NEURAL NETWORK QUANTUM STATES FOR FEW-BODY NUCLEAR SYSTEMS



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Summer school "Light-ion physics in the EIC era"

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ORGANIZATION OF THIS BRIEF COURSE

Scope of the course

- Introduce neural-network quantum states in the continuum (coordinate space)
- Applications to the nuclear quantum few-body problem

Coding goal:

• Solve the deuteron in pionless effective field theory using neural-network quantum states

Structure of the lectures:

• There will be a single 60-minute lecture plus a 60-minute coding session

Coding requirements

• We will use Jupiter notebooks; required packages are Numpy and Google-JAX (use Google Colab is highly recommended)



NUCLEAR PHYSICS

A TALE OF SCALES

• Long-range: r ~ 5 fm

Collective nuclear deformation evidenced by the characteristic rotational spectra



T. Naito et al 2021 J. Phys. B 54 165201 (2021)



A TALE OF SCALES

- Intermediate-range: r ~ 2 fm
- Formation of alpha clusters





S. Shen, et al., Nat. Comm. 14, 2777 (2023)

A TALE OF SCALES

• Short-range: r ~ 1 fm

Overlapping nucleons

Incident electron







Correlated partner proton or neutron

BROADER IMPACT



Credit: N. Rocco



"AB-INITIO" NUCLEAR THEORY



Illustration by APS / Alan Stonebraker

NUCLEAR HAMILTONIAN

Realistic nuclear Hamiltonians include two- and three-body potentials

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





NUCLEAR HAMILTONIAN

Nuclear potentials depend on the spatial position, spin, and isospin of the nucleons

Need to introduce a generalized coordinate $x \equiv {\mathbf{r}_i, s_z, t_z}$

$$s_z = +1/2 \quad \bigwedge \qquad t_z = +1/2 \quad \bigcirc \qquad t_z = -1/2 \quad \sub \qquad t_z = -1/2 \quad \underbrace \qquad t_z$$

NUCLEAR HAMILTONIAN



K. Hebeler et al., Phys. Rept. 890 (2021) 1-116

THE QUANTUM MANY-BODY PROBLEM

• Non relativistic many body theory aims at solving the many-body Schrödinger equation

$$H\Psi_n(x_1,\ldots,x_A) = E_n\Psi_n(x_1,\ldots,x_A)$$

• Nucleons are fermions, so the wave function must be anti-symmetric

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

THE QUANTUM MANY-BODY PROBLEM



A guide to Feynman diagrams in the many-body problem

THE MEAN-FIELD APPROXIMATION

Mean field: nucleons are independent particles subject to an average nuclear potential



THE MEAN-FIELD APPROXIMATION

The mean-field ground-state wave function is a Slater determinant



CONFIGURATION-INTERACTION METHODS



Image courtesy of Patrick Fasano

TACKLE LARGE SYSTEMS

Polynomially-scaling methods reach (much) larger systems with some approximations



B. S. Hu et al., Nature Phys. (2022) $z = M(\theta) + \varepsilon_{exp} + \varepsilon_{em} + \varepsilon_{model} + \varepsilon_{model},$ (1)



NEED FOR HIGH RESOLUTION

Continuum nuclear quantum Monte Carlo use a coordinate-space representation of many-body wave functions.

- They have no difficulties in treating "stiff" nuclear forces: test the convergence of nuclear EFTs;
- Access to high-momentum components of the nuclear wave functions;
- Limited to relatively light nuclear systems



R. Cruz-Torres et al., Nature Phys. 17 (2021) 3, 306

INCLUDING CORRELATIONS



INCLUDING CORRELATIONS



VARIATIONAL MONTE CARLO





VARIATIONAL MONTE CARLO

Homogeneous liquid of ⁴He atoms; interaction parametrized by the Lennard Jones potential

The wave function must be small (large) where the potential is repulsive (attractive)

$$v(r) = 4\epsilon \left[\left(\frac{\sigma}{r}\right)^{12} - \left(\frac{\sigma}{r}\right)^6 \right]$$

$$\Psi_V(R) = \prod_{i < j} f(r_{ij}) ; \quad f(r) = \exp\left[-\frac{1}{2}\left(\frac{b}{r}\right)^5\right]$$



VARIATIONAL MONTE CARLO

The **variational principle** guarantees the variational energy to be larger greater than or equal to the ground-state energy with the same quantum numbers

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

Computing the variational energy requires evaluating a high-dimensional integral

$$E_V = \frac{\langle \Psi_V | H | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle} = \frac{\sum_S \int dR \langle \Psi_V | RS \rangle \langle RS | H | \Psi_V \rangle}{\sum_S \int dR \langle \Psi_V | RS \rangle \langle RS | \Psi_V \rangle} = \frac{\sum_S \int dR | \Psi_V (R, S) |^2 \frac{\langle RS | H | \Psi_V \rangle}{\langle RS | \Psi_V \rangle}}{\sum_S \int dR | \Psi_V (R, S) |^2}$$

$$R = \{\mathbf{r}_1, \dots, \mathbf{r}_A\} \quad \longleftrightarrow \quad S = \{(s_1^z, t_1^z), \dots, (s_A^z, t_A^z)\}$$

ESTIMATING OBSERVABLES

$$E_V = \frac{\sum_S \int dR |\Psi_V(R,S)|^2 \frac{\langle RS|H|\Psi_V \rangle}{\langle RS|\Psi_V \rangle}}{\sum_S \int dR |\Psi_V(R,S)|^2} \,.$$

Leverage the central limit theorem, we can estimate the above integral as

$$E_V \simeq \frac{1}{N} \sum_{(R_i, S_i) \sim \pi_V} \frac{\langle R_i S_i | H | \Psi_V \rangle}{\langle R_i S_i | \Psi_V \rangle} \quad \longleftrightarrow \quad \pi_V(R, S) = \frac{|\Psi_V(R, S)|^2}{\sum_S \int dR |\Psi_V(R, S)|^2}$$

We use Metropolis Hastings to sample $\pi_V(R, S)$

$$P_{\rm acc} = \min\left(1, \frac{|\Psi_V(R', S')|^2}{|\Psi_V(R, S)|^2}\right)$$



Goal: sample the probability distribution

$$R_n \sim P(R) = \frac{|\Psi_V(R)|^2}{\int dR |\Psi_V(R)|^2}$$

The M(RT)² algorithm is **based on the idea of random walk**. A set of random configurations are generated by applying the transformation.

$$P_{i+1}(R_{i+1}) = \int dR_i P_i(R_i) T(R_i \to R_{i+1})$$

By recursively applying the same transformation we get

$$P_n(R_n) = \int dR_1 \dots dR_{n-1} P_1(R_1) T(R_1 \to R_2) \dots T(R_{n-1} \to R_n)$$

Under some very general conditions it can be proven that

$$\lim_{n \to \infty} P_n(R_n) = P(R) \longrightarrow P(R) \text{ only depends on } T(R_i \to R_{i+1})$$

Let us impose the asymptotic distribution to be in an "equilibrium" state, which translates into **detailed balance condition:** that point by point there is no net flux of probability

 $P(R)T(R \to R') = P(R')T(R' \to R)$

We can arbitrarily split the transition probability in two terms

$$T(R \to R') = G(R \to R')A(R \to R')$$
Proposal probability

The detailed balance implies:

 $\frac{A(R' \to R)}{A(R \to R')} = \frac{P(R)G(R \to R')}{P(R')G(R' \to R)}$

It can be easily checked that the following acceptance probability satisfies the above requirement

$$A(R \to R') = \min\left(1, \frac{P(R')G(R' \to R)}{P(R)G(R \to R')}\right)$$

A common choice for $G(R \rightarrow R')$ is a Gaussian distribution centered in zero. In this case, at each step of the propagations, the walkers are moved by $R' = R + \Delta$



- In the multiple-particle case, we need to consider a three-dimensional Gaussian for each particle.
- Since the Gaussian probability is symmetric, the acceptance probability simplifies to

$$A(R \to R') = \min\left(1, \frac{P(R')}{P(R)}\right)$$

M(RT)² APPLIED TO VMC

At this point, we can describe the Metropolis algorithm for a VMC calculation

Step 0 - Start from an arbitrary distribution of configurations on the coordinate R

Step 1 - Move the walkers according to $G(R \rightarrow R')$, such as $R' = R + \Delta$

Step 2 - Compute the acceptance probability

$$A(R \to R') = \min\left(1, \frac{|\Psi_V(R')|^2}{|\Psi_V(R)|}\right)$$

Step 3 - Accept or reject the proposed move comparing with $\xi \sim U(0,1]$

$$\frac{|\Psi_V(R')|^2}{|\Psi_V(R)|^2} \ge \xi \quad \longrightarrow \quad \text{Accept: } R = R' \quad \text{;} \quad \frac{|\Psi_V(R')|^2}{|\Psi_V(R)|^2} < \xi \quad \longrightarrow \quad \text{Reject} \quad R = R$$

M(RT)² APPLIED TO VMC



- Sample from an initial distribution
- Random Gaussian move
- Acceptance/rejection of the move
- Random Gaussian move
- Acceptance/rejection of the move

• Iterate until convergence

ESTIMATING OBSERVABLES

$$E_V = \frac{\sum_S \int dR |\Psi_V(R,S)|^2 \frac{\langle RS|H|\Psi_V \rangle}{\langle RS|\Psi_V \rangle}}{\sum_S \int dR |\Psi_V(R,S)|^2} \,.$$

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We use Metropolis Hastings to sample $\pi_V(R, S)$

$$P_{\rm acc} = \min\left(1, \frac{|\Psi_V(R', S')|^2}{|\Psi_V(R, S)|^2}\right)$$

ENERGY MINIMIZATION

The variational wave function depends on a set of variational parameters

 $|\Psi_V\rangle$ \longrightarrow $|\Psi_V(\mathbf{p})\rangle$

It is convenient to introduce the derivative operator

$$O^{i}|\Psi_{V}(\mathbf{p})\rangle \equiv \frac{\partial}{\partial p_{i}}|\Psi_{V}(\mathbf{p})\rangle$$

Recall:

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

The gradient of the energy reads

$$g_i \equiv \frac{\partial E_V}{\partial p_i} = 2\left(\frac{\langle \Psi_V | O^i H | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle} - \frac{\langle \Psi_V | H | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle} \frac{\langle \Psi_V | O^i | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle}\right) = \langle HO^i \rangle - \langle H \rangle \langle O^i \rangle \longleftrightarrow O^i = \frac{\partial P_i}{\partial p_i}$$

FIRST-ORDER UPDATES

Since the gradient is known, we can use stochastic gradient descent to update the parameters

$$\mathbf{p}^{n+1} = \mathbf{p}^n - \eta \, \mathbf{g}^n \quad \longrightarrow \quad \eta \simeq 0.001$$

Inspired by machine-learning application, we can consider RMS, Adam, or any other improved firstorder algorithm

First-order algorithms are commonly used within the NQS community, but typically require many iterations to converge

NATURAL GRADIENT

A more powerful method consists in performing an imaginary-time evolution in the variational manifold

$$\begin{split} |\bar{\Psi}_{V}(\mathbf{p}_{\tau}) &\equiv (1 - H\delta\tau) |\Psi_{V}(\mathbf{p}_{\tau})\rangle \\ \mathbf{p}_{\tau+\delta\tau} &= \operatorname*{arg\,max}_{\mathbf{p}\in R^{d}} \left(\left| \langle \bar{\Psi}_{V}(\mathbf{p}_{\tau}) | \Psi_{V}(\mathbf{p}_{\tau+\delta\tau}) \rangle \right|^{2} \right) \end{split}$$



The parameters are updated as

$$\mathbf{p}_{\tau+\delta\tau} = \mathbf{p}_{\tau} - \delta\tau S^{-1} \mathbf{g}_{\tau} \longrightarrow S_{ij} = \frac{\langle \Psi_V | O^i O^j | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle} - \frac{\langle \Psi_V | O^i | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle} \frac{\langle \Psi_V | O^j | \Psi_V \rangle}{\langle \Psi_V | \Psi_V \rangle}$$

J. Stokes, at al., Quantum 4, 269 (2020).

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

NEURAL-NETWORK QUANTUM STATES



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

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

To begin with, let us consider a prototypal problem: the 1d quantum harmonic oscillator

$$H = -\frac{\partial^2}{\partial x^2} + \frac{x^2}{2} \longrightarrow \Psi_0(x) = e^{-x^2/2} \quad ; \quad E_0 = 1/2$$

We represent the ground-state wave function with a feed-forward neural network



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FERMION MANY-BODY WAVE FUNCTIONS

The fermion antisymmetry must be built in the wave function

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

Different ansatzë have been put forward

- Slater-Jastrow: $\Psi_V(X) = e^{U(X)} \Phi_S(X) \longrightarrow \Psi_V(X) = e^{U(X)} \Phi_S(Y_{bf}(X))$
- Hidden Fermions: $\Psi_V(X) = \Phi_{HF}(X)$
- Neural Pfaffian: $\Psi_V(X) = e^{U(X)} \Phi_P(X) \longrightarrow \Psi_V(X) = e^{U(X)} \Phi_P(Y_{bf}(X))$

Hermann et al., Nature Chemistry, **12**, 891 (2020) Pfau et al., PRR **2**, 033429 (2020) J. Stokes et al., PRB, **102**, 205122 (2020) J. R. Moreno, et al., PRL **125**, 076402 (2022)

$$\Psi_{V}(X) = e^{J(X)} \times \det \begin{bmatrix} \phi_{1}(x_{1}) & \phi_{1}(x_{2}) & \cdots & \phi_{1}(x_{N}) \\ \phi_{2}(x_{1}) & \phi_{2}(x_{2}) & \cdots & \phi_{2}(x_{N}) \\ \vdots & \vdots & \ddots & \vdots \\ \phi_{N}(x_{1}) & \phi_{N}(x_{2}) & \cdots & \phi_{N}(x_{N}) \end{bmatrix}$$

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J. Stokes et al., Physical Review B **102**, 205122 (2020) *Pfau et al., Physical Review Research* **2**, 033429 (2020) *Hermann et al., Nature Chemistry* **12**, 891 (2020)

Mean-field component: Slater determinant of single-particle orbitals

$$\Phi(X) = \begin{bmatrix} \phi_1(x_1) & \phi_1(x_2) & \cdots & \phi_1(x_N) \\ \phi_2(x_1) & \phi_2(x_2) & \cdots & \phi_2(x_N) \\ \vdots & \vdots & \ddots & \vdots \\ \phi_N(x_1) & \phi_N(x_2) & \cdots & \phi_N(x_N) \end{bmatrix}$$

(If needed) The center of mass motion is automatically removed by

$$\bar{\mathbf{r}}_i = \mathbf{r}_i - \mathbf{R}_{CM}$$



"Manually" imposing permutation-invariance scales factorially with A

 $J(X) = j(x_1, x_2, x_3) + j(x_1, x_3, x_2) + j(x_2, x_1, x_3) + j(x_2, x_3, x_1) + j(x_3, x_1, x_2) + j(x_3, x_2, x_1)$

"Manually" imposing permutation-invariance scales factorially with A

 $J(X) = j(x_1, x_2, x_3) + j(x_1, x_3, x_2) + j(x_2, x_1, x_3) + j(x_2, x_3, x_1) + j(x_3, x_1, x_2) + j(x_3, x_2, x_1)$

Solution: "deep-sets"
$$\longrightarrow J(X) = \rho_F \left[\sum_i \vec{\phi}_F(\bar{\mathbf{r}}_i, \mathbf{s}_i) \right]$$



Wagstaff et al., arXiv:1901.09006 (2019)

NEURAL BACKFLOW CORRELATIONS

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



Di Luo and B. K. Clark, Phys. Rev. Lett. 122, 226401 (2019)

BACK TO NUCLEAR PHYSICS



NUCLEAR HAMILTONIAN

• Nucleon-nucleon potential fit to s-wave np scattering lengths and effective ranges



R. Schiavilla, AL, PRC 103, 054003 (2021)

NUCLEAR HAMILTONIAN

Let us examine the different spin-isospin operators

$$\tau_{ij} = \boldsymbol{\tau}_i \cdot \boldsymbol{\tau}_j = 2P_{ij}^{\tau} - 1 \quad \longrightarrow \quad \tau_{ij} |t_i^z t_j^z\rangle = 2|t_j^z t_i^z\rangle - |t_i^z t_j^z\rangle$$
$$\sigma_{ij} = \boldsymbol{\sigma}_i \cdot \boldsymbol{\sigma}_j = 2P_{ij}^{\sigma} - 1 \quad \longrightarrow \quad \sigma_{ij} |s_i^z s_j^z\rangle = 2|s_j^z s_i^z\rangle - |s_i^z s_j^z\rangle$$

The spin-isospin operator is just the combination of the above two

$$\sigma_{ij}\tau_{ij} = 4P^{\sigma}_{ij}P^{\tau}_{ij} - 2P^{\sigma}_{ij} - 2P^{\tau}_{ij} + 1$$

$$\sigma_{ij}\tau_{ij}|s_i^z t_i^z s_j^z t_j^z\rangle = 4|s_j^z t_j^z s_i^z t_i^z\rangle - 2|s_j^z t_i^z s_i^z t_j^z\rangle - 2|s_i^z t_j^z s_j^z t_i^z\rangle + |s_i^z t_i^z s_j^z t_j^z\rangle$$

LET'S SOLVE THE DEUTERON

The deuteron is the lightest nucleus, composed of a proton and a neutron



The total spin is S = 1, the total isospin is T = 0

$$\begin{split} |\Phi\rangle &= |\uparrow\uparrow\rangle \otimes \frac{1}{\sqrt{2}} \left(|pn\rangle - |np\rangle \right) \\ \Phi(S) &\equiv \langle s_1^z t_1^z s_2^z t_2^z |\Phi\rangle = \langle s_1^z s_2^z |\uparrow\uparrow\rangle \frac{1}{\sqrt{2}} \left(\langle t_1^z t_2^z |pn\rangle - \langle t_1^z t_2^z |np\rangle \right) \end{split}$$

LET'S SOLVE THE DEUTERON

Taking into account the spatial dependencies, the most general wave function can be parameterized as

 $\Psi(R,S) = e^{U(r_{12})}\Phi(S)$

• Since $U(r_{12})$ only depends on the distance $r_{12} = |\mathbf{r}_{12}|$, the wave function is symmetric under the exchange of spatial coordinates

• However, $\Phi(S)$ is antisymmetric under the exchange of spin-isospin coordinates

• Hence, the total wave function $\Psi(R, S)$ is antisymmetric under the exchange of two particles

LOCAL ENERGY

Reminder: the energy can be estimated as

$$E_V \simeq \frac{1}{N} \sum_{(R_i, S_i) \sim \pi_V} \frac{\langle R_i S_i | H | \Psi_V \rangle}{\langle R_i S_i | \Psi_V \rangle}$$

We need to compute the local energy, which is the sum of a kinetic and potential contribution

$$\frac{\langle RS|T|\Psi_V \rangle}{\langle RS|\Psi_V \rangle} = -\frac{1}{2m} \sum_i \frac{\nabla_i^2 \Psi_V(R,S)}{\Psi_V(R,S)}$$
$$= -\frac{1}{2m} \sum_i \left[\nabla_i^2 U(R) + (\nabla_i U(R))^2 \right]$$

Automatic differentiation (and tailored variants) are employed to efficiently compute it.

LOCAL ENERGY

The potential energy is particularly complicated in Nuclear Physics!

$$\frac{\langle RS|V|\Psi_V\rangle}{\langle RS|\Psi_V\rangle} = \sum_{i$$

We pre-compute the local operators corresponding to exchanging the spin, the isospin, and both.

$$P_{ij}^{\sigma}(R,S) = \frac{\langle RS|\hat{P}_{ij}^{\sigma}|\Psi_{V}\rangle}{\langle RS|\Psi_{V}\rangle} \quad ; \quad P_{ij}^{\tau}(R,S) = \frac{\langle RS|\hat{P}_{ij}^{\tau}|\Psi_{V}\rangle}{\langle RS|\Psi_{V}\rangle} \quad ; \quad P_{ij}^{\sigma\tau}(R,S) = \frac{\langle RS|\hat{P}_{ij}^{\sigma}\hat{P}_{ij}^{\tau}|\Psi_{V}\rangle}{\langle RS|\Psi_{V}\rangle}$$

LOCAL ENERGY

$$\frac{\langle RS|V|\Psi_V\rangle}{\langle RS|\Psi_V\rangle} = \sum_{i$$

LOCAL ENERGY FOR THE DEUTERON

$$|\Phi\rangle = |\uparrow\uparrow\rangle \otimes \frac{1}{\sqrt{2}} (|pn\rangle - |np\rangle) \longrightarrow \begin{cases} \sigma_{12}|\Phi\rangle = |\Phi\rangle \\ \tau_{12}|\Phi\rangle = -3|\Phi\rangle \\ \sigma_{12}\tau_{12}|\Phi\rangle = -3|\Phi\rangle \end{cases}$$

Therefore, the potential becomes effectively spin-isospin independent

$$\frac{\langle RS|V|\Psi_V\rangle}{\langle RS|\Psi_V\rangle} = \sum_{i$$

SOLVING THE DEUTERON

• Three-nucleons potential adjusted to reproduce the energy of ³H.



R. Schiavilla, AL, PRC 103, 054003 (2021)

VMC+NQS: UNDERSTAND NUCLEAR FORCES



A. Gnech, et al., Phys. Rev. Lett. 133 (2024) 14, 142501

NEUTRON STARS





BACKUP

GITHUB REPOSITORY

https://github.com/alelovato/EIC_School/

Let us consider the prototypal problem of a collection of A independent (decoupled) quantum Harmonic oscillators, in N dimensions with unit mass

$$H = -\frac{1}{2} \sum_{i=1}^{A} \nabla_{i}^{2} + \sum_{i=1}^{A} \frac{\mathbf{r}_{i}^{2}}{2}$$

We assume a variational wave function of the exponential form

$$\Psi_V(R) \equiv \langle R | \Psi_V \rangle = \exp\left(-b \sum_{i=1}^A \mathbf{r}_i^2\right)$$

Note that the exact ground state is recovered by taking b = 1/2

$$\Psi_0(R) = \exp\left(-\frac{1}{2}\sum_{i=1}^A \mathbf{r}_i^2\right) \longrightarrow E_0 = A \times N \times \frac{1}{2}$$

The local energy is the sum of the kinetic and potential contributions

$$E_L(R) \equiv \frac{\langle R|H|\Psi_V \rangle}{\langle R|\Psi_V \rangle} = \frac{\langle R|T|\Psi_V \rangle}{\langle R|\Psi_V \rangle} + \frac{\langle R|V|\Psi_V \rangle}{\langle R|\Psi_V \rangle}$$

The kinetic energy involves the second derivative of the variational wave function

$$T_L(R) = -\frac{1}{2} \sum_{i=1}^{A} \frac{\nabla_i^2 \Psi_V(R)}{\Psi_V(R)}$$

Let us first compute the first derivative analytically

$$\nabla_i \Psi_V(R) = -2b\mathbf{r}_i \Psi_V(R)$$

So that:

$$\nabla_i^2 \Psi_V(R) = (-2Nb + 4b^2 \mathbf{r}_i^2) \Psi_V(R) \longrightarrow T_L(R) = \sum_{i=1}^A (bN - 2b^2 \mathbf{r}_i^2)$$

The potential energy is more immediate to evaluate, as it is diagonal in coordinate space

$$V_L(R) = \sum_{i=1}^A \frac{\mathbf{r}_i^2}{2}$$

The total energy then reads

$$E_L(R) = \sum_{i=1}^{A} \left[bN + \left(\frac{1}{2} - 2b^2\right) \mathbf{r}_i^2 \right]$$

Question: what happens when b = 1/2?

Question 2: do you notice anything "strange"?

The potential energy is more immediate to evaluate, as it is diagonal in coordinate space

$$V_L(R) = \sum_{i=1}^A \frac{\mathbf{r}_i^2}{2}$$

The total energy then reads

$$E_L(R) = \sum_{i=1}^{A} \left[bN + \left(\frac{1}{2} - 2b^2\right) \mathbf{r}_i^2 \right]$$

Question: what happens when b = 1/2?

$$E_L(R) = A \times N \times \frac{1}{2} = E_0$$

Question 2: do you notice anything "strange"?

ZERO-VARIANCE PRINCIPLE

The Monte Carlo variance vanishes for exact variational wave functions

$$\sigma_{E_V} = \sqrt{\frac{\langle E_L^2 \rangle - \langle E_L \rangle^2}{N_s - 1}} \quad ; \quad \langle E_L \rangle \simeq \frac{1}{N_s} \sum_{R_n} E_L(R_n) \quad ; \quad \langle E_L^2 \rangle \simeq \frac{1}{N_s} \sum_{R_n} E_L^2(R_n)$$

For an exact variational wave function, the local energy is independent of the sample

$$E_L(R) \equiv \frac{\langle R|H|\Psi_V\rangle}{\langle R|\Psi_V\rangle} = E_0 \frac{\langle R|\Psi_V\rangle}{\langle R|\Psi_V\rangle} = E_0$$

As a consequence:

$$\begin{cases} \langle E_L \rangle = E_0 \\ \langle E_L^2 \rangle = E_0^2 \end{cases} \longrightarrow \sigma_{E_V} = 0$$

REFERENCES

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

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

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

J. R. Moreno, et al., PRL 125, 076402 (2022)