

Formalizm Matrix-Product States dla kwantowego modelu Isinga

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Zdefiniowanie problemu

Układ fizyczny

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L} c_{(\sigma_1 \dots \sigma_L)} |\sigma_1 \dots \sigma_L\rangle$$



Hamiltonian układu

$$\hat{H} = - \sum_{n=1}^{L-1} \hat{S}_n^x \hat{S}_{n+1}^x - \gamma \sum_{n=1}^L \hat{S}_n^z$$

gdzie:

\hat{S}_n^x, \hat{S}_n^z - operatory spinowe

$$\gamma = \frac{\hbar}{J}$$

Formalizm Matrix-Product States(MPS)

MPS dla otwartych warunków brzegowych(OBC)

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L} \sum_{a_1, \dots, a_{L-1}}^{d_1, \dots, d_L \quad D_1, \dots, D_{L-1}} M_{1, a_1}^{\sigma_1} M_{a_1, a_2}^{\sigma_2} \dots M_{a_{L-2}, a_{L-1}}^{\sigma_{L-1}} M_{a_{L-1}, 1}^{\sigma_L} |\sigma_1 \dots \sigma_L\rangle$$

MPS dla okresowych warunków brzegowych(PBC)

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L} \sum_{a_1, \dots, a_L}^{d_1, \dots, d_L \quad D_1, \dots, D_L} M_{a_1, a_2}^{\sigma_1} M_{a_2, a_3}^{\sigma_2} \dots M_{a_{L-1}, a_L}^{\sigma_{L-1}} M_{a_L, a_1}^{\sigma_L} |\sigma_1 \dots \sigma_L\rangle$$

Zapis dowolnego stanu w postaci MPS

Wejście

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L} c_{(\sigma_1 \dots \sigma_L)} |\sigma_1 \dots \sigma_L\rangle$$

Krok 1

$$c_{(\sigma_1 \dots \sigma_L)} = \Psi_{\sigma_1, (\sigma_2 \dots \sigma_L)} = \sum_{a_1}^{D_1} U_{\sigma_1, a_1} S_{a_1, a_1} V_{a_1, (\sigma_2 \dots \sigma_L)}^\dagger$$

$$U_{\sigma_1, a_1} = A_{1, a_1}^{\sigma_1}$$

$$c_{a_1, (\sigma_2 \dots \sigma_L)} = S_{a_1, a_1} V_{a_1, (\sigma_2 \dots \sigma_L)}^\dagger$$

Zapis dowolnego stanu w postaci MPS

Krok 2

$$c_{a_1,(\sigma_2 \dots \sigma_L)} = \Psi_{(a_1 \sigma_2),(\sigma_3 \dots \sigma_L)} = \sum_{a_2}^{D_2} U_{(a_1 \sigma_2),a_2} S_{a_2,a_2} V^\dagger_{a_2,(\sigma_3 \dots \sigma_L)}$$

$$U_{(a_1 \sigma_2),a_2} = A_{a_1,a_2}^{\sigma_2}$$

$$c_{a_2,(\sigma_3 \dots \sigma_L)} = S_{a_2,a_2} V^\dagger_{a_2,(\sigma_3 \dots \sigma_L)}$$

Wyjście

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L}^{d_1, \dots, d_L} \sum_{a_1, \dots, a_{L-1}}^{D_1, \dots, D_{L-1}} A_{1,a_1}^{\sigma_1} A_{a_1,a_2}^{\sigma_2} \dots A_{a_{L-2},a_{L-1}}^{\sigma_{L-1}} A_{a_{L-1},1}^{\sigma_L} |\sigma_1 \dots \sigma_L\rangle$$

Lewostronnie unormowany MPS

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L} \sum_{a_1, \dots, a_{L-1}}^{d_1, \dots, d_L, D_1, \dots, D_{L-1}} A_{1, a_1}^{\sigma_1} A_{a_1, a_2}^{\sigma_2} \dots A_{a_{L-2}, a_{L-1}}^{\sigma_{L-1}} A_{a_{L-1}, 1}^{\sigma_L} |\sigma_1 \dots \sigma_L\rangle$$

$$\begin{aligned} \sum_{\sigma_n} \left(A^{\sigma_n \dagger} A^{\sigma_n} \right)_{a_n, a'_n} &= \sum_{a_{n-1}, \sigma_n} A_{a_n, a_{n-1}}^{\sigma_n \dagger} A_{a_{n-1}, a'_n}^{\sigma_n} \\ &= \sum_{a_{n-1}, \sigma_n} U_{a_n, (a_{n-1}, \sigma_n)}^\dagger U_{(a_{n-1}, \sigma_n), a'_n} = \delta_{a_n, a'_n} \end{aligned}$$

Iloczyn wewnętrzny

$$\langle \phi | \psi \rangle = \sum_{\sigma_1, \dots, \sigma_L} \tilde{M}^{\sigma_L \dagger} \dots \tilde{M}^{\sigma_1 \dagger} M^{\sigma_1} \dots M^{\sigma_L}$$

Wartość oczekiwana

$$\langle \psi | \hat{O} | \psi \rangle = \sum_{\substack{\sigma_1, \dots, \sigma_L \\ \sigma'_i}} \tilde{M}^{\sigma_L \dagger} \dots \tilde{M}^{\sigma_i \dagger} \dots \tilde{M}^{\sigma_1 \dagger} O^{\sigma_i, \sigma'_i} M^{\sigma_1} \dots M^{\sigma'_i} \dots M^{\sigma_L}$$

gdzie:

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L} M^{\sigma_1} \dots M^{\sigma_L} |\sigma_1, \dots, \sigma_L\rangle$$

$$|\phi\rangle = \sum_{\sigma_1, \dots, \sigma_L} \tilde{M}^{\sigma_1} \dots \tilde{M}^{\sigma_L} |\sigma_1, \dots, \sigma_L\rangle$$

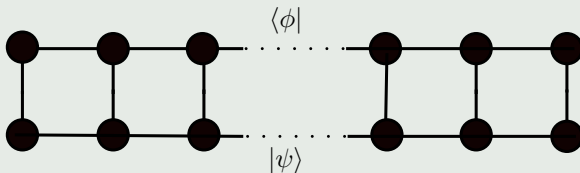
$$\hat{O} = \sum_{\sigma_i, \sigma'_i} O^{\sigma_i, \sigma'_i} |\sigma_i\rangle \langle \sigma'_i|$$

Graficzna reprezentacja

MPS

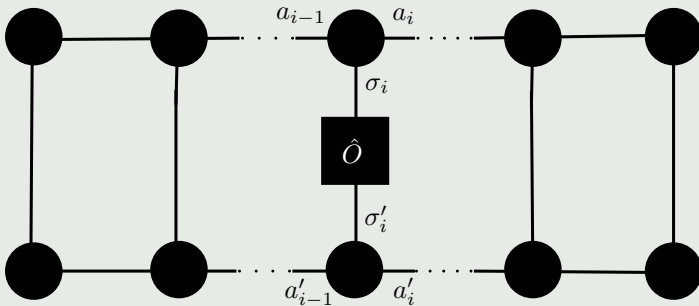


Iloczyn wewnętrzny



Graficzna reprezentacja

Wartość oczekiwana



Metoda wariacyjna w dziedzinie MPS (VMPS)

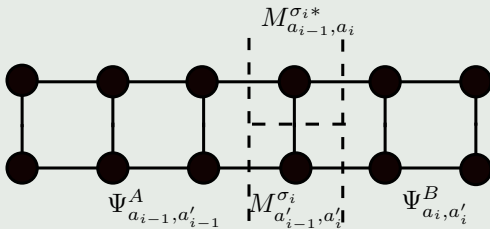
Podstawa metody

$$\langle \psi | \hat{H} | \psi \rangle = \sum_n \sum_m c_m^* c_n E_n \langle \psi_m | \psi_n \rangle = \sum_n E_n |c_n|^2 \geq E_0$$

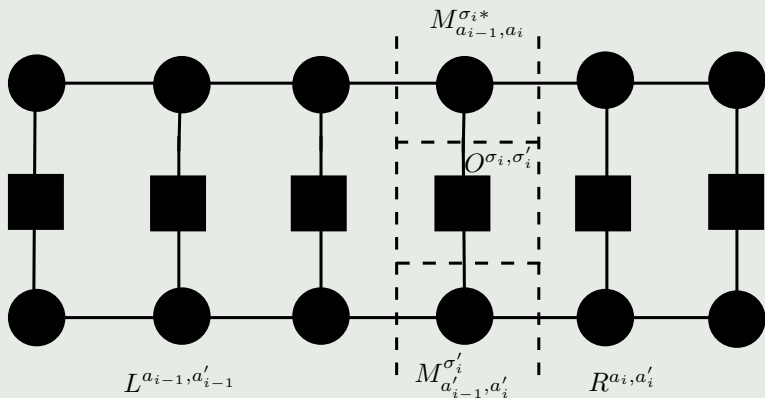
Metoda mnożników Lagrange'a

$$L = \langle \psi | \hat{H} | \psi \rangle - \lambda \langle \psi | \psi \rangle$$

Graficzna reprezentacja $\langle \psi | \psi \rangle$



Graficzna reprezentacja $\langle \psi | \hat{H} | \psi \rangle$



Minimalizacja na i -tym węźle

$$\begin{aligned}
 0 &= \sum_{\sigma'_i} \sum_{a'_{i-1}, a'_i} L^{a_{i-1}, a'_{i-1}} O^{\sigma_i, \sigma'_i} R^{a_i, a'_i} M^{\sigma'_i}_{a'_{i-1}, a'_i} \\
 &- \lambda \sum_{a'_{i-1}, a'_i} \Psi^{A}_{a_{i-1}, a'_{i-1}} M^{\sigma_i}_{a'_{i-1}, a'_i} \Psi^{B}_{a_i, a'_i}
 \end{aligned}$$

Minimalizacja na i -tym węźle

$$H_{(\sigma_i a_{i-1} a_i)(\sigma'_i a'_{i-1} a'_i)}{}^{ef} = \sum_{b_{i-1}, b_i} L^{a_{i-1}, a'_{i-1}} O_{b_{i-1}, b_i}^{\sigma_i, \sigma'_i} R^{a_i, a'_i}$$

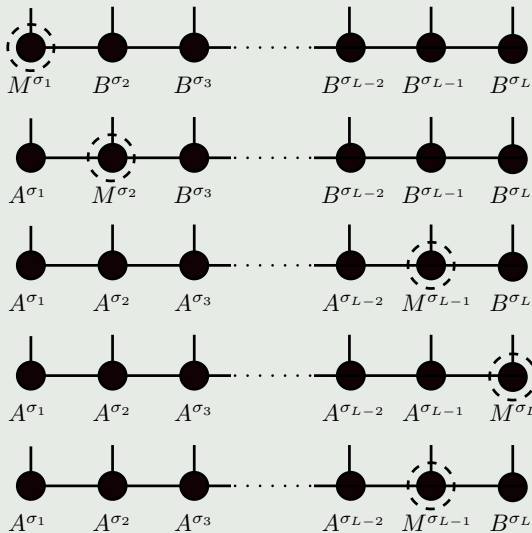
$$N_{(\sigma_i a_{i-1} a_i)(\sigma'_i a'_{i-1} a'_i)} = \Psi_{a_{i-1}, a'_{i-1}}^A \Psi_{a_i, a'_i}^B \delta_{\sigma_i, \sigma'_i}$$

$$\nu_{(\sigma'_i a'_{i-1} a'_i)} = M_{a'_{i-1}, a'_i}^{\sigma'_i}$$

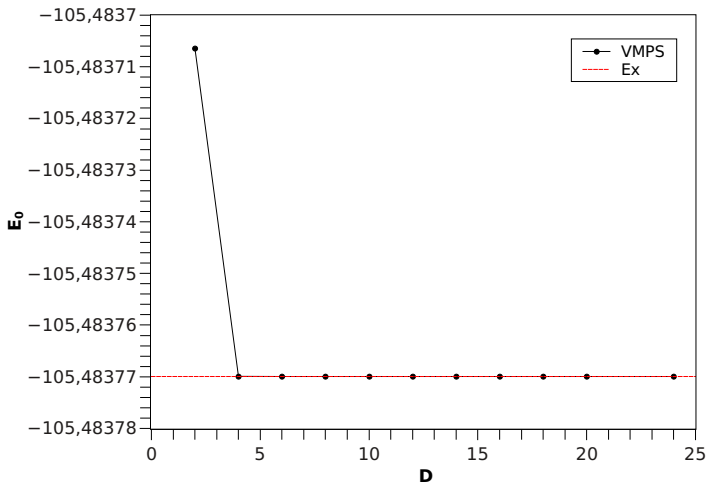
Końcowy wzór

$$H^{ef} \nu = \lambda N \nu$$

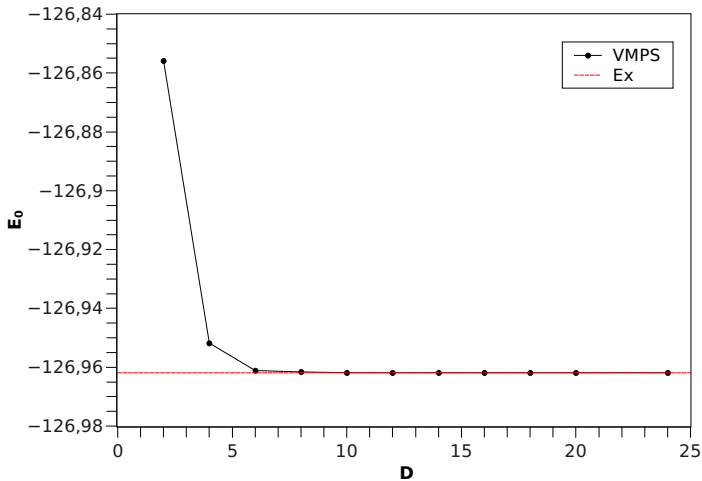
Algorytm VMPS



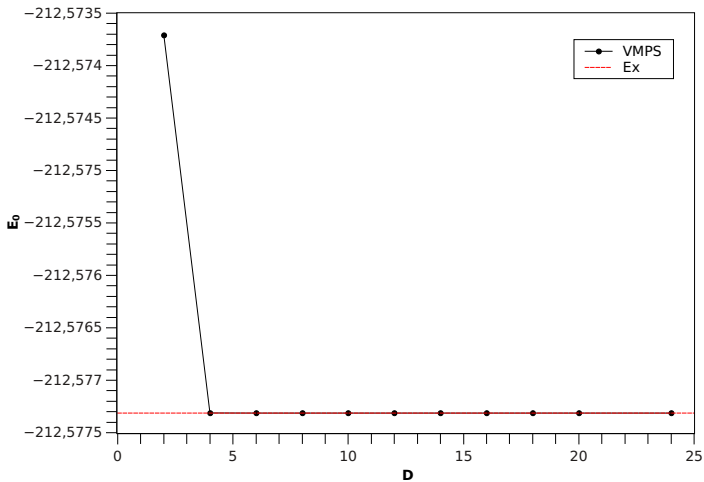
$E_0(D)$, dla $L = 100$, $\gamma = 0.5$



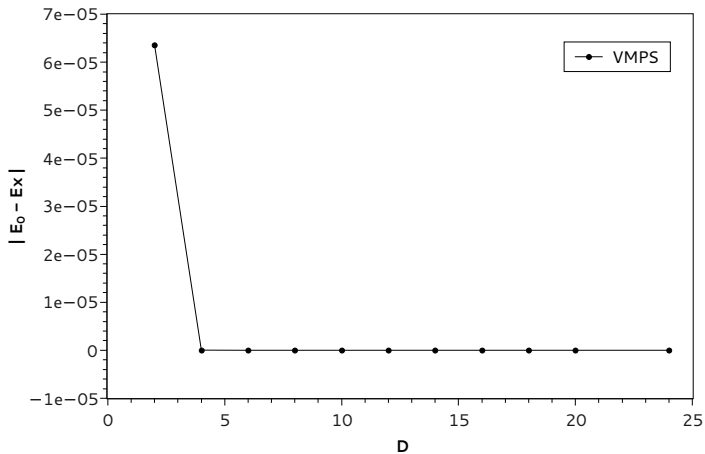
$E_0(D)$, dla $L = 100$, $\gamma = 1$



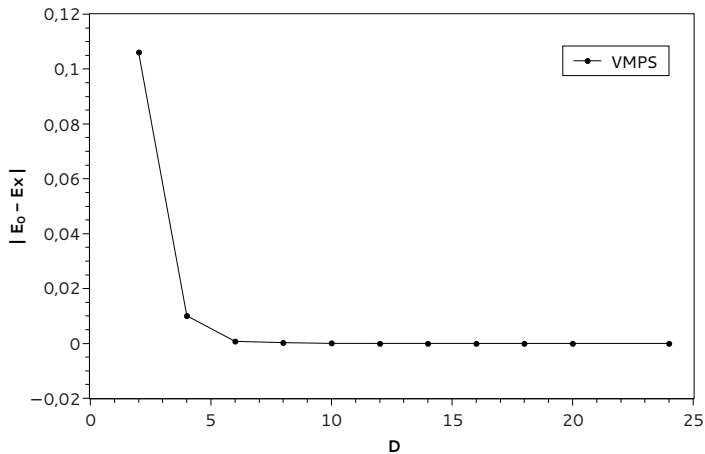
$E_0(D)$, dla $L = 100$, $\gamma = 2$



$|E_0 - E_x|(D)$, dla $L = 100$, $\gamma = 0.5$



$|E_0 - E_x|(D)$, dla $L = 100$, $\gamma = 1$



$|E_0 - E_x|(D)$, dla $L = 100$, $\gamma = 2$

