

Matrix Product State Methods for Excitations

Jesse Osborne

University of Queensland
School of Mathematics and Physics

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Infinite matrix product states

$$|\Psi\rangle = \begin{array}{ccccccc} & A_1^{s_1} & A_2^{s_2} & A_3^{s_3} & \dots & A_N^{s_N} & \\ & \circ & \circ & \circ & \dots & \circ & \\ & | & | & | & & | & \\ & & & & & & \end{array} .$$

Enforce translation invariance and take $N \rightarrow \infty$,

$$|\Psi\rangle = \dots \begin{array}{cccccc} & A^{s_1} & A^{s_2} & A^{s_3} & A^{s_4} & A^{s_5} & \\ & \circ & \circ & \circ & \circ & \circ & \\ & | & | & | & | & | & \\ & & & & & & \end{array} \dots .$$

- Observables can be calculated by fixed-point relations.
- No finite-size effects: only finite entanglement (bond dim.).
- How to get information about excitations?
(Dynamics vs statics.)

MPS excitation ansatz

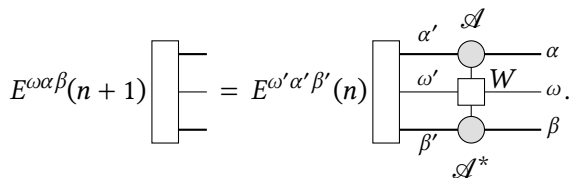
$$|\Phi_k[B]\rangle = \sum_n e^{ikn} \dots \text{---} \underset{\text{A}}{\text{O}} \text{---} \underset{\text{A}}{\text{O}} \text{---} \underset{\text{B}}{\text{O}} \text{---} \underset{\tilde{\text{A}}}{\text{O}} \text{---} \underset{\tilde{\text{A}}}{\text{O}} \text{---} \dots$$

n

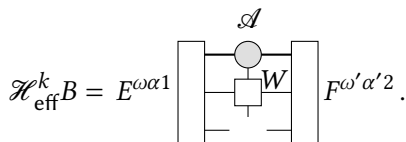
$$= \dots \text{---} \underset{\mathcal{A}}{\text{O}} \text{---} \underset{\mathcal{A}}{\text{O}} \text{---} \underset{\mathcal{A}}{\text{O}} \text{---} \underset{\mathcal{A}}{\text{O}} \text{---} \underset{\mathcal{A}}{\text{O}} \text{---} \dots, \quad \mathcal{A} = \begin{pmatrix} e^{ik} A & B \\ 0 & \tilde{A} \end{pmatrix}.$$

- Generalisation of single mode approximation $B = \hat{a}^\dagger A$.
- B can be optimised for each k .
- Block triangular structure: reminiscent of MPOs.
- Reuse ground state data instead of starting from scratch.
- Non-injective: needed to differentiate gap from boundary term.

EA fixed point equations



- Only need $\omega' \leq \omega$, $\alpha' \leq \alpha$, $\beta' \leq \beta \Rightarrow$ Solve recursively.
- Same algorithm as iMPS, but with extra indices α, β .



JO and I. P. McCulloch, in preparation.

Cf. L. Michel and I. P. McCulloch, arXiv:1008.4667.

Expectation values

Expectation values are polynomials in system size L :

$$\langle \hat{H} \rangle = \frac{\langle \Phi | \hat{H} | \Phi \rangle}{\langle \Phi | \Phi \rangle} = \frac{EL^2 + \Delta L}{L} = EL + \Delta.$$

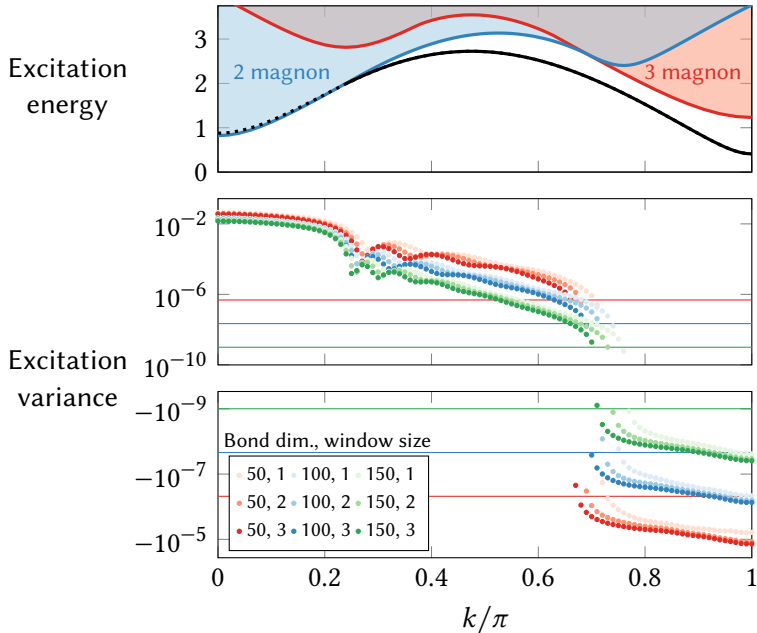
Can calculate the variance:

$$\langle \hat{H}^2 \rangle - \langle \hat{H} \rangle^2 = \sigma_E^2 L + \sigma_\Delta^2.$$

Since we only have an approximation to the GS:

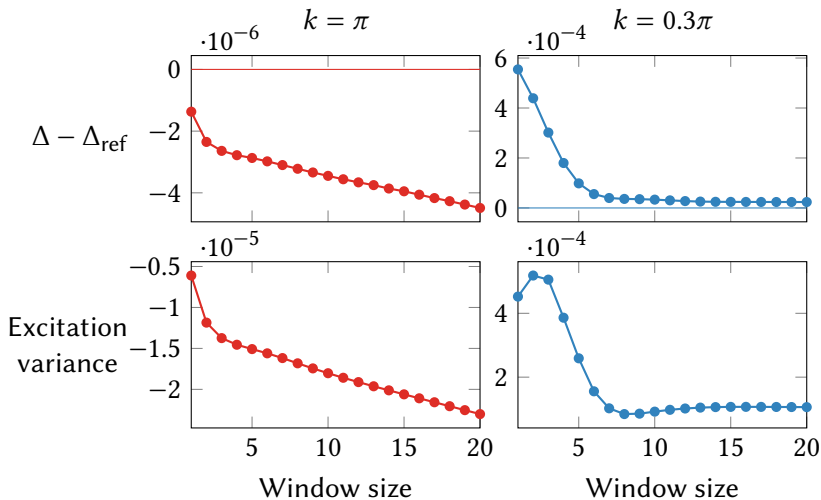
- Δ can be less than the actual value.
- σ_Δ^2 can be negative.
- Bound by error in GS.

Spin-1 Heisenberg model



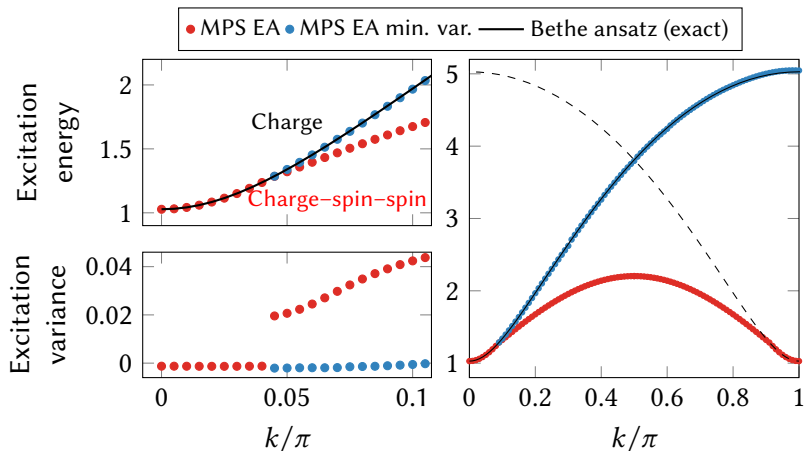
Spin-1 Heisenberg model

Tuning window size (bond dim. 50)



Minimising energy variance

Hubbard model chargons, $U/t = 5$



Compare V. Zauner-Stauber et al., Phys. Rev. B **97**, 235155 (2018).

Multi-particle states

$$\begin{aligned}
 |Y_{k_1, k_2}\rangle = & \sum_n \left[q_1 \sum_{m>n} e^{i(k_1 n + k_2 m)} \dots \text{---} \underset{\substack{| \\ n}}{\text{red}} \text{---} \dots \text{---} \underset{\substack{| \\ m}}{\text{blue}} \text{---} \dots \right. \\
 & + q_2 \sum_{m<n} e^{i(k_2 m + k_1 n)} \dots \text{---} \underset{\substack{| \\ m}}{\text{blue}} \text{---} \dots \text{---} \underset{\substack{| \\ n}}{\text{red}} \text{---} \dots \\
 & \left. + e^{i(k_1 + k_2)n} \dots \text{---} \underset{\substack{| \\ n}}{\text{pink}} \text{---} \text{---} \text{---} \text{---} \text{---} \dots \right],
 \end{aligned}$$

‘Correction window’ W

$$\mathcal{A} = \begin{pmatrix} e^{i(k_1 + k_2)A} & q_1 B & q_2 C & W \\ & e^{ik_2 A} & & C \\ & & e^{ik_1 A} & B \\ & & & A \end{pmatrix}.$$

Generalises to ≥ 3 excitations.

Conclusion

- Excitation ansatz wavefunctions are triangular MPSs.
- Can calculate fixed point relations recursively.

Thanks to:

- Ian McCulloch (NTHU).

Code available: <https://github.com/mptoolkit>

Documentation(?): <https://mptoolkit.qusim.net>