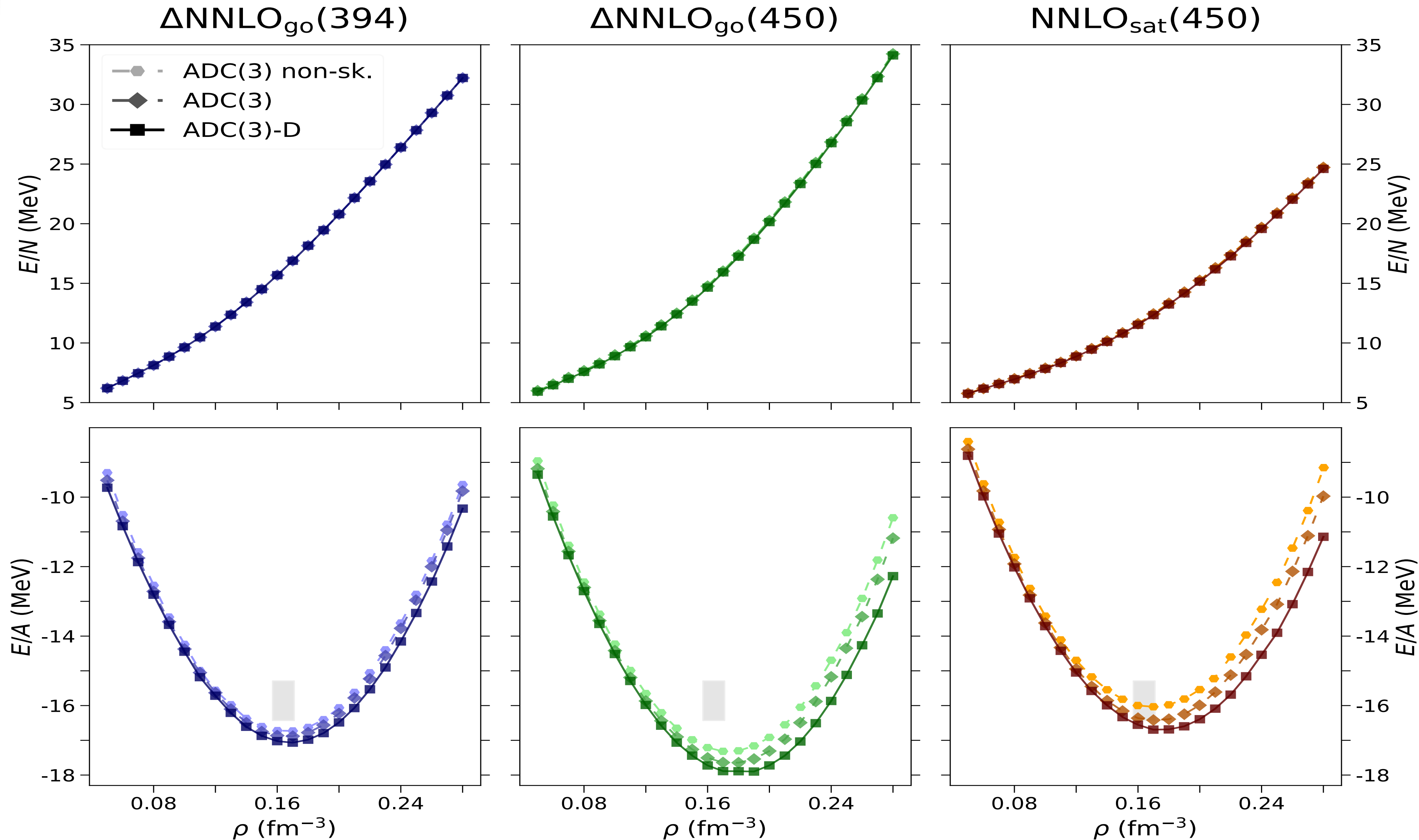
In  $\tilde{\Sigma}(\omega)$ , replace  $t^{(0)}$  with converged  $T_2$  amplitudes



$$(t^{(0)})_{ij}^{ab} = \frac{\langle ab|v|ij\rangle_A}{\epsilon_i + \epsilon_j - \epsilon_a - \epsilon_b}$$



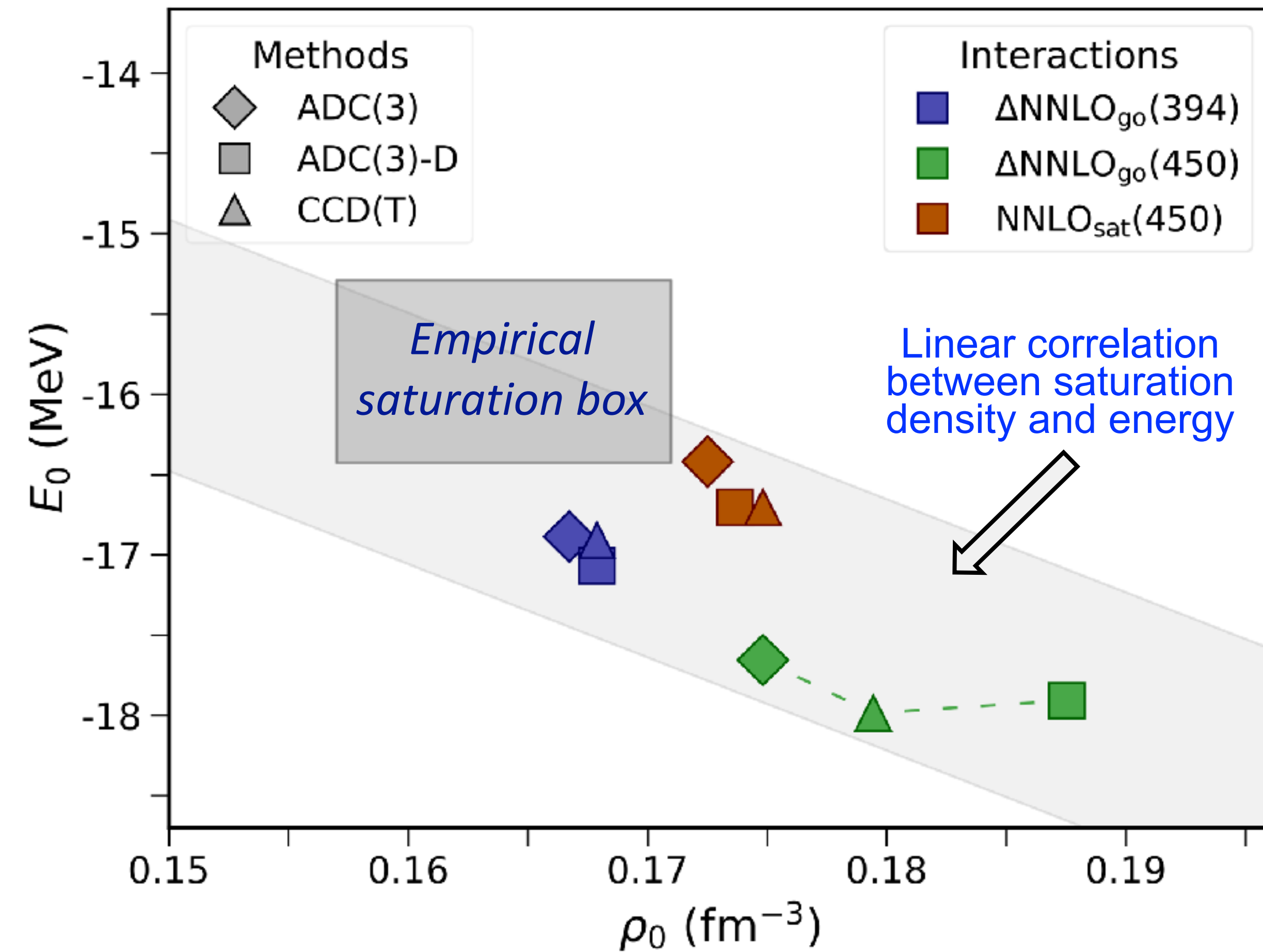
# Equations of state ( $T=0$ )



EOS available at:

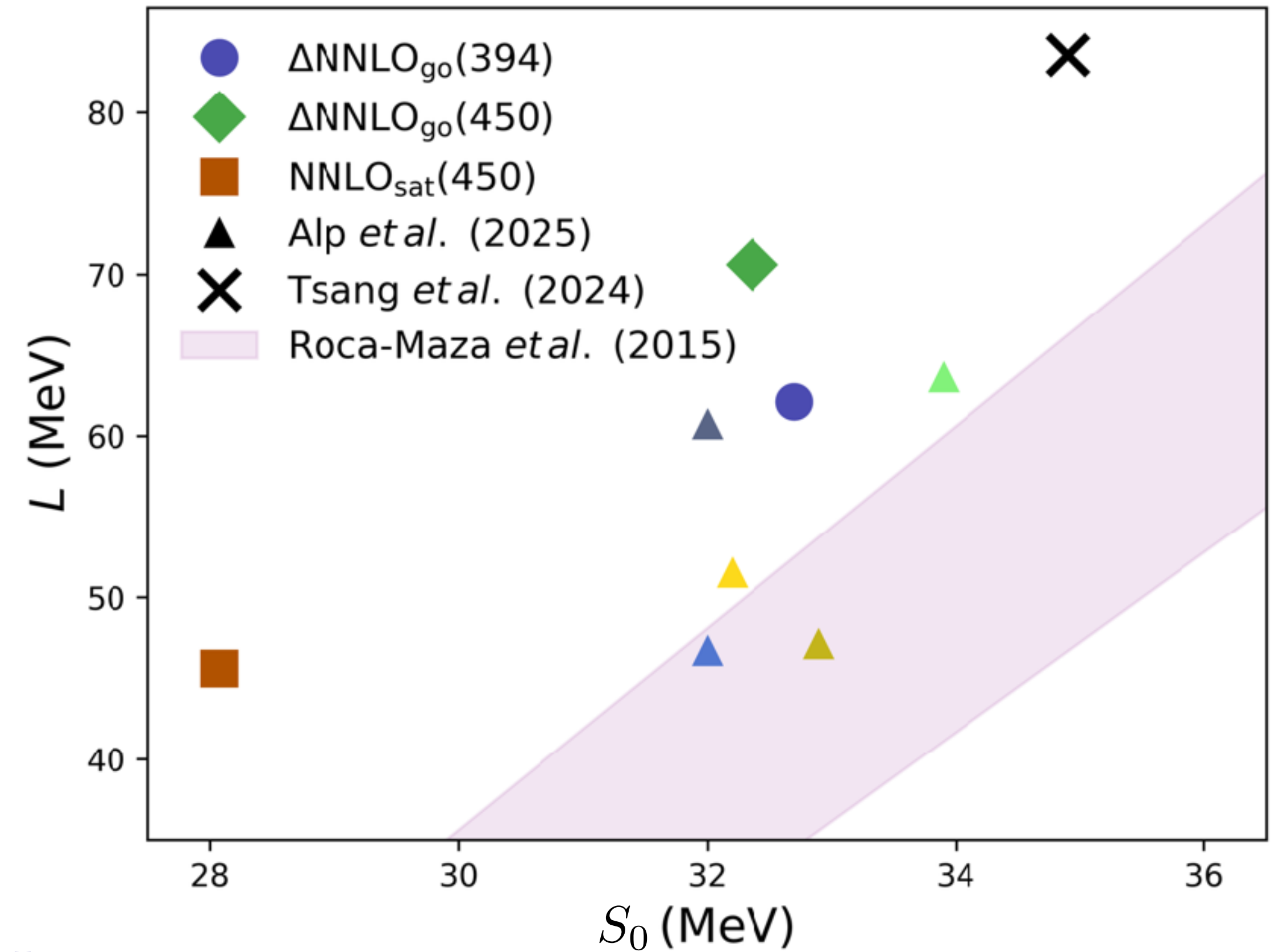
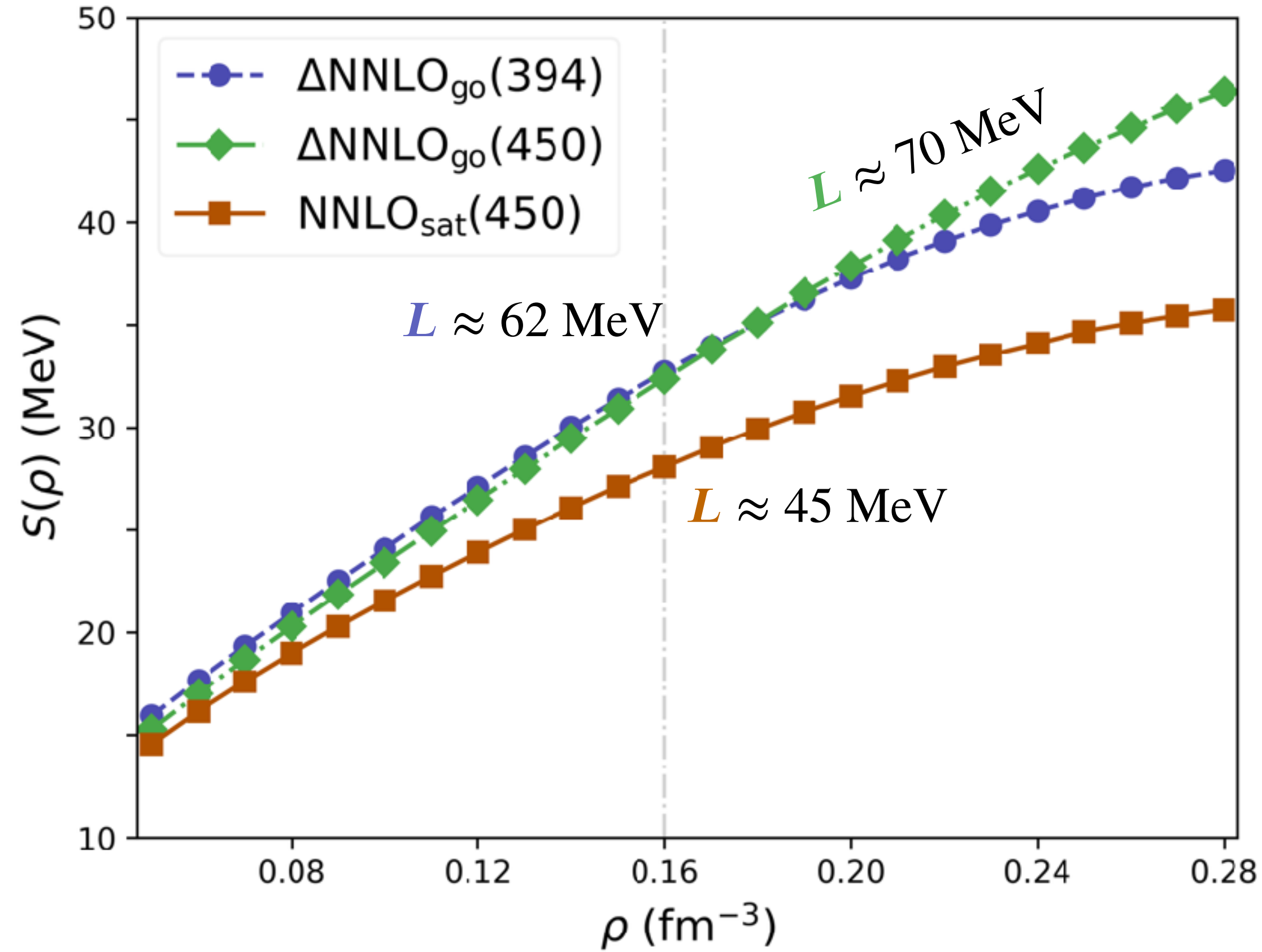


# Saturation point of symmetric nuclear matter



- ◆ Generally good agreement across many-body methods
- ◆ **Interaction-dependence** is the dominant uncertainty

# Symmetry energy



$$S(\rho) = S(\rho_0) + \frac{L}{3\rho_0}(\rho - \rho_0) + \dots$$

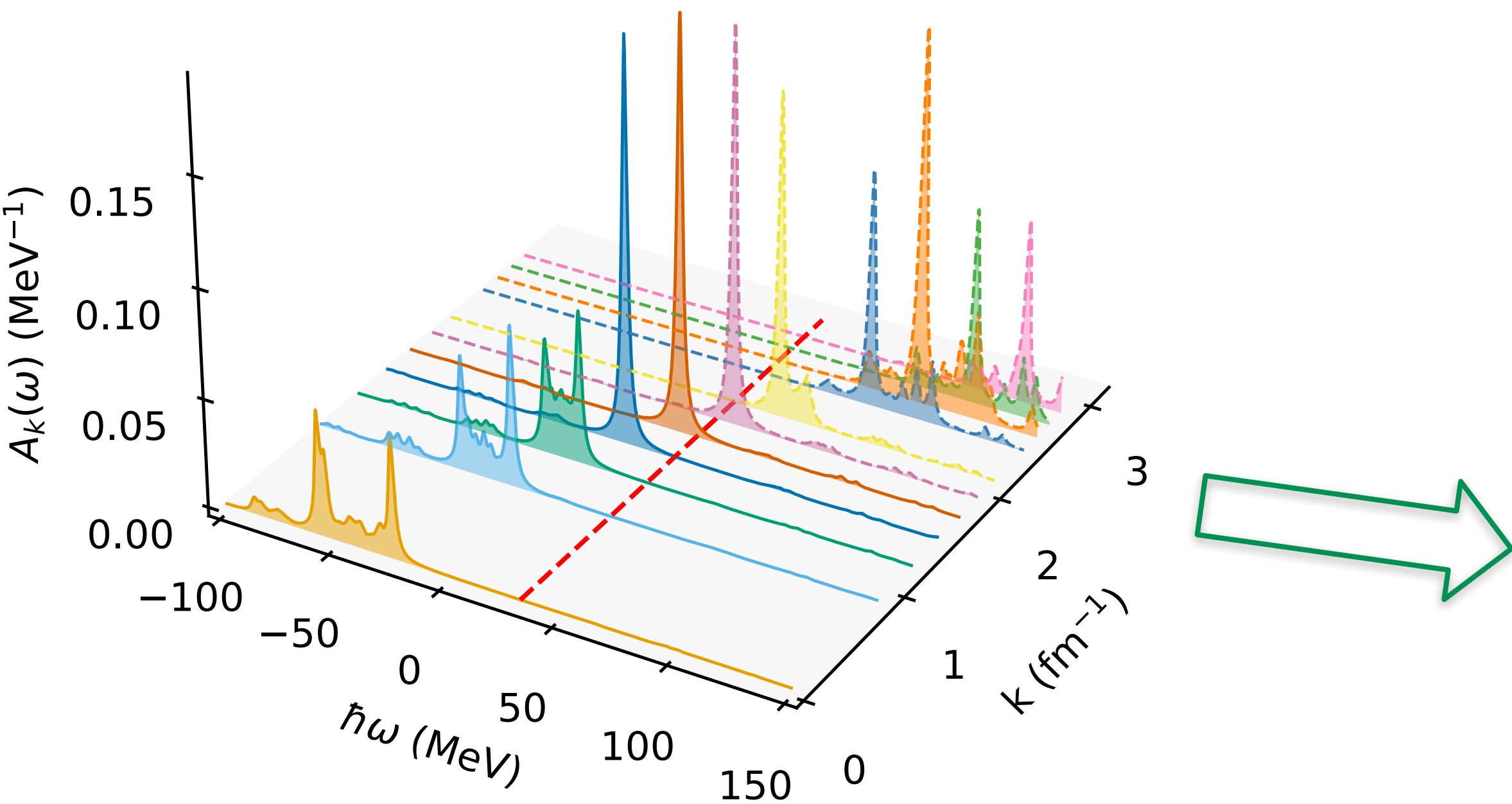
F. Marino, CB, and G. Colò, arXiv:2601.03763 [nucl-th]

Alp *et al.*, arXiv:2504.18259  
 Burgio *et al.*, Front. Astron. Space Sci. **11**, 1505560 (2024)  
 Lynch and Tsang, Phys. Lett. B **830**, 137098 (2022)  
 Roca-Maza *et al.*, Phys. Rev. C **92**, 064304 (2015)



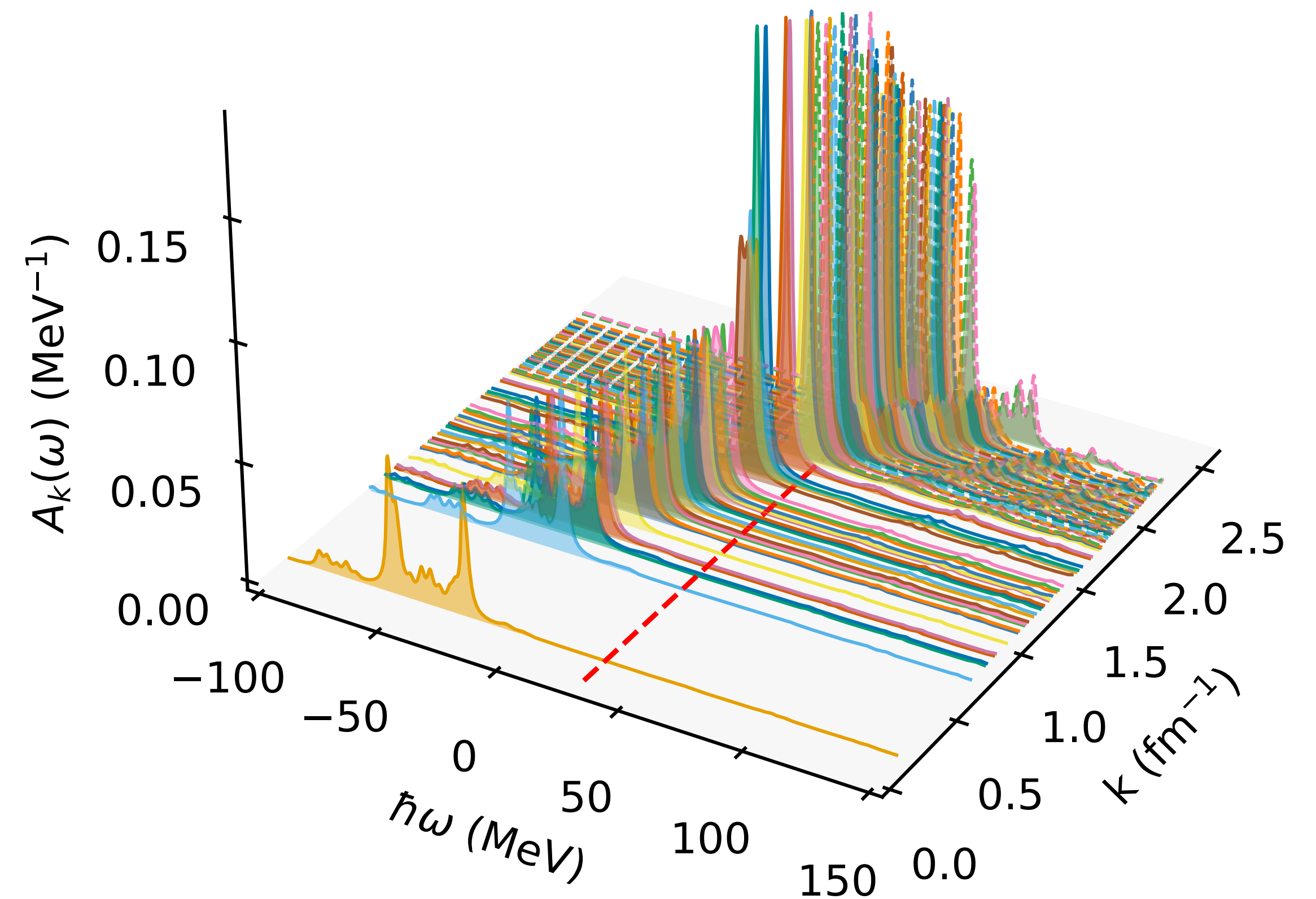
# Removing finite size effects - Twisted Angle BC

## Ordinary PBC



## sp-TABC: chose just one set of twisted angles

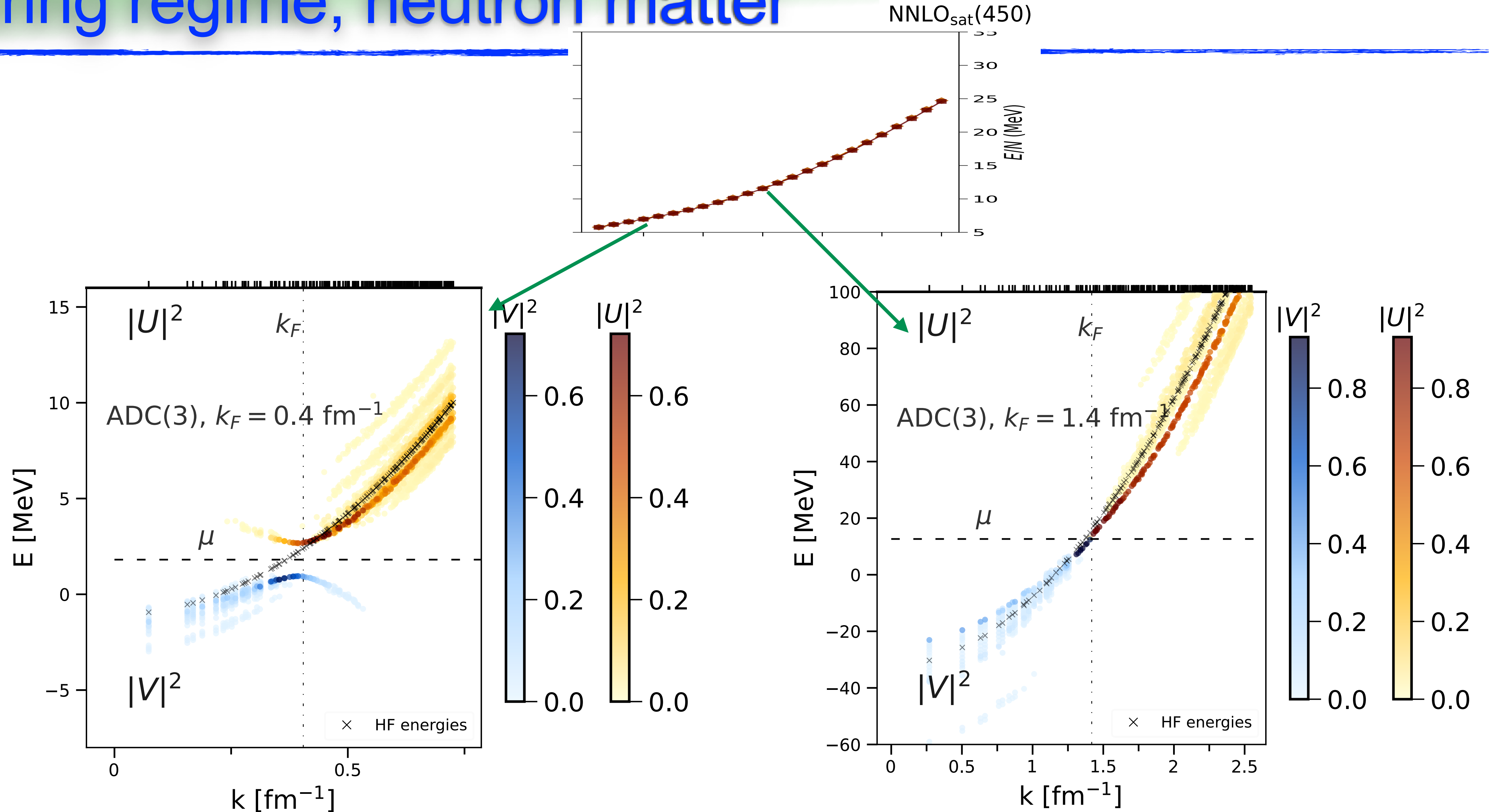
- Finer resolution in momenta
- Better description of the thermodynamic Limit



$$S^p(\mathbf{p}, \omega) = \sum_n |\langle \Psi_n^{A+1} | \psi_{\mathbf{p}}^\dagger | \Psi_0^A \rangle|^2 \delta(\hbar\omega - (E_n^{A+1} - E_0^A))$$

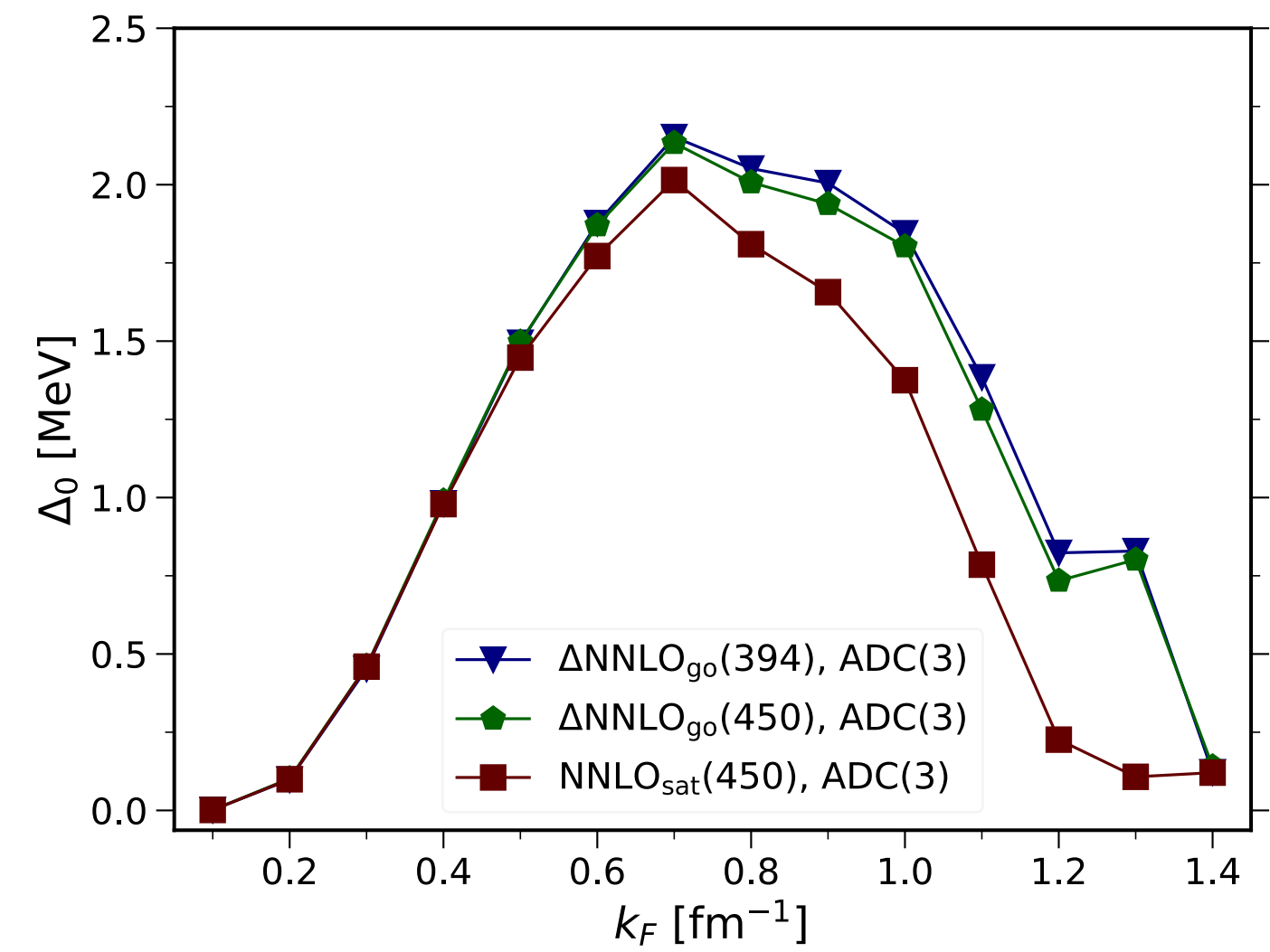
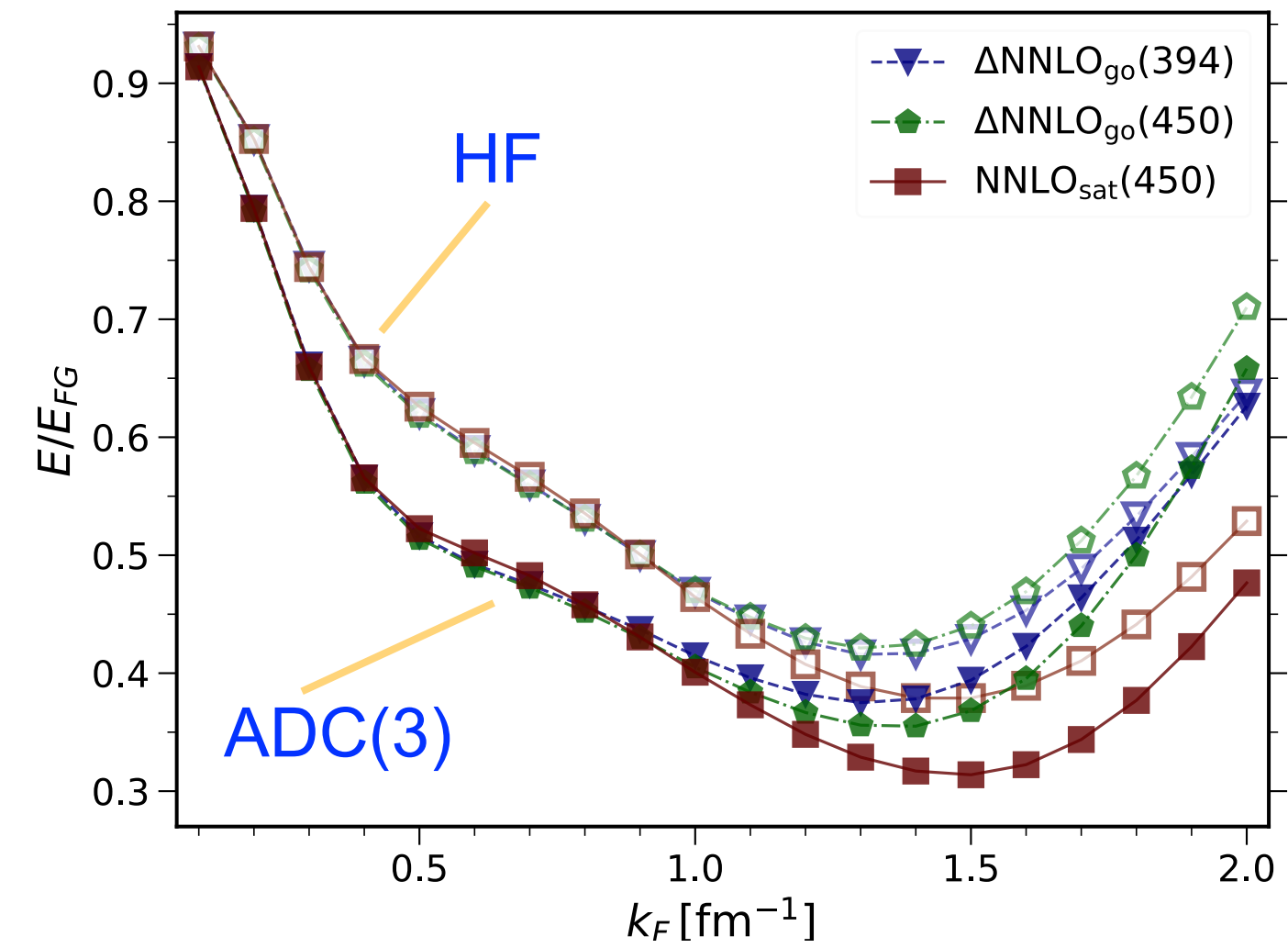
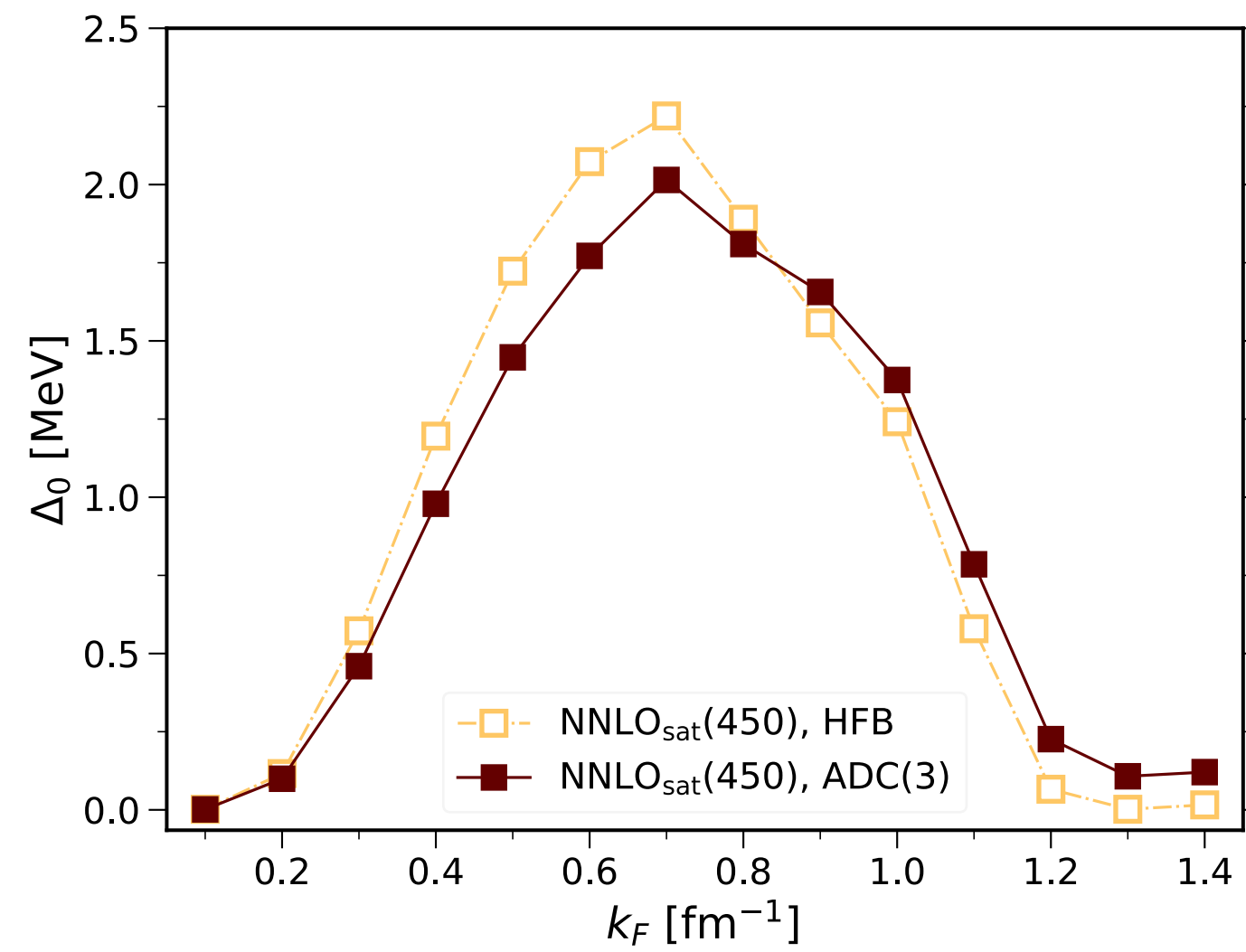
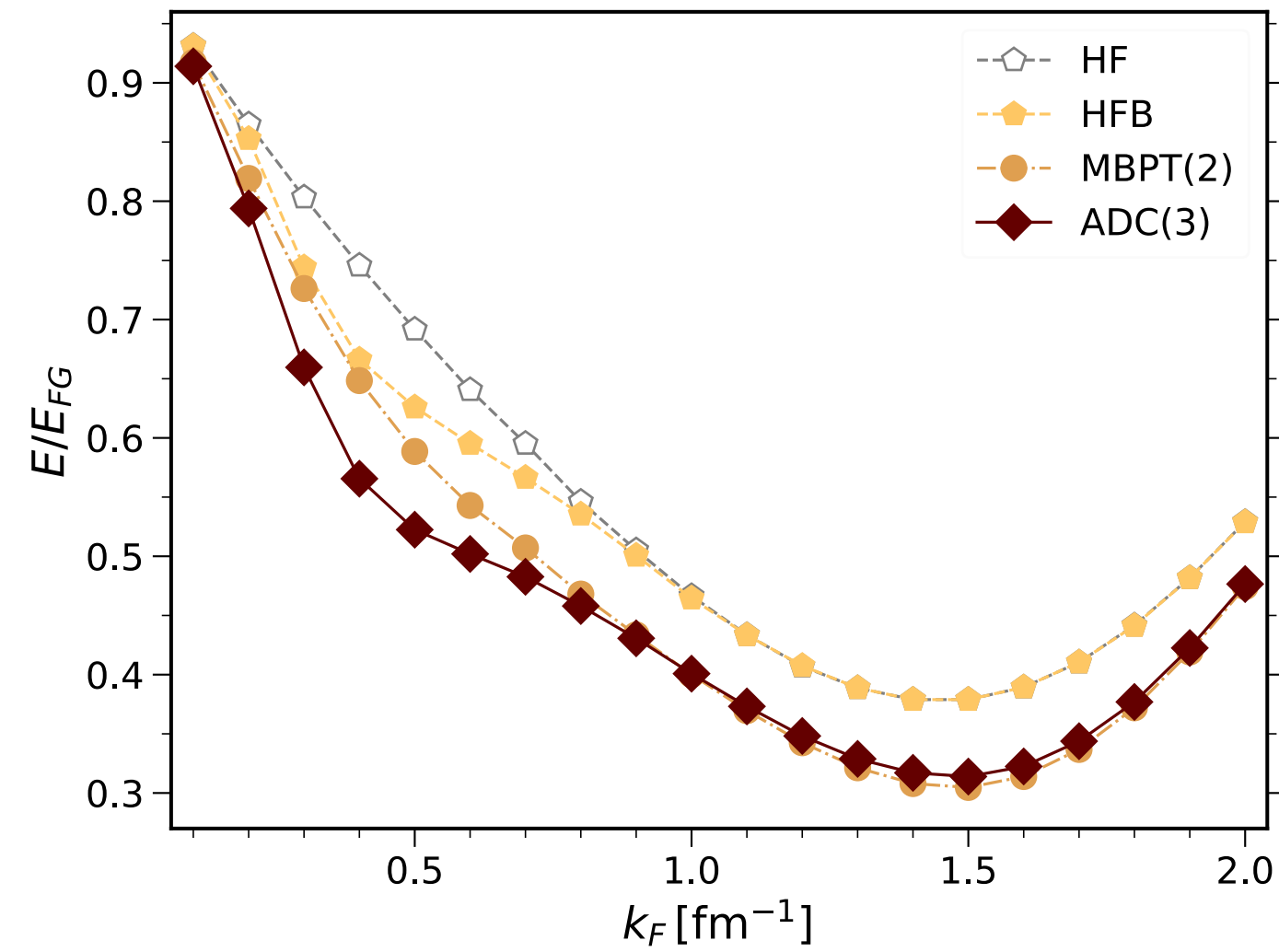
$$S^h(\mathbf{p}, \omega) = \sum_k |\langle \Psi_k^{A-1} | \psi_{\mathbf{p}} | \Psi_0^A \rangle|^2 \delta(\hbar\omega - (E_0^A - E_k^{A-1}))$$

# Pairing regime, neutron matter



# Pairing in neutron matter

NNLO<sub>sat</sub>(450), sp-TABC



# Green's function theory beyond ADC(3)?

# DiagMC !!

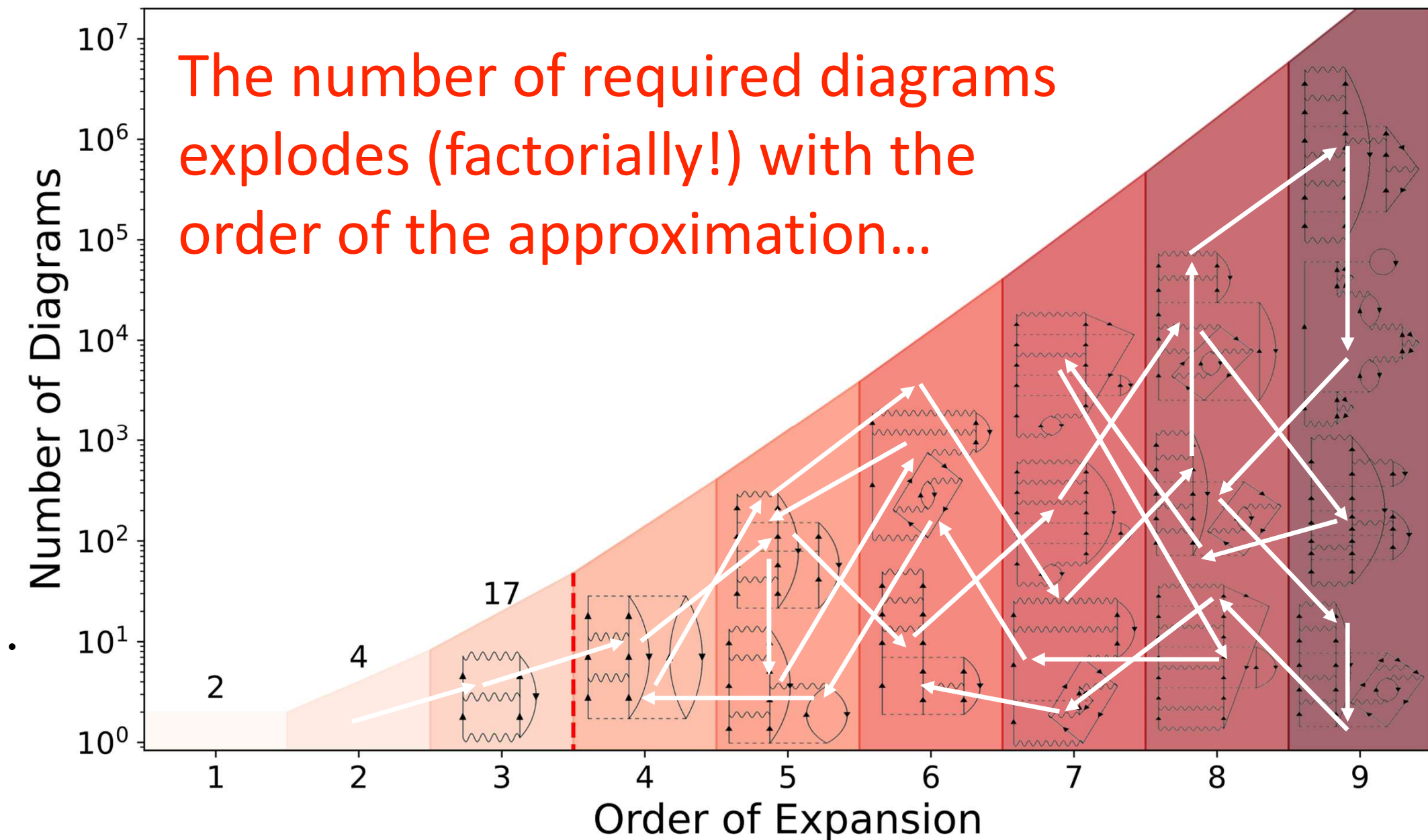
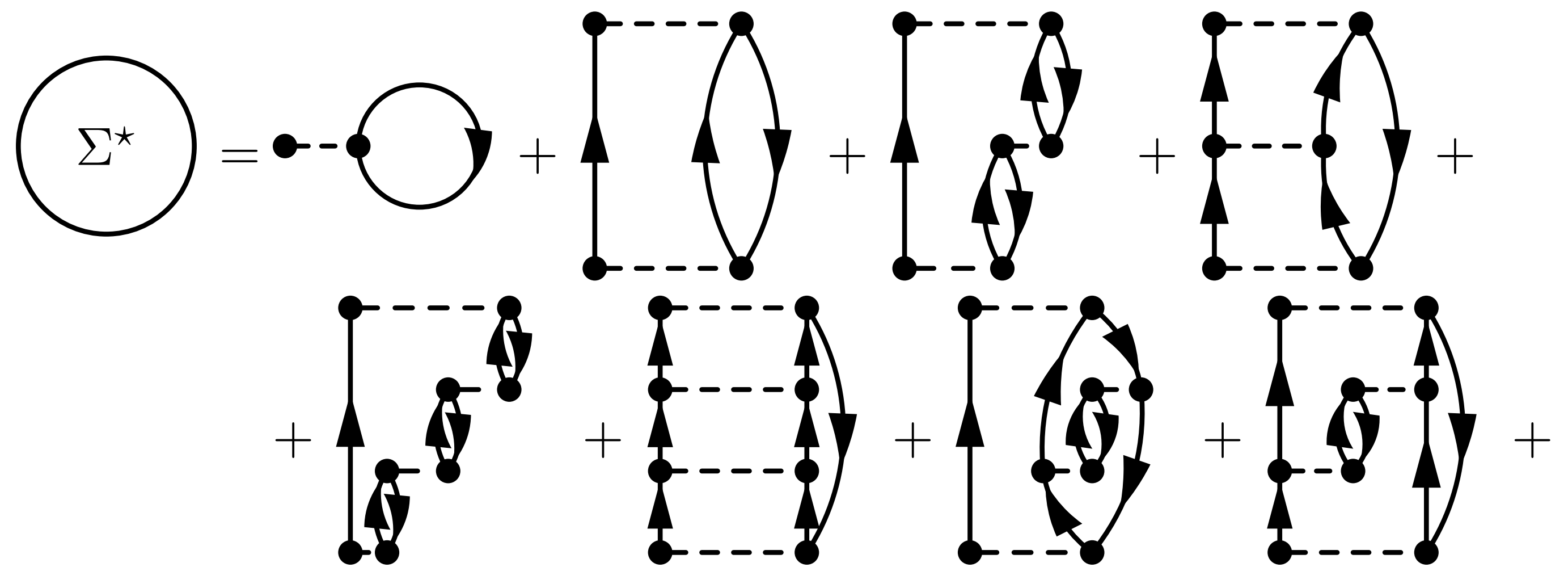
The Green's function is found as the exact solution of the Dyson equation:

$$G_{\alpha\beta}(\omega) = G_{\alpha\beta}^{(0)}(\omega) + \sum_{\gamma\delta} G_{\alpha\gamma}^{(0)}(\omega) \Sigma_{\gamma\delta}^*(\omega) G_{\delta\beta}(\omega)$$

S. Brolli  
(PhD candidate,  
Milan)



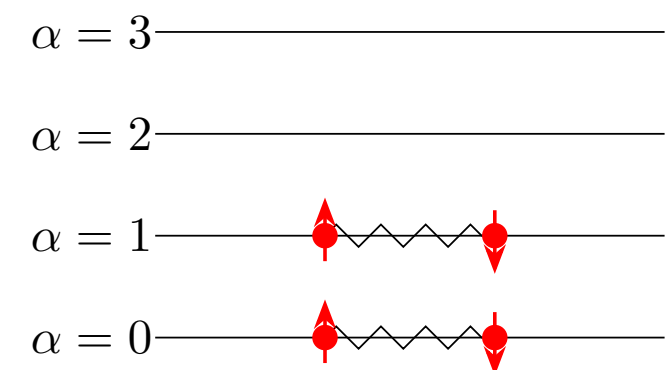
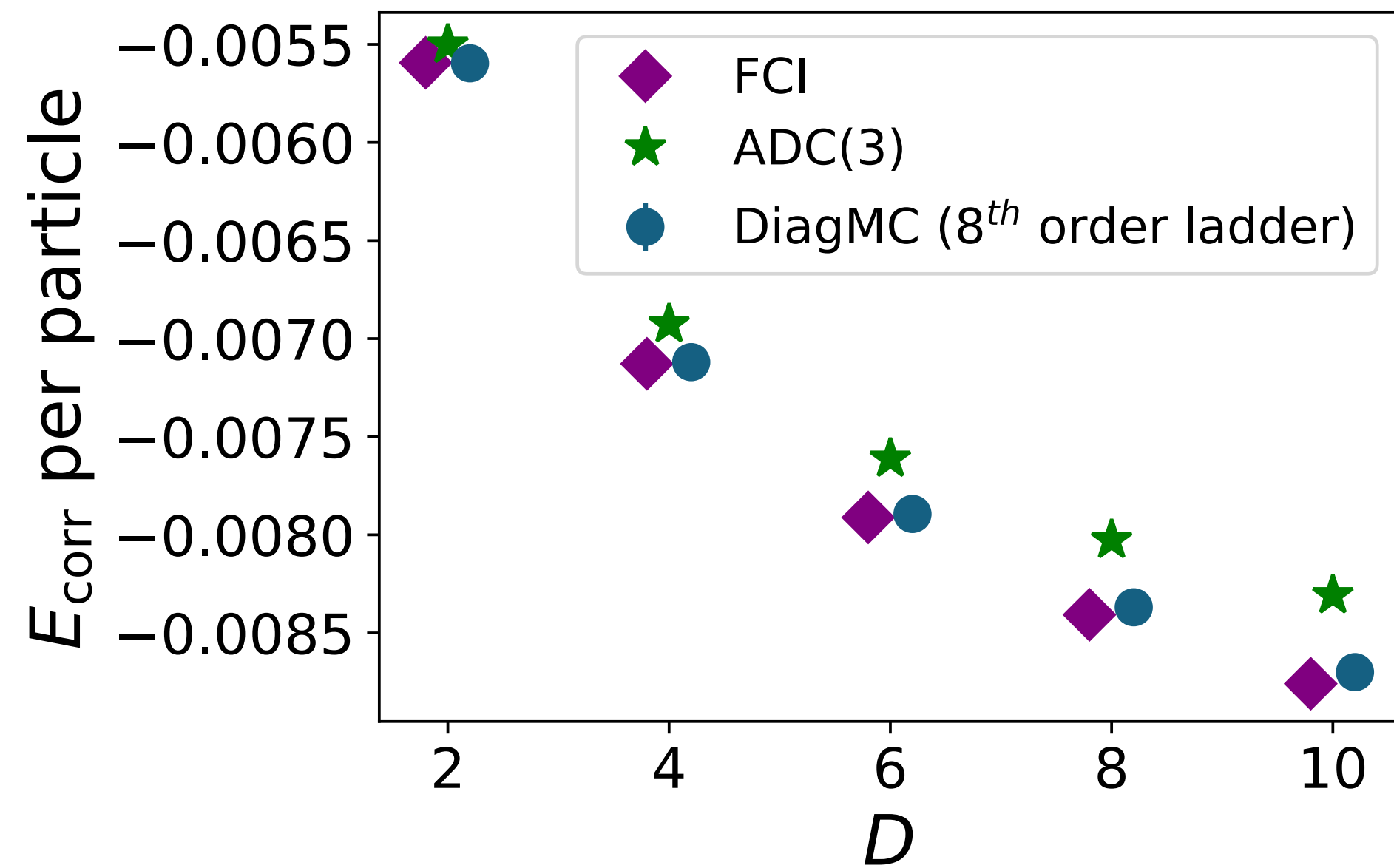
It requires knowing the self-energy which is the sum of an *infinite series* of Feynman diagrams:



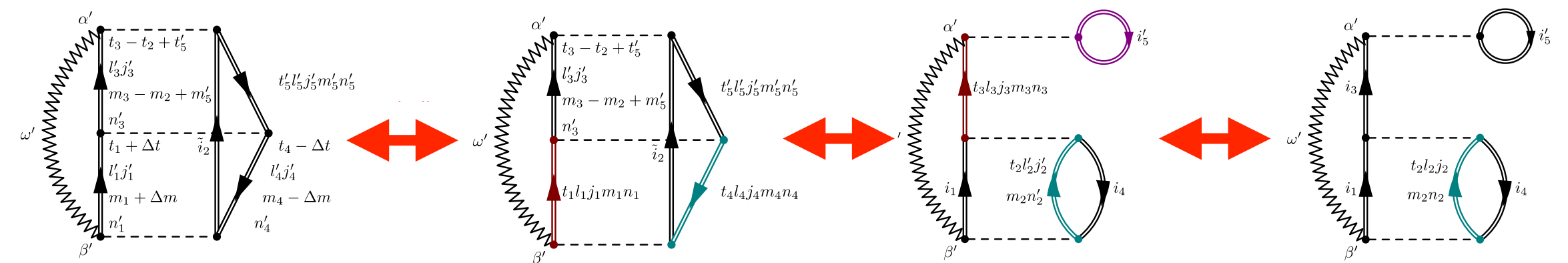
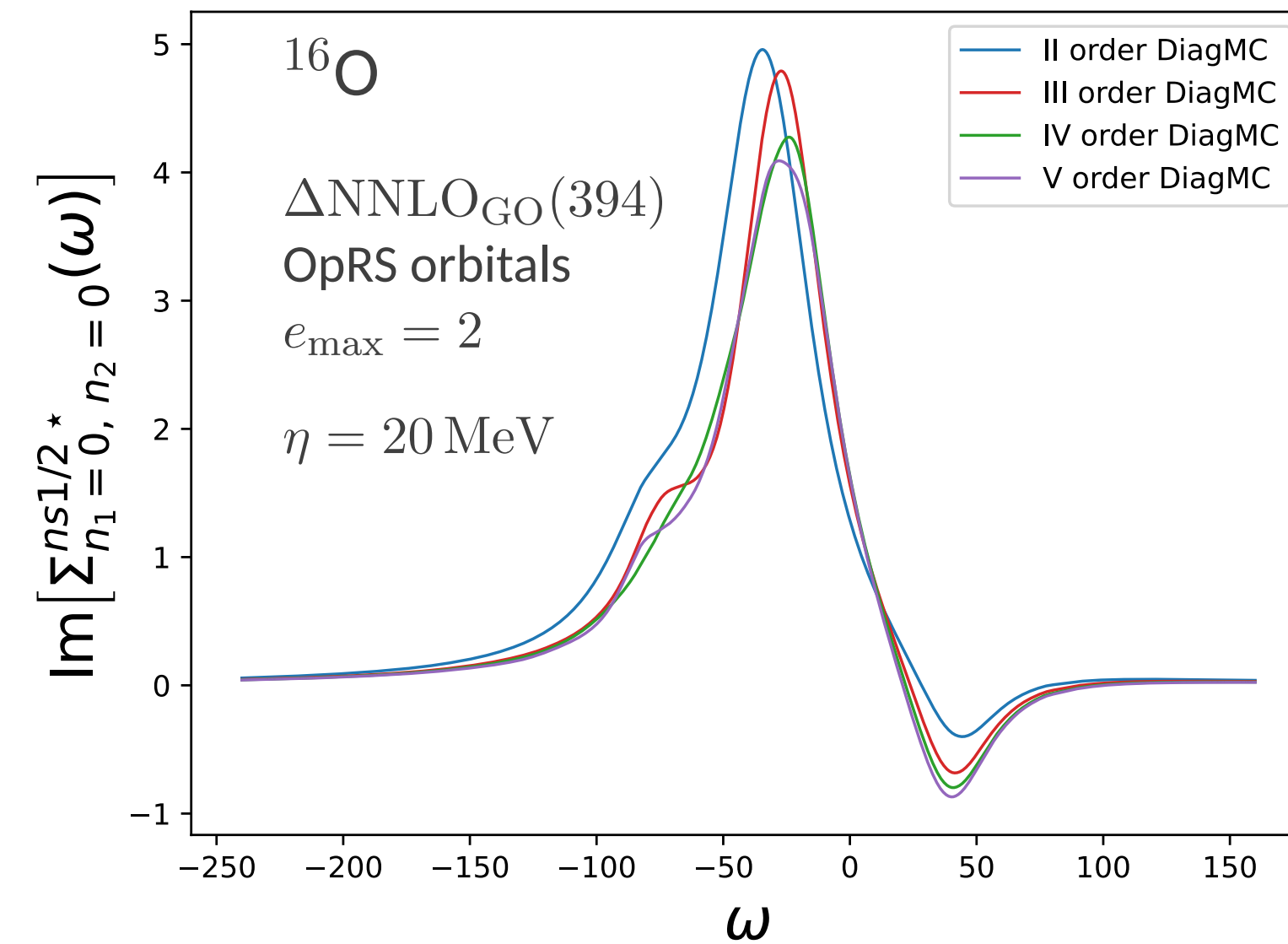
Diagrammatic Monte Carlo (DiagMC) *samples diagrams in their topological space* using a Markov chain.

[Brolli, CB, Vigezzi, Phys. Rev. Lett. **134**, 182502 (2025)]

## Richardson pairing model model up to $D=10$ levels:



## Up to 5th order calculations are now possible in nuclei:



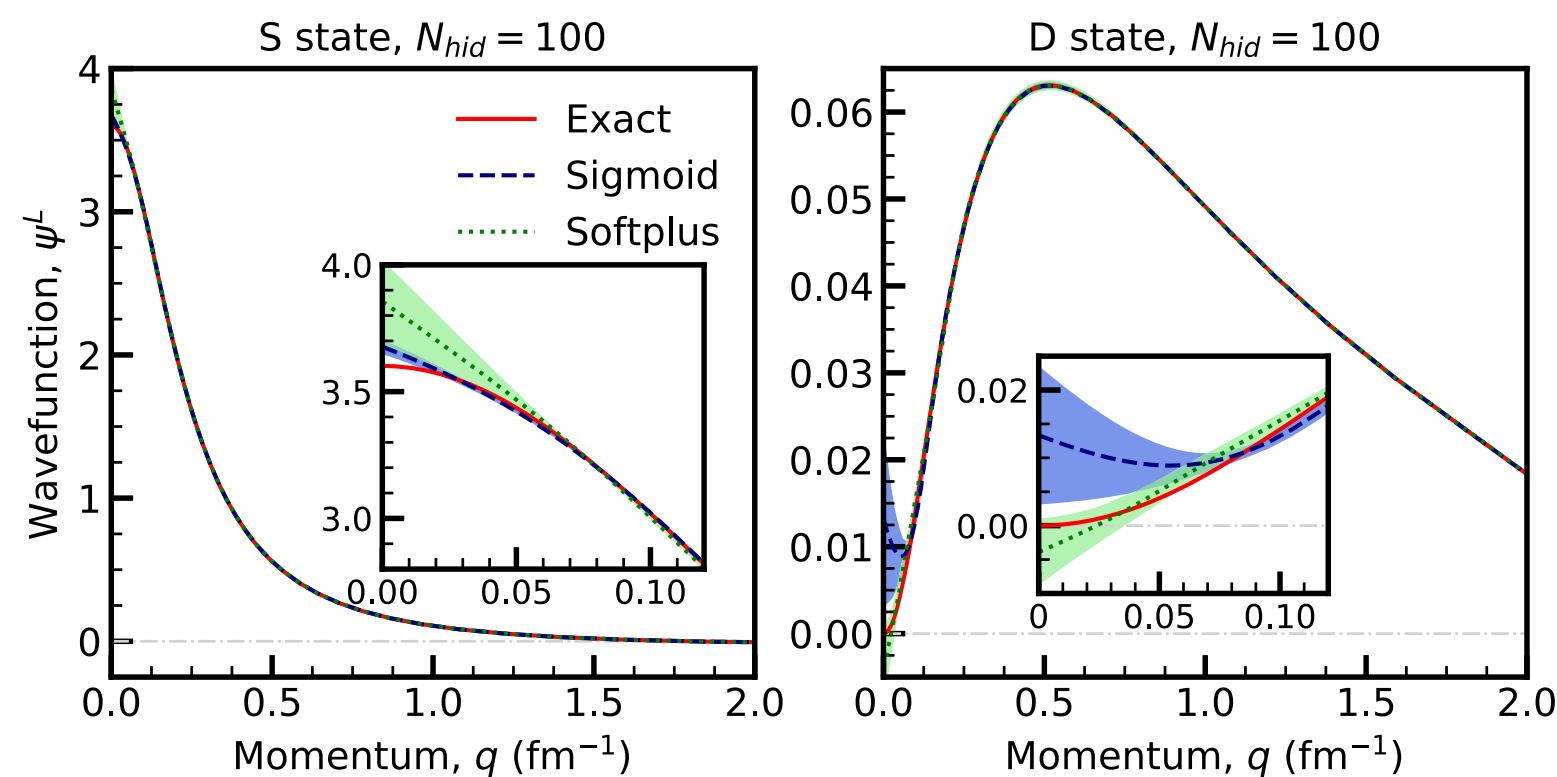
# Neural Network Quantum States (NQS):

$$\frac{\langle \Psi_V | H | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle} = E_V \geq E_0 \quad |\Psi_V^J\rangle = \prod_{i < j < k} \left( 1 - \sum_{\text{cyc}} u(r_{ij}) u(r_{jk}) \right) \prod_{i < j} f(r_{ij}) |\Phi\rangle$$

In short: a VMC with a NN trial wave function!

neural networks

The  
deuteron:



Light nuclei:

	$\Lambda$	VMC-ANN	VMC-JS	GFMC	GFMC <sub>c</sub>
<sup>2</sup> H	4 fm <sup>-1</sup>	-2.224(1)	-2.223(1)	-2.224(1)	-
	6 fm <sup>-1</sup>	-2.224(4)	-2.220(1)	-2.225(1)	-
<sup>3</sup> H	4 fm <sup>-1</sup>	-8.26(1)	-7.80(1)	-8.38(2)	-7.82(1)
	6 fm <sup>-1</sup>	-8.27(1)	-7.74(1)	-8.38(2)	-7.81(1)
<sup>4</sup> He	4 fm <sup>-1</sup>	-23.30(2)	-22.54(1)	-23.62(3)	-22.77(2)
	6 fm <sup>-1</sup>	-24.47(3)	-23.44(2)	-25.06(3)	-24.10(2)

Keeble, Rios, Phys. Lett. B **809**, 135743 (2020)

Rozalén Sarmiento, Keeble, EPJ Plus **139**, 189 (2024)

C. Adams, G. Carleo, A. Lovato, N. Rocco,  
Phys. Rev. Lett. **127**, 022502 (2021)



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In short: a VMC with a NN trial wave function!

*neural networks*

- **NQS often surpass traditional VMC**, with accuracy similar to GFMC/DMC but a a fraction of the cost
- Sometimes referred as “**science driven learning**”: it is not supervised and the cost function is the variational principle.
- However: the **variational principle is the only guiding principle** exploited, what about the rest??

➔ Use Physics Informed Neural Networks (PINN) !

# Physics Informed Neural Networks

- ◆ Generally used for solving ODE and PDE: NN represent the solution  $f(\mathbf{x})$ ,  $\partial f(\mathbf{x})/\partial x_i, \dots$
- ◆ General PINN cost function:

$$\mathcal{L}_{PINN} = \mathcal{L}_{PDE} + \mathcal{L}_{phys} + \mathcal{L}_{data}$$

constraint from  
diff. equations

Physics constraints  
(e.g. boundary conditions)

Any known (measured) data

# Physics Informed Neural Networks for nuclei

- ◆ The Schrödinger problem is an integro-differential eq.: use NN to represent  $\Psi(\mathbf{x}_1, \mathbf{x}_2, \dots)$
- ◆ Many-body PINN cost function:

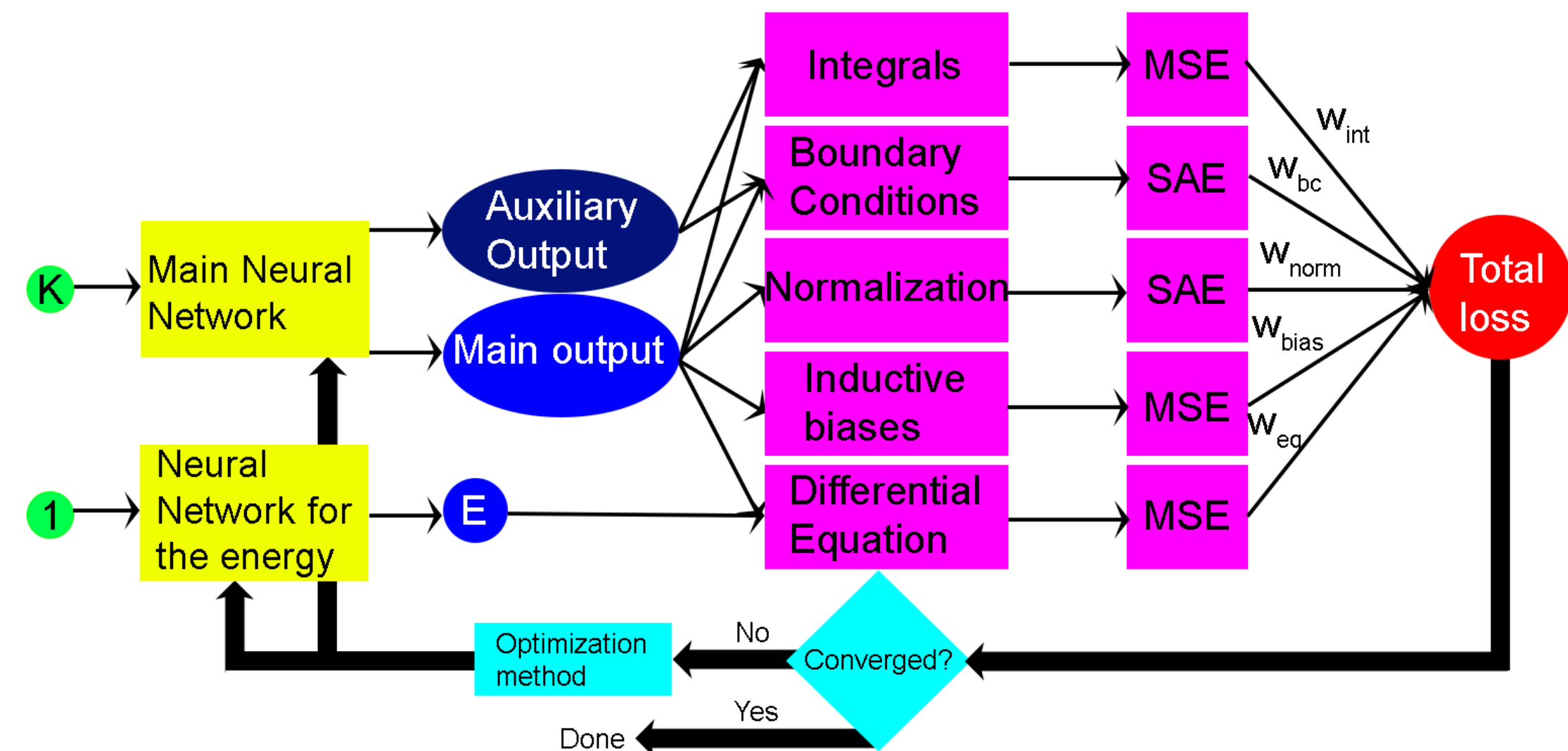
$$\mathcal{L}_{PINN} = \mathcal{L}_{PDE} + \mathcal{L}_{phys} + \cancel{\mathcal{L}_{data}}$$

Schrod. eq.

$$\mathcal{L}_{PDE} = \sum_{i=0}^{N_c} |[H\psi](x_i) - E\psi(x_i)|^2 \quad \text{with} \quad E = \frac{\langle \psi | H | \psi \rangle}{\mathcal{I}[\psi]}$$

Physics constraints

$$\mathcal{L}_{phys} = w_{int} \mathcal{L}_{int} + w_{BCs} \mathcal{L}_{BCs} + w_{norm} \mathcal{L}_{norm} + w_{var} \mathcal{L}_{var} + \dots$$



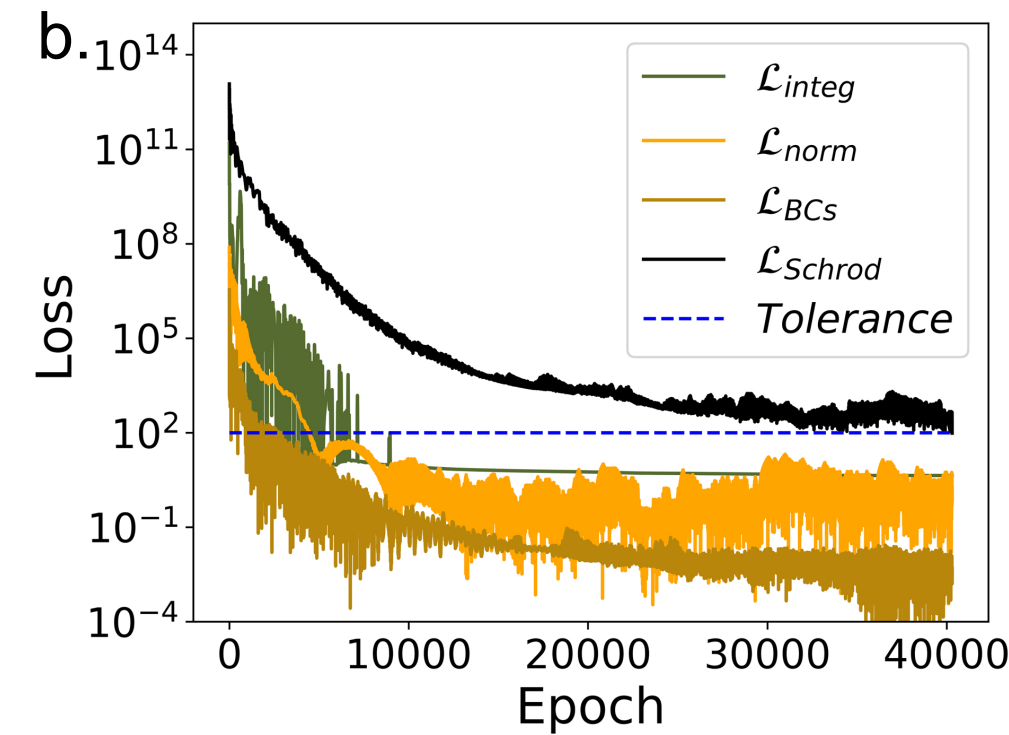
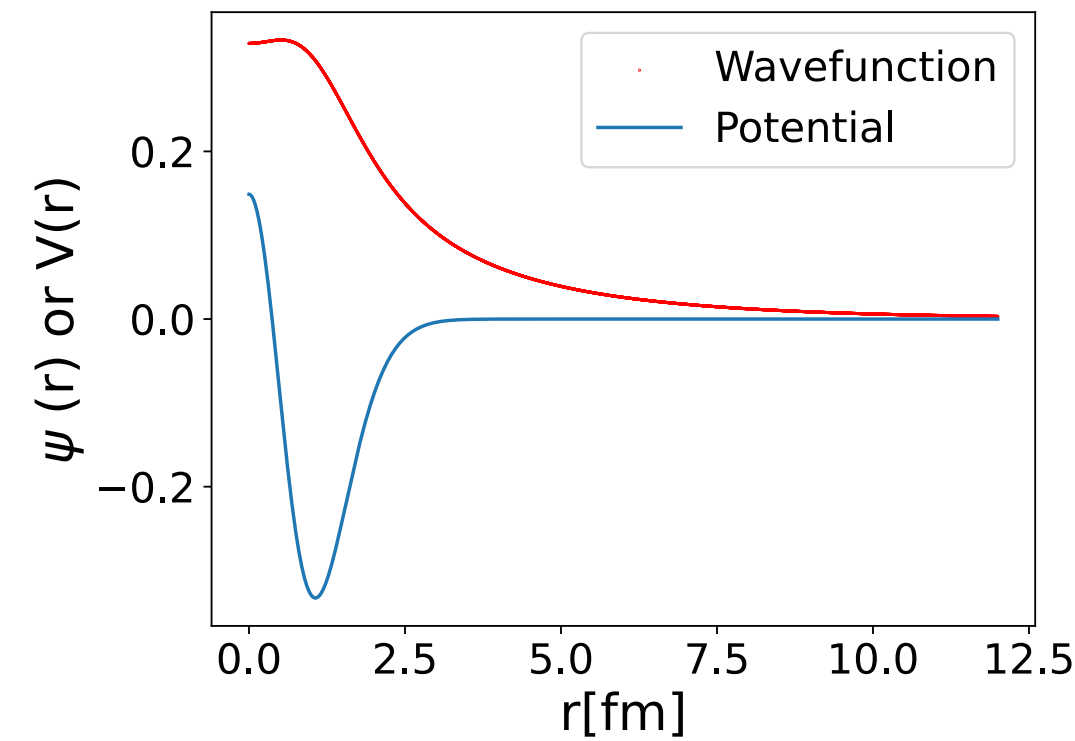
- ◆ Through boundary conditions, one may target excited (and scattering) states.

L. Brevi, A. Mandarino, E. Prati,  
Technologies 12, 174 (2024); New J. Phys. 26, 103015 (2024).

# Physics Informed Neural Networks for the deuteron

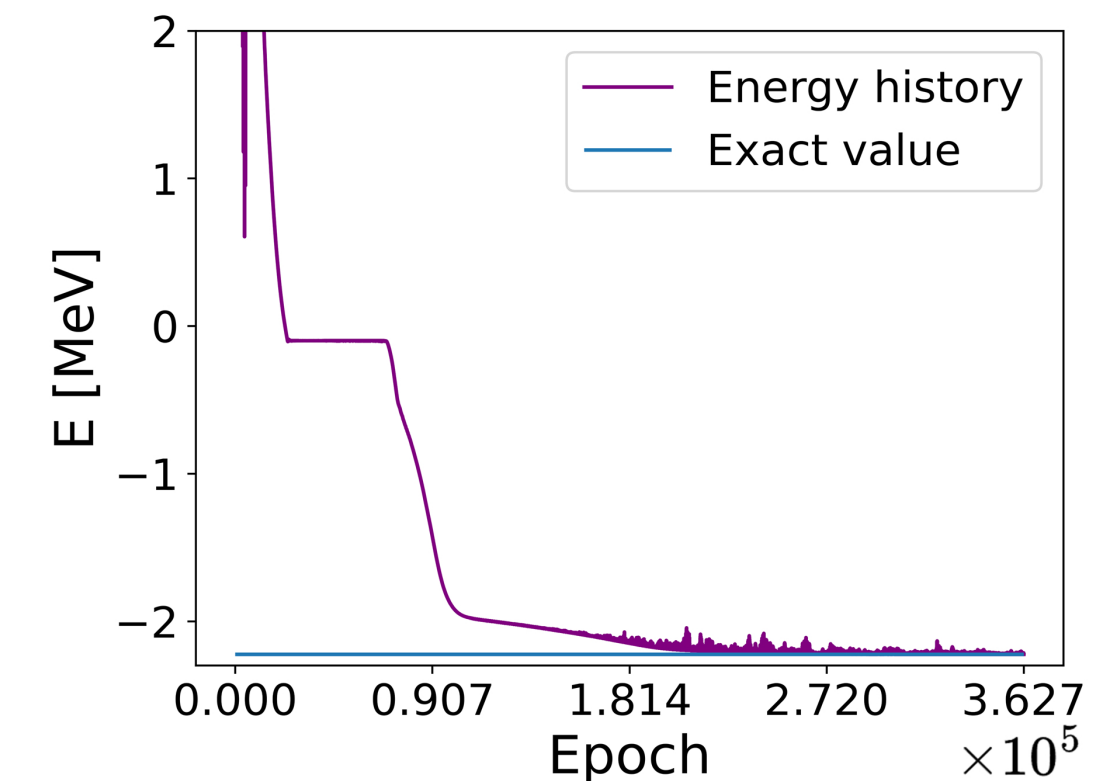
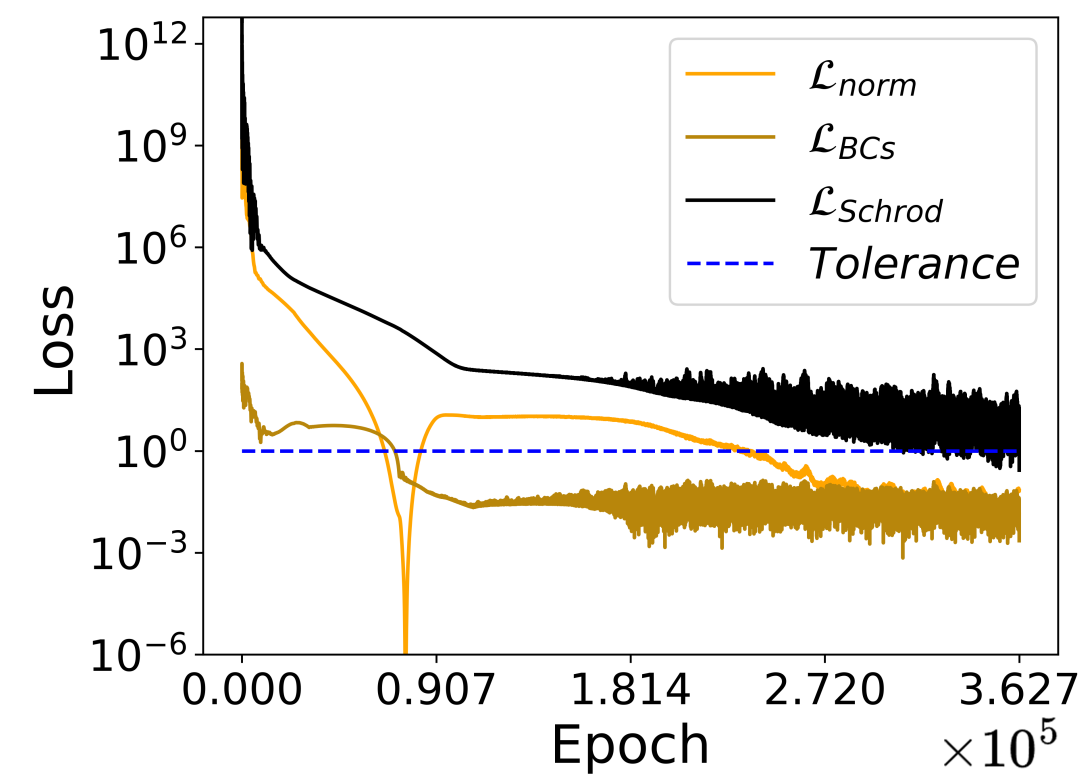
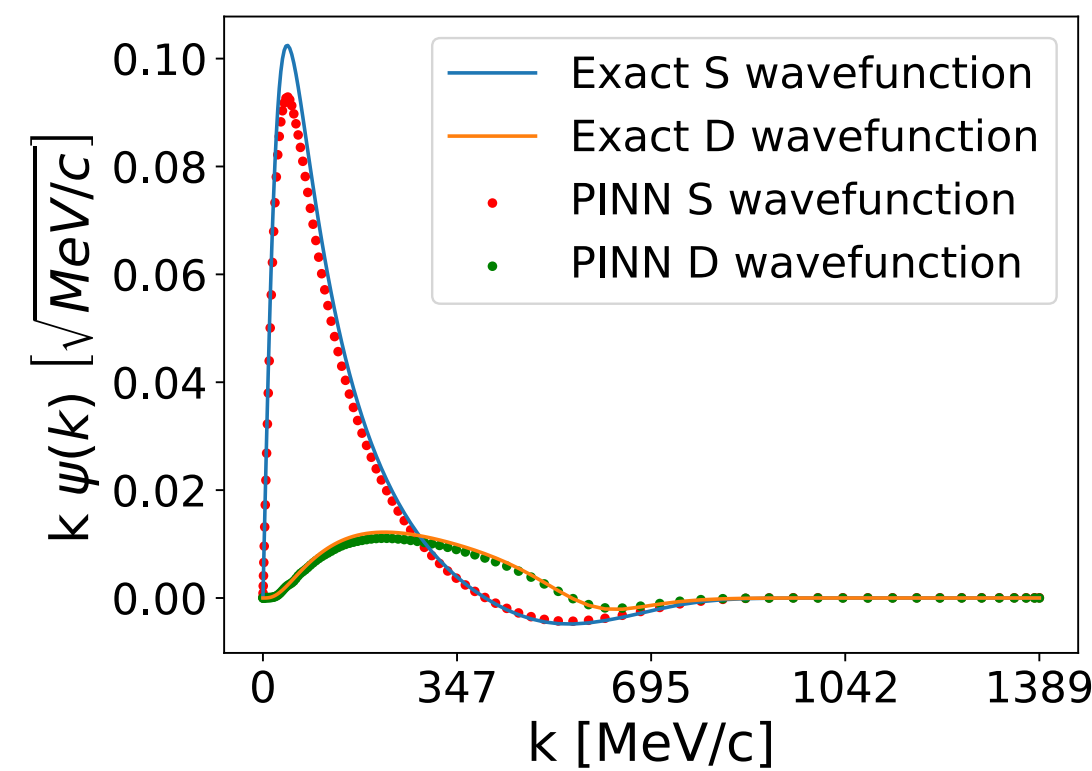
L. Brevi, A. Mandarino, CB, E. Prati, arXiv:2602.11193

## Coordinate space, Minnesota potential:



	$Err_{E,exp}$	$Err_{E,Num}$	$1 - \mathcal{F}_\psi$
Minnesota	0.01	$1 \times 10^{-8}$	
N4LO(550)	$-5.96 \times 10^{-6}$	$-5.21 \times 10^{-5}$	$1 \times 10^{-5}$
CD-Bonn	$-6.01 \times 10^{-4}$	$-2.76 \times 10^{-7}$	$1 \times 10^{-9}$

## Momentum space with N4LO(550) from EMN:



$$\mathcal{L}_{phys} = w_{ODE} \mathcal{L}_{ODE} + (w_{int} \mathcal{L}_{int} + w_{norm} \mathcal{L}_{norm}) + w_{BCs} \mathcal{L}_{BCs} + w_{var} \mathcal{L}_{var}$$

# Toward Lattice Simulations with Neural Networks

Work in progress...

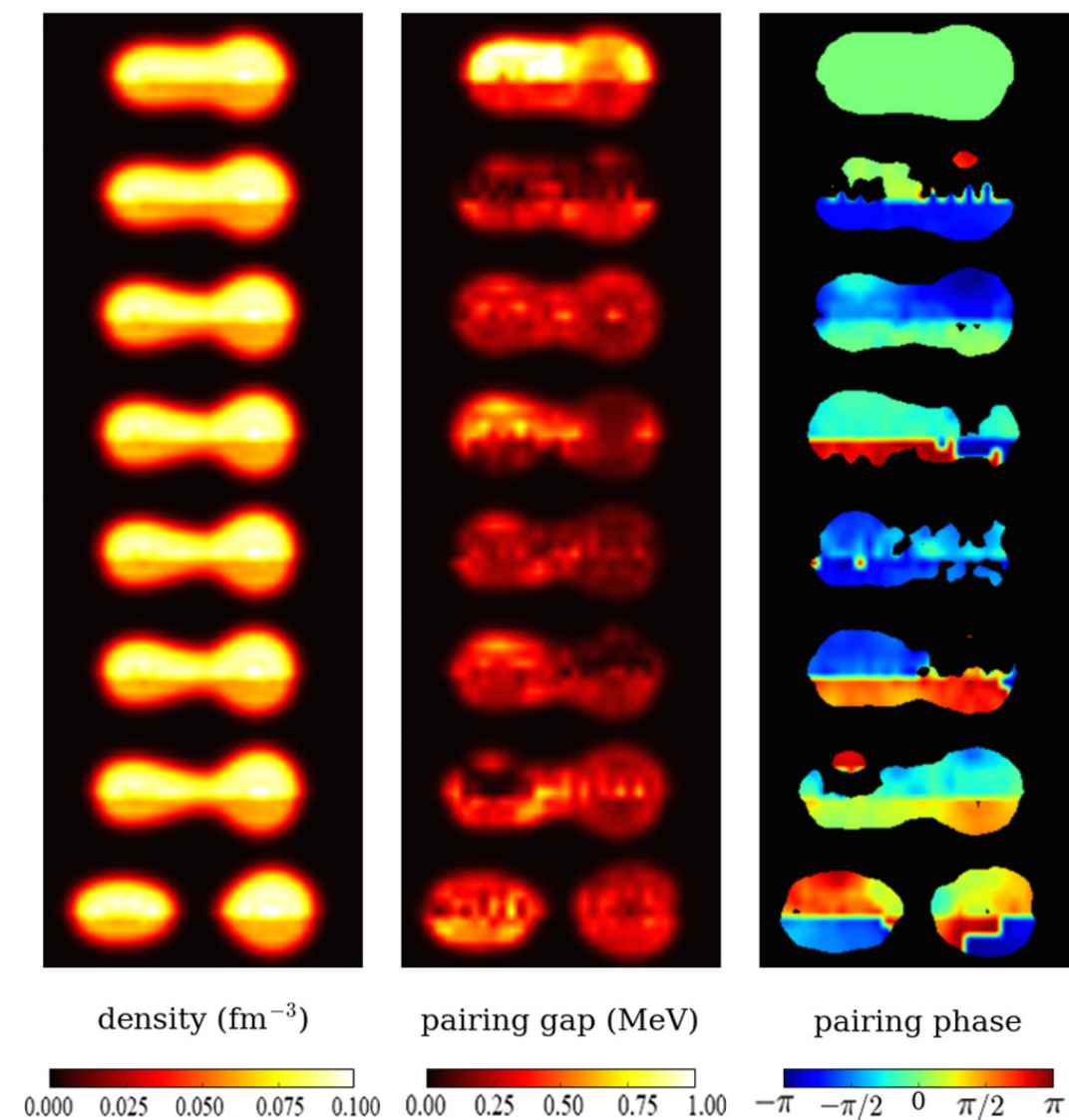
L. Lazzarino, G. Borroni, E. Redaelli - BSc theses U. Of Milan



# Why NQS on the Lattice?

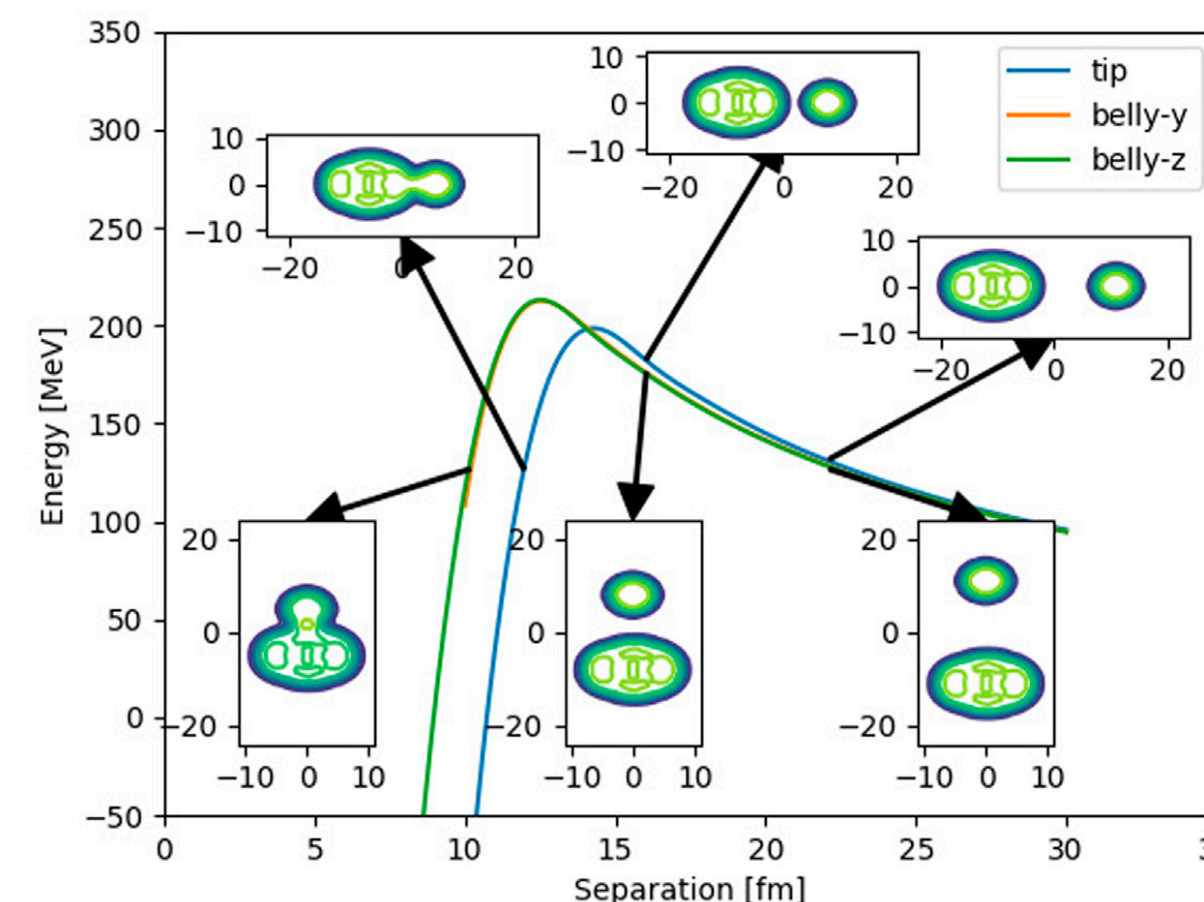
## Advantages of a lattice-NQS for nuclei:

- Fermi-Pauli statistics comes for free (Fock space).
- Not tied to spherical or partially deformed ansätze (full deformation, etc...).
- Transfer learning (train few-nucleon first).
- Many-body dynamics.



Fission of  $^{240}\text{Pu}$ :

- time dependent DFT inspired, in 3D
- $30 \times 30 \times 60 \text{ fm}^3$  box
- $24 \times 24 \times 48 = 27,000$  pts mesh

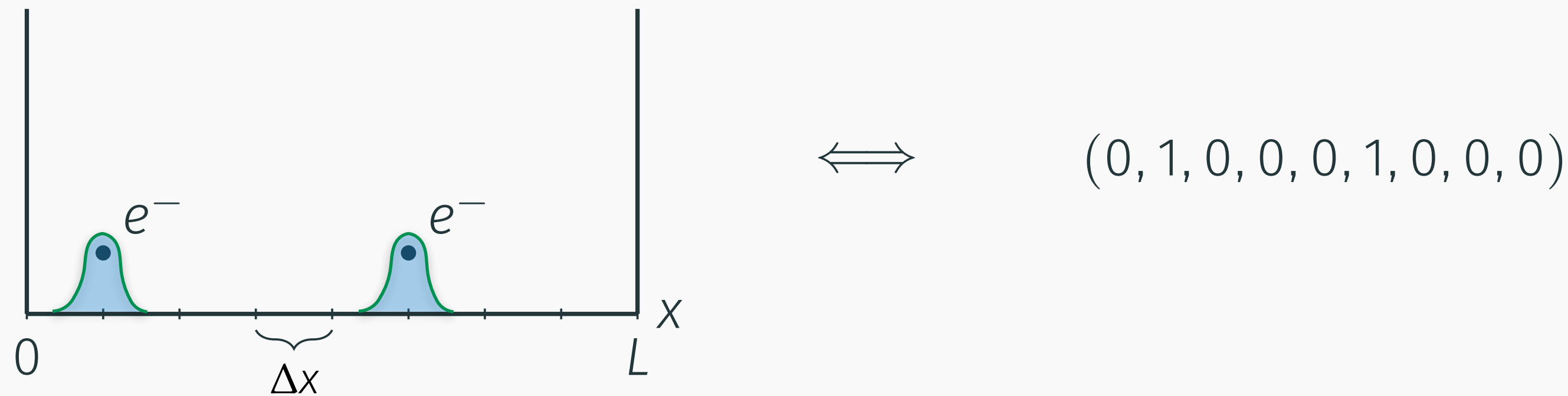


Mean-field simulations of Es-254 + Ca-48 heavy-ion reactions

P. Stevenson et al.,  
Frontiers **10**, 1019285 (2022)

# Confined fermions w/ a discrete coordinate space mesh

- Discretise coordinate space
- Use occupation number to locate particles



*no need to worry about antisymmetrization!*

- Use a Fock space basis to represent particle configurations:

$$|\psi\rangle = \prod_i \psi^\dagger(x_i) |0\rangle = |n_0=0, n_1=1, n_2=0, n_3=0, n_4=0, n_5=1, \dots, n_L=0\rangle$$

- Can be mapped into a system of spins (*with fixed magnetisation*):

$$\langle x|\psi\rangle \rightarrow \langle S|\psi\rangle \Leftrightarrow \begin{cases} c_{\uparrow\uparrow\uparrow\dots} \doteq \langle \uparrow\uparrow\uparrow \dots | \psi \rangle = \psi(\uparrow\uparrow\uparrow \dots) \\ c_{\downarrow\uparrow\uparrow\dots} \doteq \langle \downarrow\uparrow\uparrow \dots | \psi \rangle = \psi(\downarrow\uparrow\uparrow \dots) \\ \vdots \\ c_{\downarrow\downarrow\downarrow\dots} \doteq \langle \downarrow\downarrow\downarrow \dots | \psi \rangle = \psi(\downarrow\downarrow\downarrow \dots) \end{cases}$$

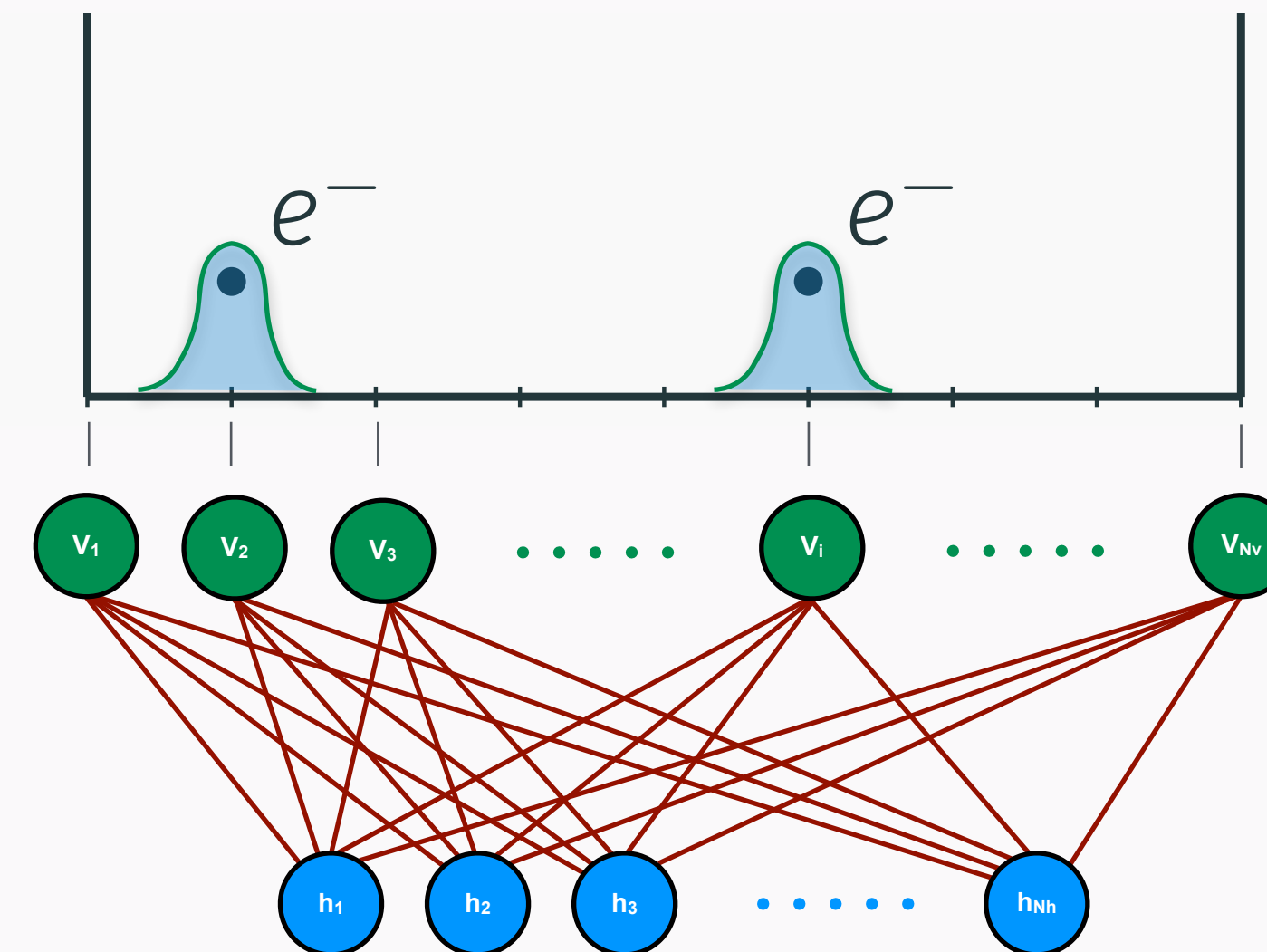
# NQS representation

- Use a Restricted Boltzmann Machine with complex parameter to represent the w.f.:

$$\mathcal{P}(\mathbf{v} \cap \mathbf{h}) = \frac{1}{\mathcal{Z}} \exp(\mathbf{a}^T \mathbf{v} + \mathbf{b}^T \mathbf{h} + \mathbf{h}^T \underline{\underline{W}} \mathbf{v})$$

with:

$$\begin{cases} \mathbf{v} \in \{-1, 1\}^{N_v} \\ \mathbf{h} \in \{-1, 1\}^{N_h} \\ W \in \text{Mat}_{N_h \times N_v}(\mathbb{C}) \end{cases} \begin{cases} \mathbf{a} \in \mathbb{C}^{N_v} \\ \mathbf{b} \in \mathbb{C}^{N_h} \\ W \in \text{Mat}_{N_h \times N_v}(\mathbb{C}) \end{cases}$$



NOTE: The RBM architecture will be independent from the particle number.

- Marginalize w.r.t. the hidden nodes:

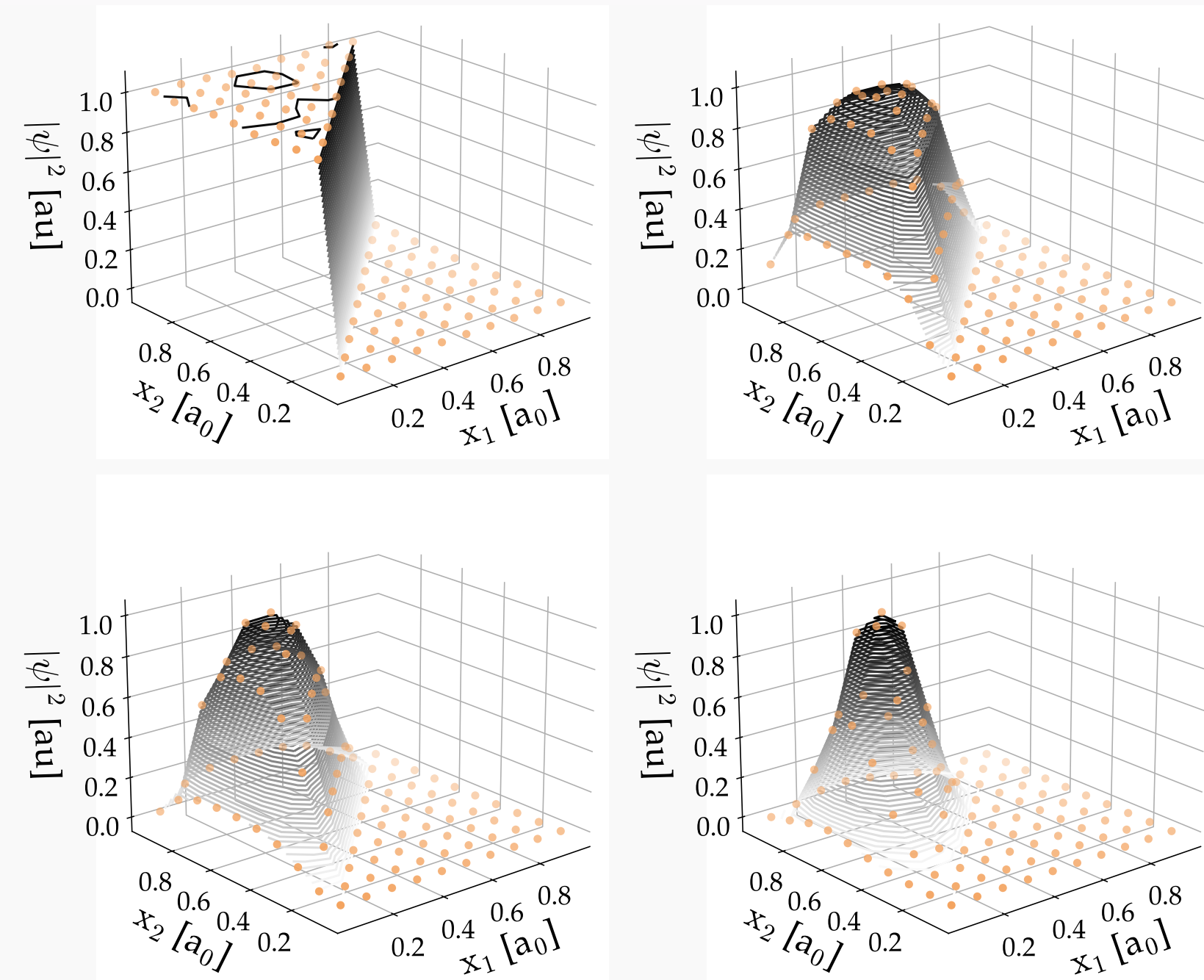
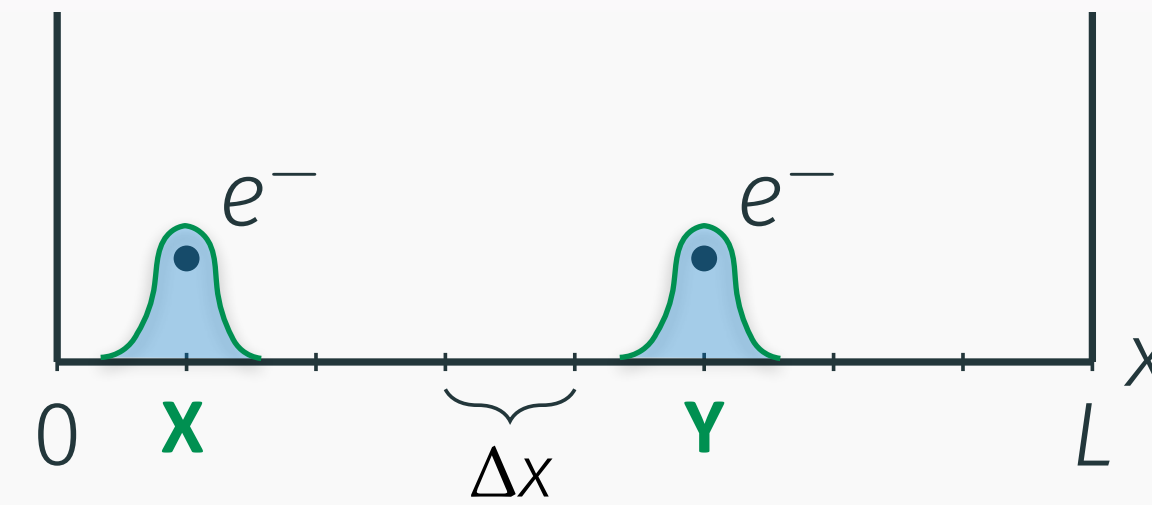
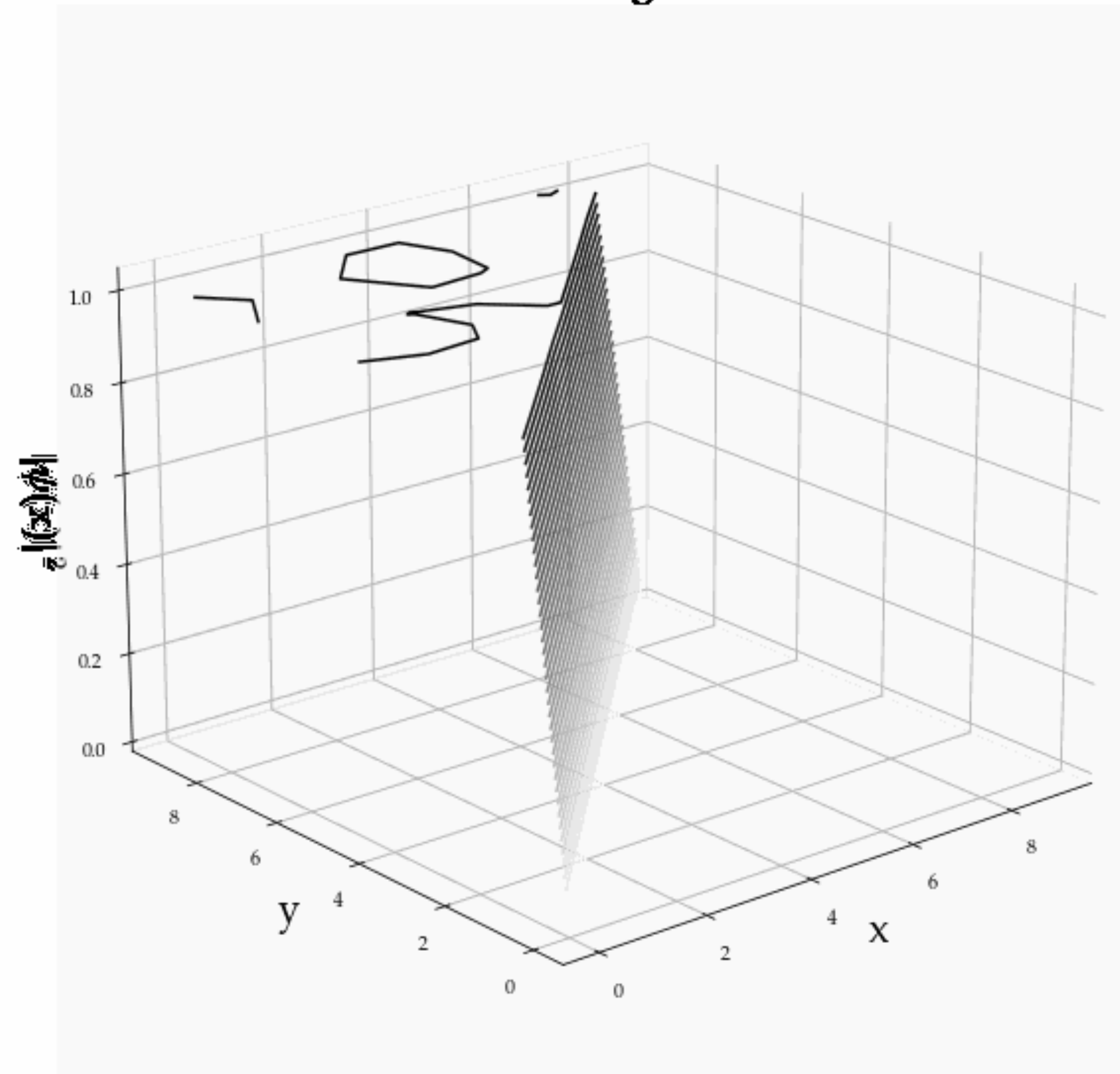
$$\langle x | \psi \rangle \rightarrow \mathcal{P}(\mathbf{v}) = \sum_{\{\mathbf{h}\}} \mathcal{P}(\mathbf{v} \cap \mathbf{h}) \quad \longrightarrow$$

## Restricted Boltzmann Machine

$$\psi(\mathbf{v}) = 2^{N_h} \exp(\mathbf{a}^{(0)T} \mathbf{v}) \prod_{i=1}^{N_h} \left[ \exp(K^{(i)} + \mathbf{a}^{(i)T} \mathbf{v}) \cosh(\omega_i(\mathbf{v})) \right]$$

# Two fermions — NQS wave function

Neural network ground state

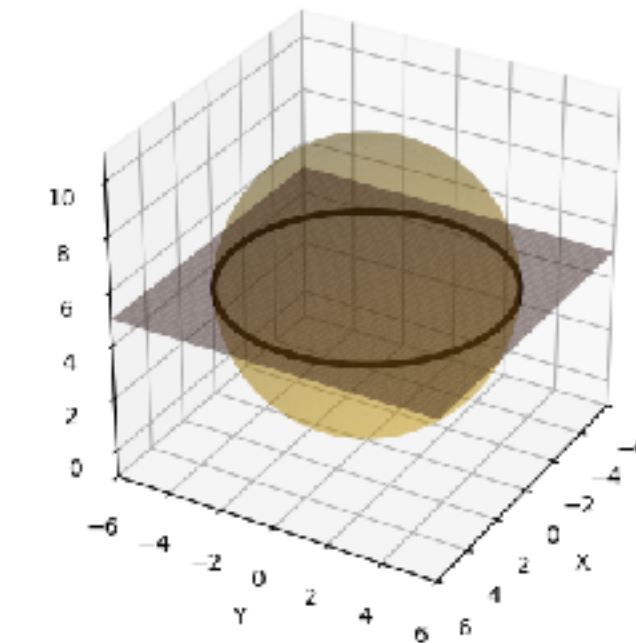
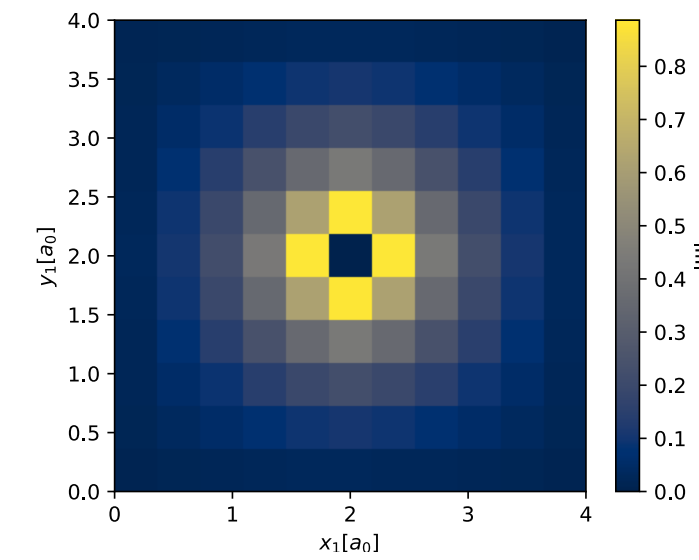
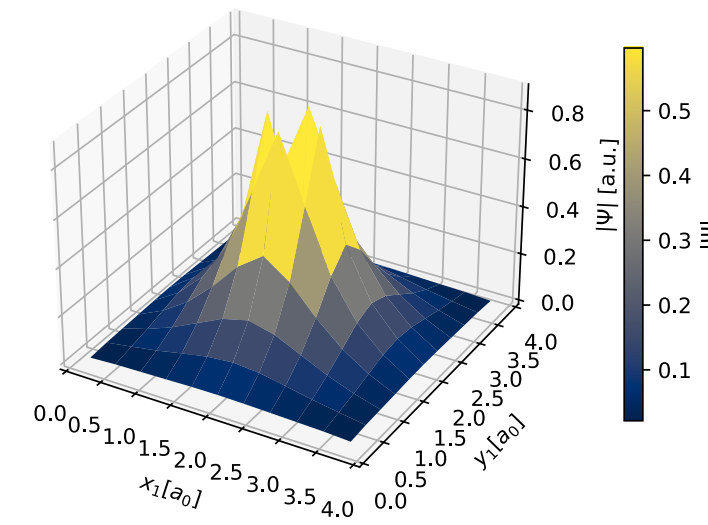


**Figure 2:** Physical learning process of an RBM with  $N_h = 20$  from the starting wave function (upper left). The NN seems to be learning boundaries and antisymmetry in the following iterations (number 15, 30 and 300 are reported).

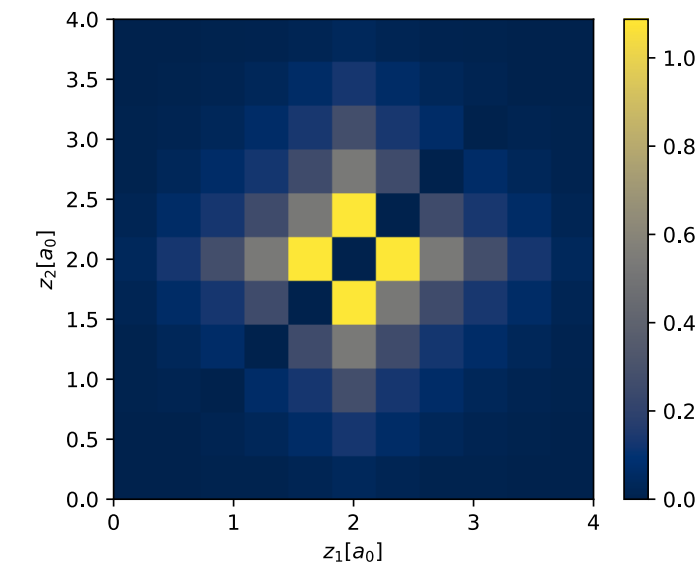
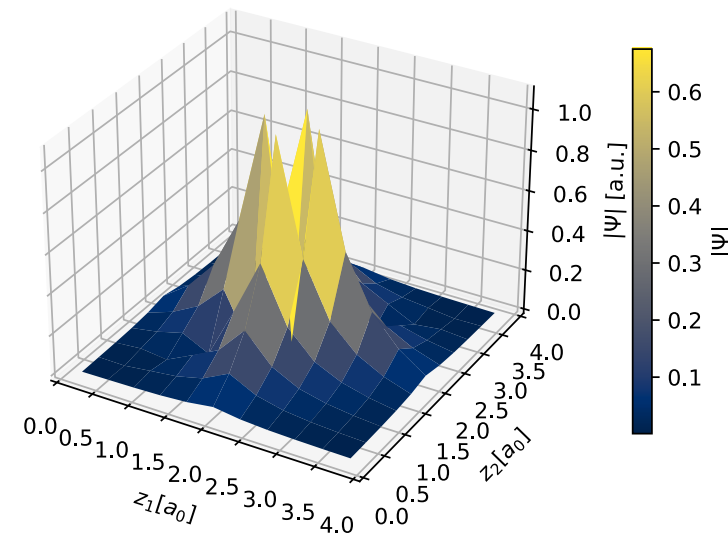
# Examples for 2- and 3-fermion systems

Lithium atom  
(3 electrons)

Motion of one electron on the equatorial plan, with the other two (spin up and down) near the atom's center.



Two spin-up electrons on a vertical line passing through the nucleus; and with the third (spin-down) near the atom's center.



The deuteron  
(proton-neutron)

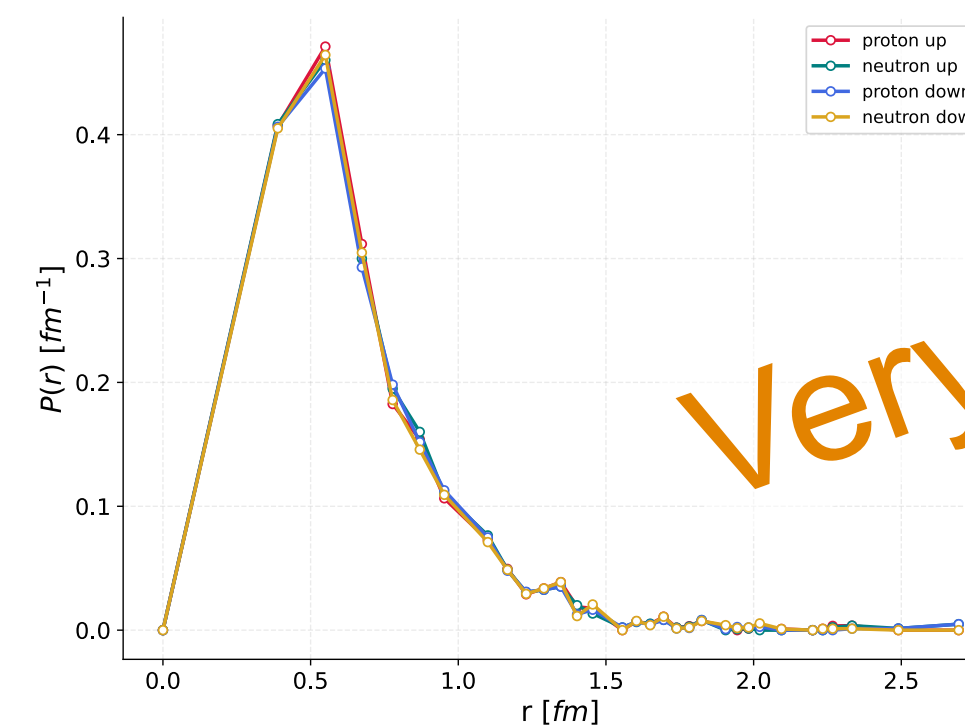
Extension to a **4-fold RBM lattice** to account for spin and isospin.

Semi-realistic Argonne V6 potential

$$V = \sum_{i=1}^6 V_i(r) \hat{O}_i$$

$$O_1 = \mathbb{I}, \quad O_2 = \boldsymbol{\sigma}_1 \cdot \boldsymbol{\sigma}_2, \quad O_3 = \boldsymbol{\tau}_1 \cdot \boldsymbol{\tau}_2, \quad O_4 = (\boldsymbol{\sigma}_1 \cdot \boldsymbol{\sigma}_2)(\boldsymbol{\tau}_1 \cdot \boldsymbol{\tau}_2),$$

$$O_5 = S_{12}, \quad O_6 = S_{12}(\boldsymbol{\tau}_1 \cdot \boldsymbol{\tau}_2)$$



Very preliminary!

# Recent Progress in Many-Body Theories (RPMBT-23)

23° International Conference on  
Recent Progress in Many-Body Theories

## RPMBT23

September 14-18, 2026 - Milan, Italy



<https://rpmbt23.mi.infn.it/>

### Conference topics

Condensed Matter Physics

Computational Quantum  
Many-Body Physics

Strongly Correlated  
Electronic Systems

Nuclear and  
Subnuclear Physics

Atomic and  
Molecular Physics

Quantum Fluids,  
Ultracold Gases

Quantum Field Theory,  
Astrophysics

Quantum Information and  
Computation

Quantum Chemistry,  
Quantum Magnetism

<https://rpmbt23.mi.infn.it/>

RPMBT-23 conference in Milan, Sept 14-18, 2026

School on Quantum Simulation, Sept. 9-11, 2026

Abstract submission opens in a few days!

*(...and Thank you for your*

*Attention!!!!)*

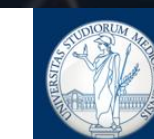
Satellite event:

### School on Quantum Simulation

Physics Department  
Università degli Studi di Milano  
September 9 - 11, 2026, Milan, Italy



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