

Lower Bounding Ground State Energies of Local Hamiltonians

Tractable relaxations of the quantum marginal problem

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European Research Council

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Local Hamiltonians and Marginal States

$$H = \sum_{i \in I} h_{a_i} \quad a_i \text{ local patches on a lattice } \Lambda$$

$$\begin{aligned} E_0 = \min \langle \psi | H | \psi \rangle &= \min \sum_{i \in I} \langle \psi | h_{a_i} | \psi \rangle = \\ \text{s.t. } \|\psi\| = 1 &\quad \text{s.t. } \|\psi\| = 1 \end{aligned}$$

$$\begin{aligned} &= \min \sum_{i \in I} \text{Tr}(h_{a_i} \rho_{a_i}) \\ &\quad \text{s.t. } \{\rho_{a_i}\}_{i \in I} \text{ are all marginals of some } \psi \in \mathcal{H}_\Lambda \\ &\quad \text{Tr}(\rho_{a_i}) = 1 \end{aligned}$$

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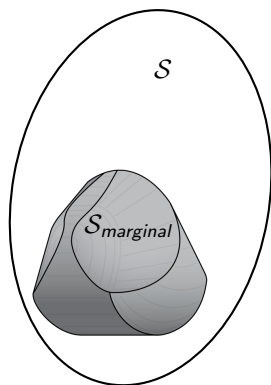
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Convex Sets of Quantum States



$$\{\rho_{a_i} \geq 0\}_{i \in I} \quad \longleftrightarrow \quad \vec{\rho} \in S$$

$$H = \sum_{i \in I} h_{a_i} \quad \longleftrightarrow \quad \vec{H}$$

$$\sum_{i \in I} \text{Tr}(h_{a_i} \rho_{a_i}) \quad \longleftrightarrow \quad \vec{H} \cdot \vec{\rho}$$

It's Hard...

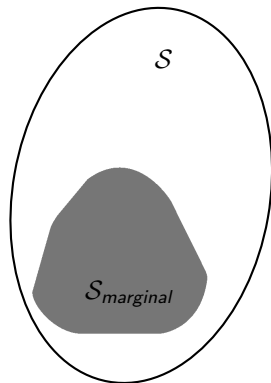
- ▶ Local energy problem is QMA-complete
[Aharonov et al. Comm Math Phys (2009)]

- ▶ Variational Methods:

$$E_{\mathcal{V}} = \min_{\vec{\rho} \in \mathcal{V}} \vec{H} \cdot \vec{\rho} \implies E_0 \leq E_{\mathcal{V}}$$

- ▶ Relaxation methods:

$$E_{\mathcal{R}} = \min_{\vec{\rho} \in \mathcal{R}} \vec{H} \cdot \vec{\rho} \implies E_0 \geq E_{\mathcal{R}}$$



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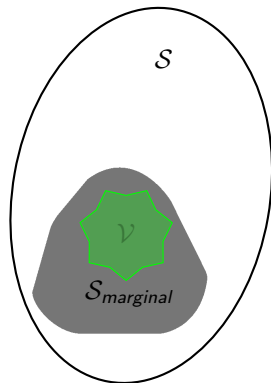
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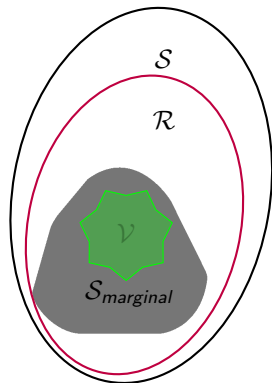
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In this Talk

- ▶ Our relaxation method
 - ▶ expand exact constraint into a hierarchy of constraints
 - ▶ coarse-grain and compress the variables
 - ▶ use tensor networks for coarse-graining maps
- ▶ Results for 1D translation-invariant models

How to relax?

$$E_0 = \min \langle \psi | H | \psi \rangle = \min \sum_{i \in I} \text{Tr}(h_{a_i} \rho_{a_i}) \geq \min \sum_{i \in I} \text{Tr}(h_{a_i} \rho_{a_i})$$

s.t. $\|\psi\| = 1$ s.t. $\{\rho_{a_i}\}_{i \in I} \in \mathcal{S}_{\text{marginal}}$ s.t. ... ???

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Which constraints NOT to relax?

- ▶ Nearest neighbors in 1D: $H = \sum_i h_{i,i+1}$
- ▶ $\{\rho_{i,i+1}\}_i$ marginals of $\psi \in \mathcal{H}_\Lambda \Rightarrow \{\rho_{i,i+1}\}_i$ **mutually compatible**

$$\text{Tr}_1(\rho_{1,2}) = \rho_2 = \text{Tr}_{1,3,4,\dots,N}(|\psi\rangle\langle\psi|) = \text{Tr}_3(\rho_{2,3})$$

- ▶ Keep the conditions $\text{Tr}_i(\rho_{i,i+1}) = \text{Tr}_{i+2}(\rho_{i+1,i+2})$

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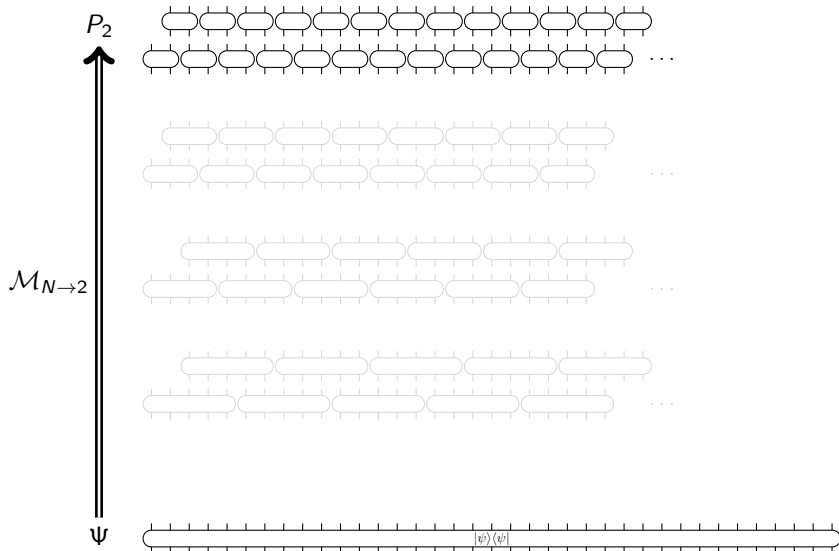
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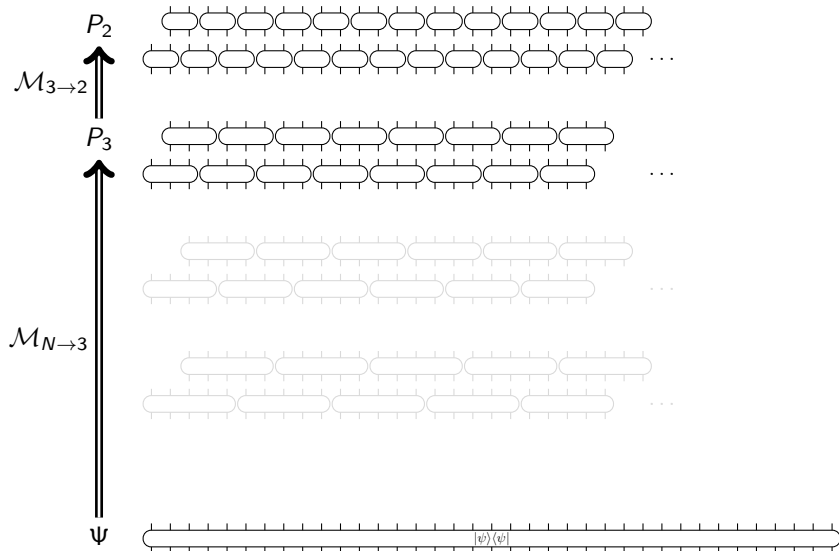
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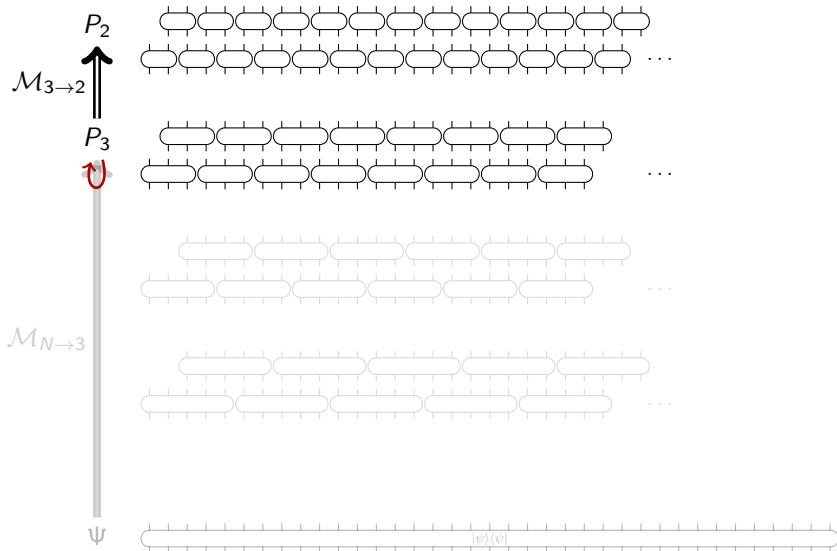
Ingredient 1: a hierarchy of constraints



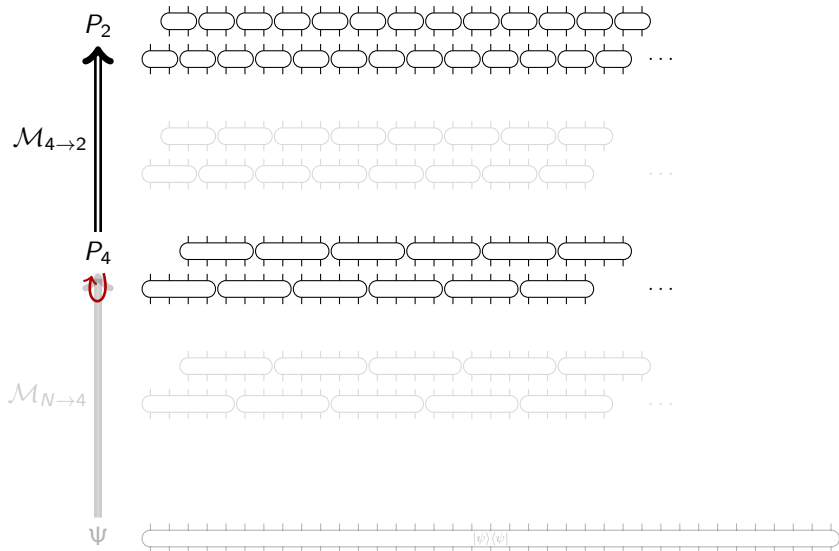
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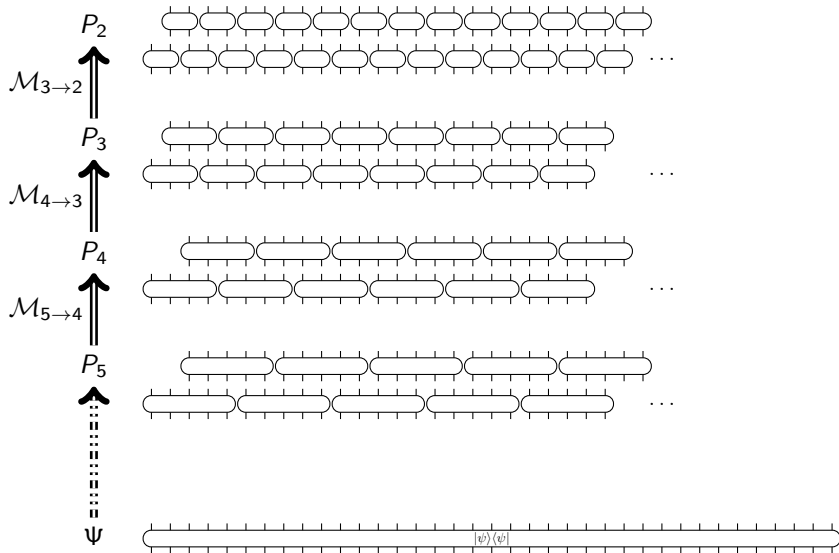
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Ingredient 2: relaxation via coarse-graining and compression

$$\min \mathcal{F}(\rho_{1,2,3}, \rho_{2,3,4})$$

$$\text{s.t. } \rho_{1,2,3} = \omega_{1,2,3,4}$$

$$\rho_{2,3,4} = \omega_{1,2,3,4}$$

$$\left\{ \tilde{\omega}_{1,0,4} \geq 0 \right\} \supseteq \left\{ \begin{array}{c} \text{---} \\ \text{---} \\ \omega_{1,2,3,4} \\ \text{---} \\ \text{---} \end{array} \left| \omega_{1,2,3,4} \geq 0 \right. \right\}$$

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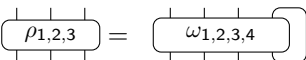
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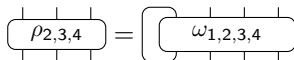
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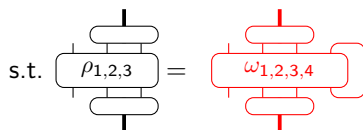
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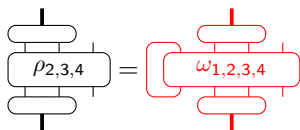
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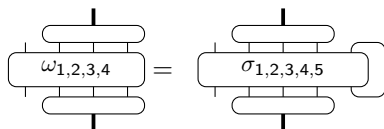
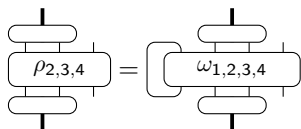
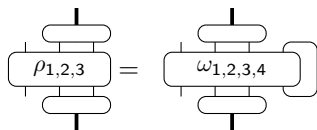
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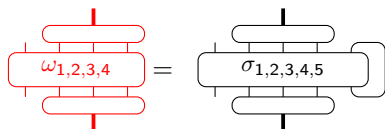
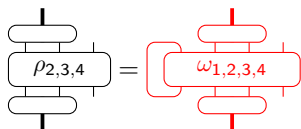
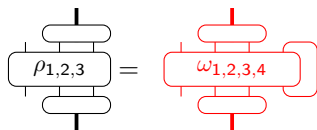
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The last ingredient: composable coarse-graining maps



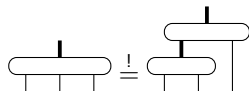
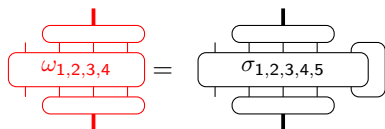
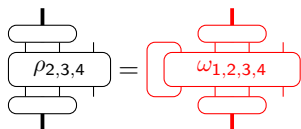
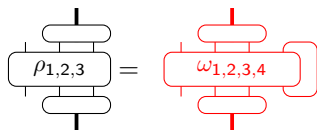
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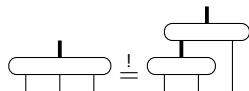
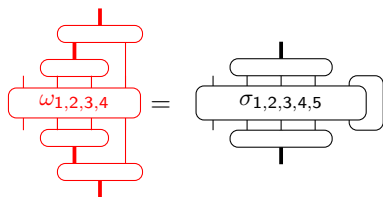
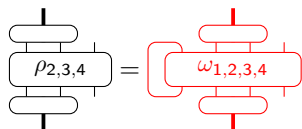
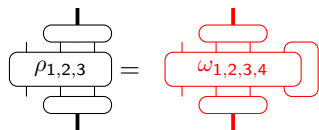
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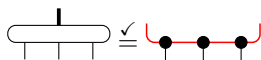
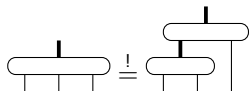
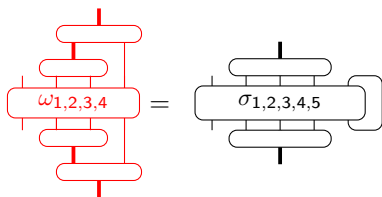
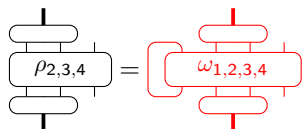
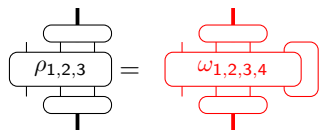
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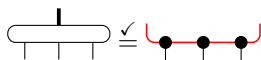
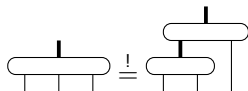
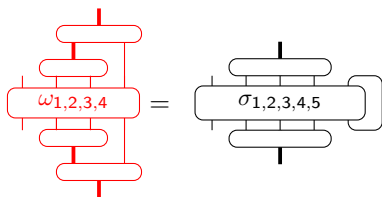
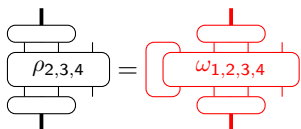
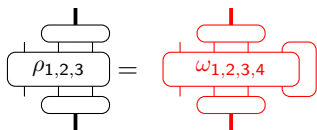
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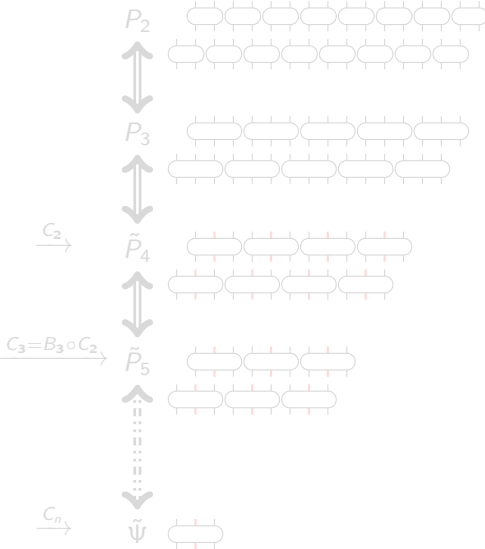
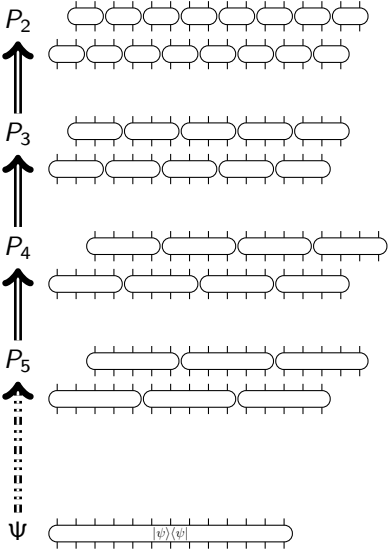
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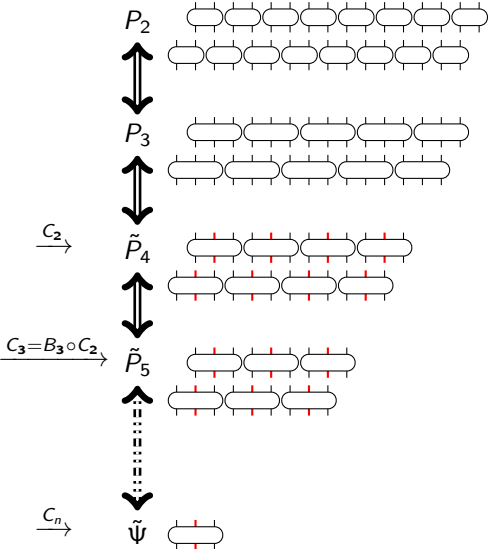
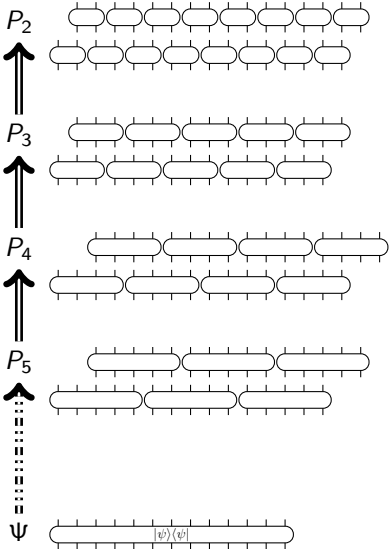


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Method overview



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Application:
The 1D Translation-Invariant
Local Energy Problem

The translation-invariant problem

$$H = \sum_{i=-\infty}^{\infty} h_{i,i+1}, \quad h_{i,i+1} \equiv h \quad \forall i$$

$$e_{TI} := \min \text{Tr}(h\rho_{\{1,2\}}) \\ \text{s.t. } \rho_{\{1,2\}} \in \mathcal{S}_{\text{marginal}}^{TI}$$

Relax
 \implies

$$e_{LTI}^n := \min \text{Tr}(h\rho_{\{1,2\}}) \\ \text{s.t. } \text{Tr}\rho_{\{1,2\}} = 1, \\ \rho_{\{1,2\}} = \text{Tr}_{\{3,4,\dots,n\}}\sigma_{\{1,2,\dots,n\}}, \\ \sigma \geq 0, \\ \text{Tr}_{\{1\}}\sigma = \text{Tr}_{\{n\}}\sigma$$



Models:

- ▶ $H_{TFI}(h_z) = -\sum_i X_i X_{i+1} - h_z \sum_i Z_i$ (Transverse field Ising)
- ▶ $H_{XXZ}^S(\Delta) = \sum_i X_i X_{i+1} + Y_i Y_{i+1} + \Delta Z_i Z_{i+1}$ (XXZ $S = 1/2$ and $S = 1$)

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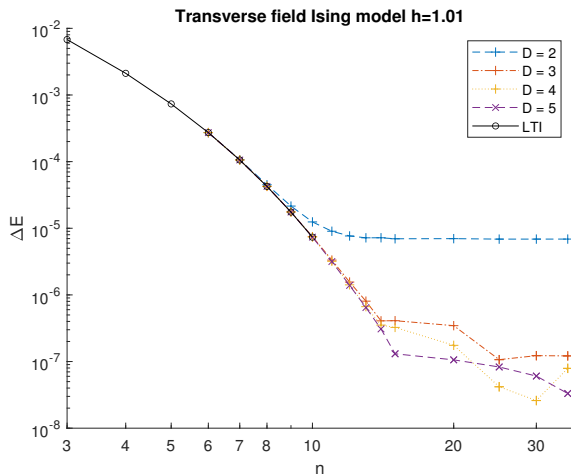
Translation invariant relaxation scheme

Exact 5-body Loc.T.I. problem	Relaxation due to applied C.G. maps	Compressed relaxation

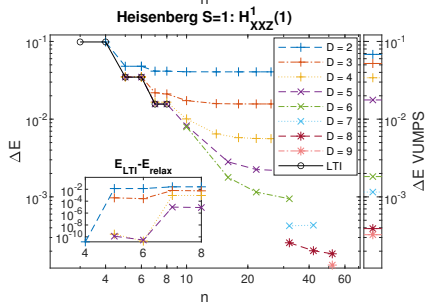
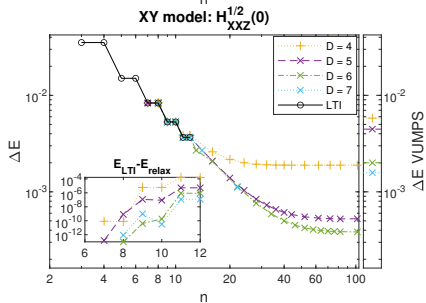
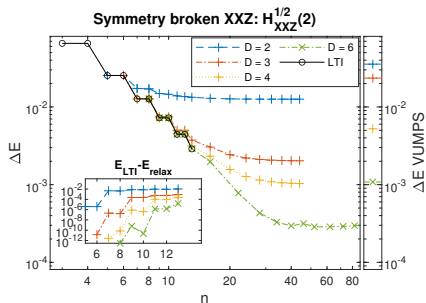
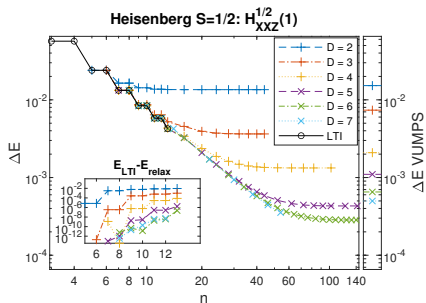
The MPS tensor is the VUMPS solution for the same Hamiltonian [Zauner-Stauber et al. PRB (2018)].

Results

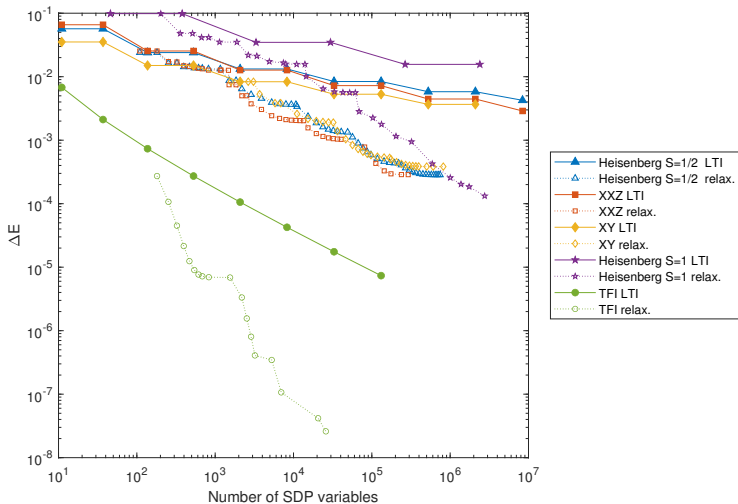
Results: transverse field Ising model



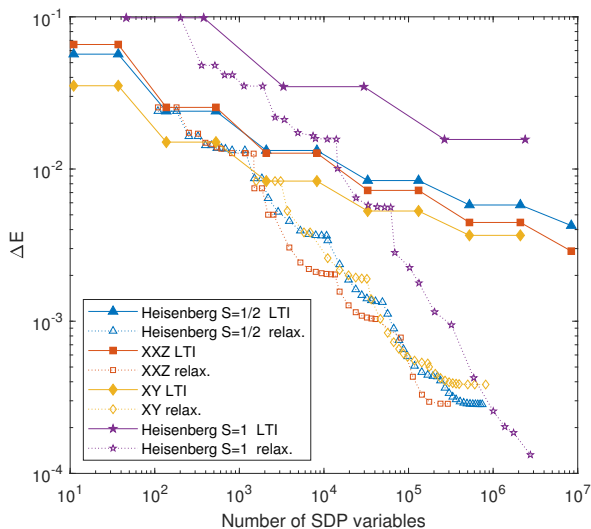
Results: XXZ models for $S=1/2$ and $S=1$ systems



Results: memory scaling vs. precision



Results: Memory scaling vs. precision



Discussion

Why does this work?

- ▶ we enforce the constraints only in subspaces spanned by MPS tensors
- ▶ keep constraints relevant to the given Hamiltonian
- ▶ optimal variational MPS \Leftrightarrow optimal coarse-graining for relaxation

Not in this talk

- ▶ Works also for finite systems
- ▶ One can optimize over the coarse-graining maps
- ▶ Issues with SDP solvers: memory scaling is not everything
- ▶ Ideas beyond 1D and beyond lattice spin systems

Thanks!

Not in this talk

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Thanks!

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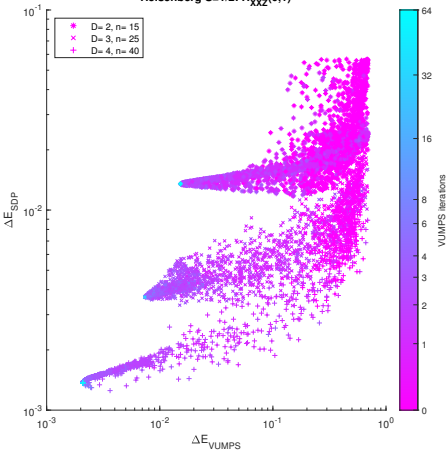
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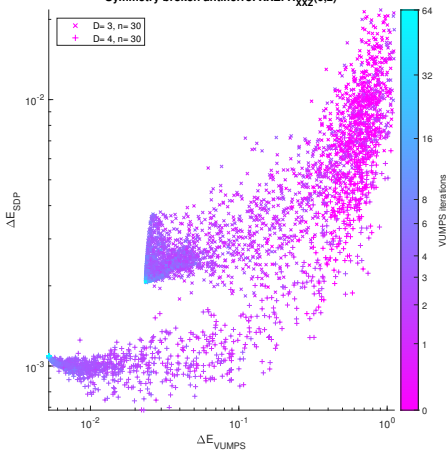
Thanks!

Results: Relaxation precision vs. VUMPS precision

Heisenberg S=1/2: $H_{XXZ}^{1/2}(0,1)$



Symmetry broken anti.ferro. XXZ: $H_{XXZ}^{1/2}(0,2)$



Near phase boundary

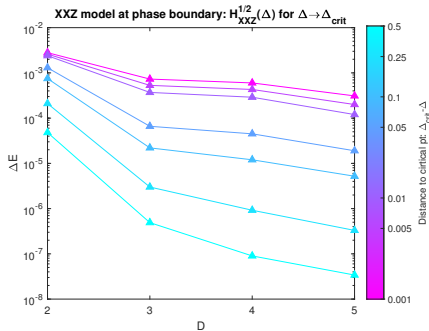
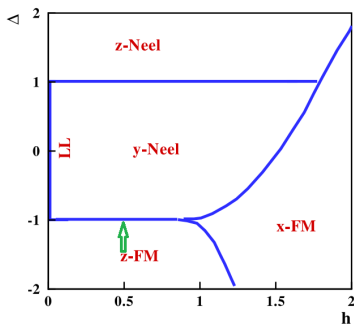


Fig. 1 Ground state magnetic phase diagram of the 1D spin-1/2 XXZ model in a transverse magnetic field. The z -Neel and y -Neel phases represent the staggered magnetization along z and y axis respectively. In the z -FM and x -FM regions, the long-range magnetization exist along the z and x axis, respectively (Color figure online)

Moradmard et al. J Supercond Nov Magn 27, (2014).