

Lower bounds on ground-state energies of local Hamiltonians through the renormalization group

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Outline

$H = \sum_i h_i$, h_i act locally, $\dim \mathcal{H} = d^N$

$$\text{Find } \min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle$$

Intro:

- ▶ Locality \Rightarrow **variational** vs. **relaxation** (aka **bootstrap**) approaches
- ▶ **Variational** principle + **renormalization group**
 - \Rightarrow tensor-networks methods (e.g. DMRG)
- ▶ ...still need good lower bounds,
 - existing methods scale as $\exp(n)$

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Our method:

- ▶ **Relaxation + renormalization group** \Rightarrow efficient* lower bounds

"Corner of Hilbert Space" — The Variational Approach

$H = \sum_i h_i$, h_i act locally, $\dim \mathcal{H} = d^N$

$$\min_{\psi \in \mathcal{C} \subset \mathcal{H}} \langle \psi | H | \psi \rangle \approx \min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle$$

"Corner of Hilbert Space" — The Variational Approach

$H = \sum_i h_i$, h_i act locally,

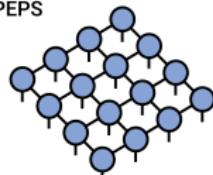
$$\dim \mathcal{H} = d^N$$

$$\min_{\psi \in \mathcal{C} \subset \mathcal{H}} \langle \psi | H | \psi \rangle \geq \min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle \geq ?$$

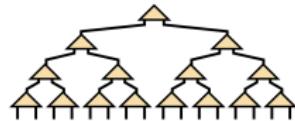
Matrix Product State /
Tensor Train



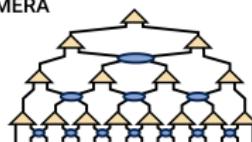
PEPS



Tree Tensor Network /
Hierarchical Tucker



MERA



[tensornetwork.org]

[Cirac and Verstraete, J.Phys.A (2009)]

Lower Bounds — Relaxation Methods

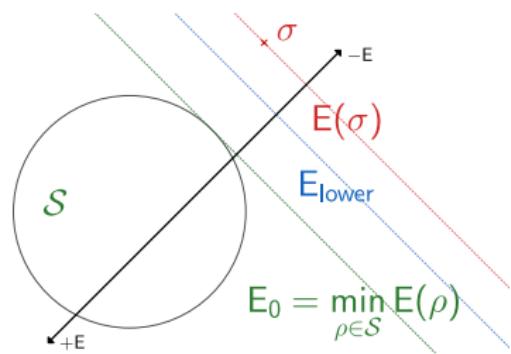
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Applications

- ▶ Central to physics and chemistry:
Certify variational solutions,
benchmark methods
- ▶ Quantum information and foundations:
detection, certification, falsification
- ▶ Quantum complexity theory
 $\text{MAX-}k\text{-SAT} \xrightarrow{\text{quantum}} k\text{-local } H \text{ min. energy}$



Lower Bounds Through Constraints Relaxation

$$\min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle$$

$$H = \sum_i h_i, \quad h_i = h_{a_i} \otimes \mathbb{I}_{a_i^c}$$

Lower Bounds Through Constraints Relaxation

$$\min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle = \min_{\psi \in \mathcal{H}} \sum_i \langle \psi | h_i | \psi \rangle$$

$$H = \sum_i h_i, \quad h_i = h_{a_i} \otimes \mathbb{I}_{a_i^c}$$

Lower Bounds Through Constraints Relaxation

$$\begin{aligned}\min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle &= \min_{\psi \in \mathcal{H}} \sum_i \langle \psi | h_i | \psi \rangle \\ &= \min_{\{\rho_i\} \leftarrow \psi} \sum_i \text{Tr}(h_i \rho_i)\end{aligned}$$

" $\{\rho_i\} \leftarrow \psi$ " \Leftrightarrow there exists a state $\psi \in \mathcal{H}$ such that ρ_i are its reduced states. **Hard!!**

$$H = \sum_i h_i, \quad h_i = h_{a_i} \otimes \mathbb{I}_{a_i^c}$$

ρ_i local reduced state on $a_i \subset \Lambda$

Lower Bounds Through Constraints Relaxation

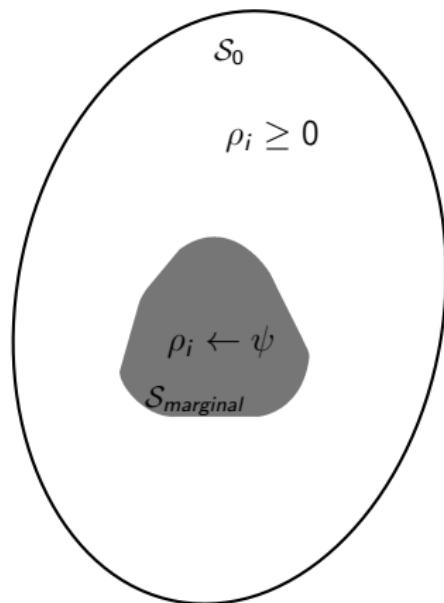
$$\begin{aligned}\min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle &= \min_{\psi \in \mathcal{H}} \sum_i \langle \psi | h_i | \psi \rangle \\ &= \min_{\{\rho_i\} \leftarrow \psi} \sum_i \text{Tr}(h_i \rho_i) \\ &\geq \min_{\{\rho_i\} \in \mathcal{S}_?} \sum_i \text{Tr}(h_i \rho_i) \quad (\text{Relaxation})\end{aligned}$$

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ρ_i local reduced state on $a_i \subset \Lambda$

Convex Sets of Local States



Convex Sets of Local States

XXZ chain:

$$H = \sum_i (\sigma_x^i \sigma_x^{i+1} + \sigma_y^i \sigma_y^{i+1} + \Delta \sigma_z^i \sigma_z^{i+1})$$

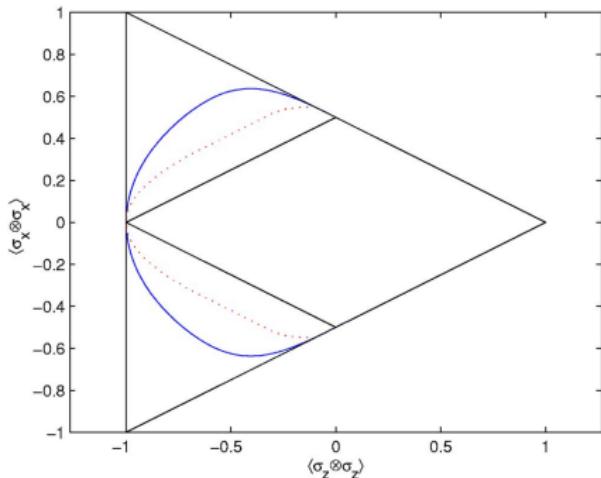
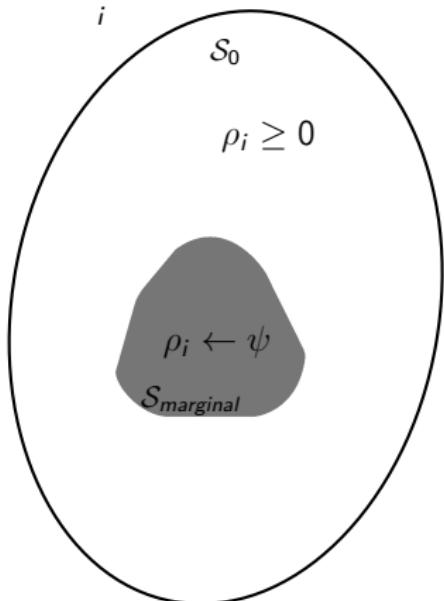
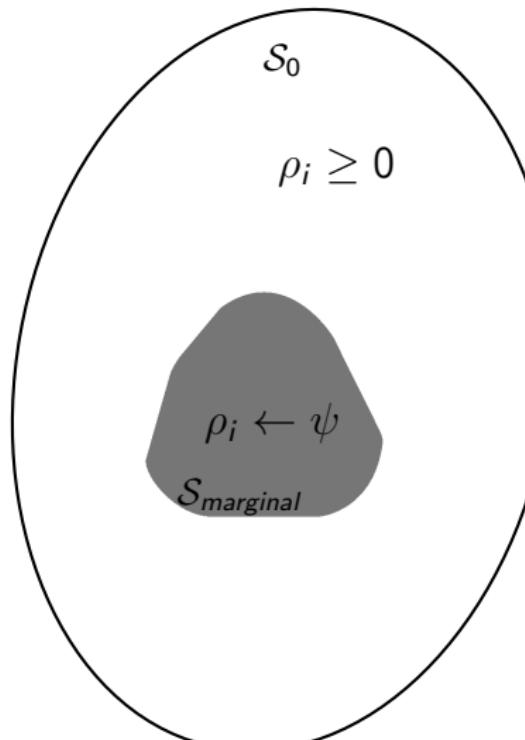


FIG. 1. (Color online) Convex sets of the possible reduced density operators of translational invariant states in the XX-ZZ plane: the big triangle represents all positive density operators, the inner parallelogram represents the separable states, the union of the separable cone and the convex hull of the full curved line is the complete convex set in the case of a 1D geometry, and the dashed lines represent extreme points in the 2D case of a square lattice. The singlet corresponds to the point with coordinates $(-1, -1)$.

[Verstraete and Cirac, PRB (2006)]

Existing methods — Complete Hierarchies

$$\min_{\{\rho_i\} \leftarrow \psi} \sum_i \text{Tr}(h_i \rho_i)$$

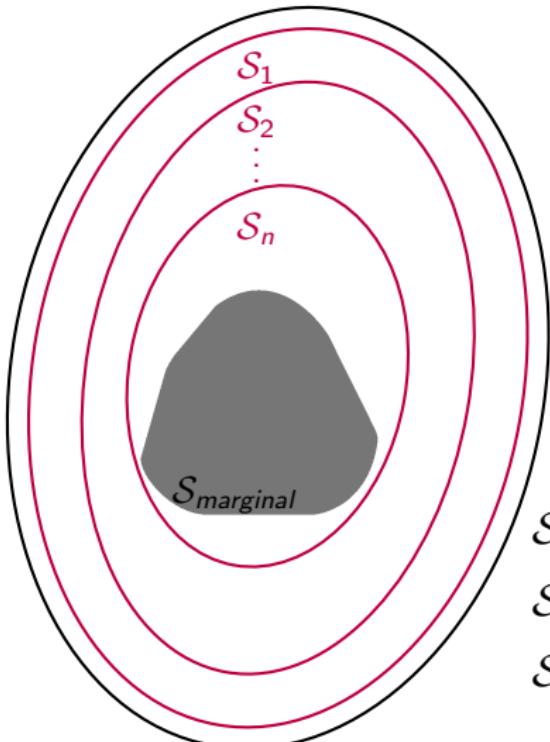


$$S_0 = \{\rho_i \geq 0\}$$

$$S_{marginal} = \{\{\rho_i\} \leftarrow \psi\}$$

Existing methods — Complete Hierarchies

$$\min_{\{\rho_i\} \leftarrow \psi} \sum_i \text{Tr}(h_i \rho_i)$$



$$S_0 = \{\rho_i \geq 0\}$$

$$S_{marginal} = \{\{\rho_i\} \leftarrow \psi\} = S_\infty$$

$$S_0 \supset S_1 \supset S_2 \supset \dots \supset S_n \supset \dots \supset S_{marginal}$$

Existing methods — Complete Hierarchies

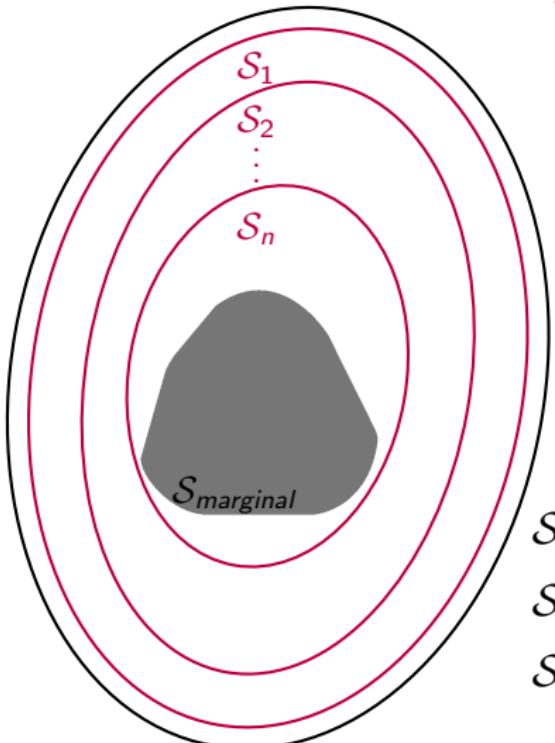
$$\min_{\{\rho_i\} \leftarrow \psi} \sum_i \text{Tr}(h_i \rho_i) \geq \dots \geq \min_{\{\rho_i\} \in \mathcal{S}_n} \sum_i \text{Tr}(h_i \rho_i)$$

⋮

$$\geq \min_{\{\rho_i\} \in \mathcal{S}_2} \sum_i \text{Tr}(h_i \rho_i)$$

$$\geq \min_{\{\rho_i\} \in \mathcal{S}_1} \sum_i \text{Tr}(h_i \rho_i)$$

$$\geq \min_{\{\rho_i\} \in \mathcal{S}_0} \sum_i \text{Tr}(h_i \rho_i)$$



$$\mathcal{S}_0 = \{\rho_i \geq 0\}$$

$$\mathcal{S}_{marginal} = \{\{\rho_i\} \leftarrow \psi\} = \mathcal{S}_\infty$$

$$\mathcal{S}_0 \supset \mathcal{S}_1 \supset \mathcal{S}_2 \supset \dots \supset \mathcal{S}_n \supset \dots \supset \mathcal{S}_{marginal}$$

History and Related Methods

► **Anderson bounds:**

[Anderson, Limits on the Energy of the Antiferromagnetic Ground State, Phys. Rev. 83, (1951)]

► **Quantum chemistry: (RDMT)**

[Mazziotti and Rice, Reduced-Density-Matrix Mechanics, Wiley & Sons (2007)]

[Klyachko, Quantum marginal problem and N-representability, J. Phys: Conf. Ser. 36, (2006)]

► **Quantum information / foundations: (NPA hierarchy)**

Bounding the set of quantum correlations

[Navascués, Pironio & Acín, Phys. Rev. Lett. 98, (2007)]

► **Optimization, complexity theory:**

Lasserre/Parrilo/SOS hierarchy [Lasserre, Global optimization with polynomials and the problem of moments, SIAM J. Optim. 11, (2001)]

► **Bootstrap methods:**

Conformal bootstrap Poland, Rychkov & Vichi, Rev. Mod. Phys. 91, (2019),
Random matrix models Lin, J. High Energ. Phys. 90 (2020) , 3D classical Ising model [Cho et. al. arxiv:2206.12538], Gap in 1D quantum Ising model [Nancarrow & Xin, arXiv:2211.03819] ...

Example: The LTI Hierarchy

Setting:

$H = \sum_i \tau_i(h)$, translation-invariant, nearest-neighbor interactions

$$\begin{aligned} E_{\text{TI}} := \min_{\rho^{(2)}, \psi_{\text{TI}}} & \text{Tr} (h \rho^{(2)}) \\ \text{s.t. } & \rho^{(2)} \leftarrow \psi_{\text{TI}} \end{aligned}$$

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$$E_{\text{TI}} = \min_{\{\rho^{(m)}\}, \psi_{\text{TI}}} \text{Tr}(h\rho^{(2)})$$

$$\text{s.t. } \rho^{(2)} \leftarrow \rho^{(3)} \leftarrow \dots \leftarrow \rho^{(n-1)} \leftarrow \rho^{(n)} \leftarrow \psi_{\text{TI}}$$

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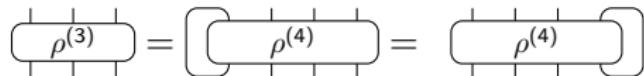
$$E_{\text{TI}} = \min_{\{\rho^{(m)}\}, \psi_{\text{TI}}} \text{Tr}(h\rho^{(2)})$$

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$$\rho^{(m-1)} \leftarrow \rho^{(m)}$$



$$\rho^{(m-1)} = \text{Tr}_L(\rho^{(m)}) = \text{Tr}_R(\rho^{(m)})$$



Example: The LTI Hierarchy

Setting:

$H = \sum_i \tau_i(h)$, translation-invariant, nearest-neighbor interactions

$$E_{\text{TI}} \geq E_{\text{LTI}}(n) = \min_{\{\rho^{(m)}\}, \forall m} \text{Tr}(h\rho^{(2)})$$

$$\text{s.t. } \rho^{(2)} \leftarrow \rho^{(3)} \leftarrow \dots \leftarrow \rho^{(n-1)} \leftarrow \rho^{(n)} \quad \cancel{\leftarrow \psi} \cancel{\leftarrow \text{TI}}$$

$$\rho^{(m-1)} \leftarrow \rho^{(m)}$$

\Updownarrow

$$\rho^{(m-1)} = \text{Tr}_L(\rho^{(m)}) = \text{Tr}_R(\rho^{(m)})$$

$$\boxed{\rho^{(3)}} = \boxed{\rho^{(4)}} = \boxed{\rho^{(4)}}$$

Renormalization of the LTI Constraints

We relax the constraints $\rho^{(3)} \leftarrow \rho^{(4)} \leftarrow \rho^{(5)}$

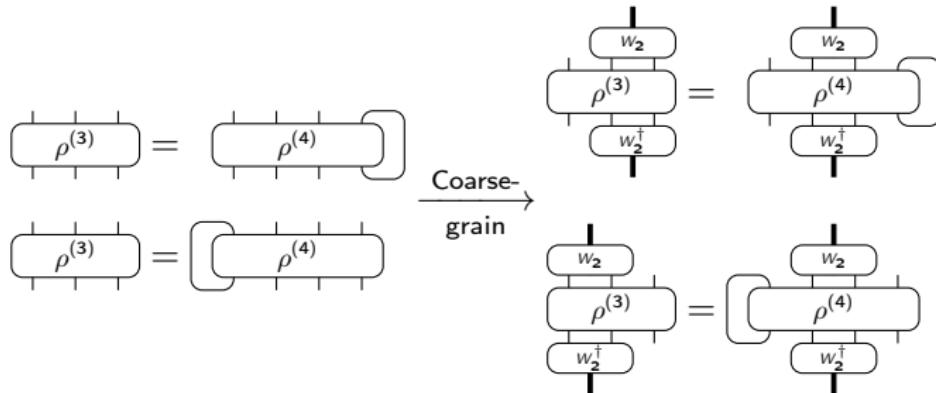
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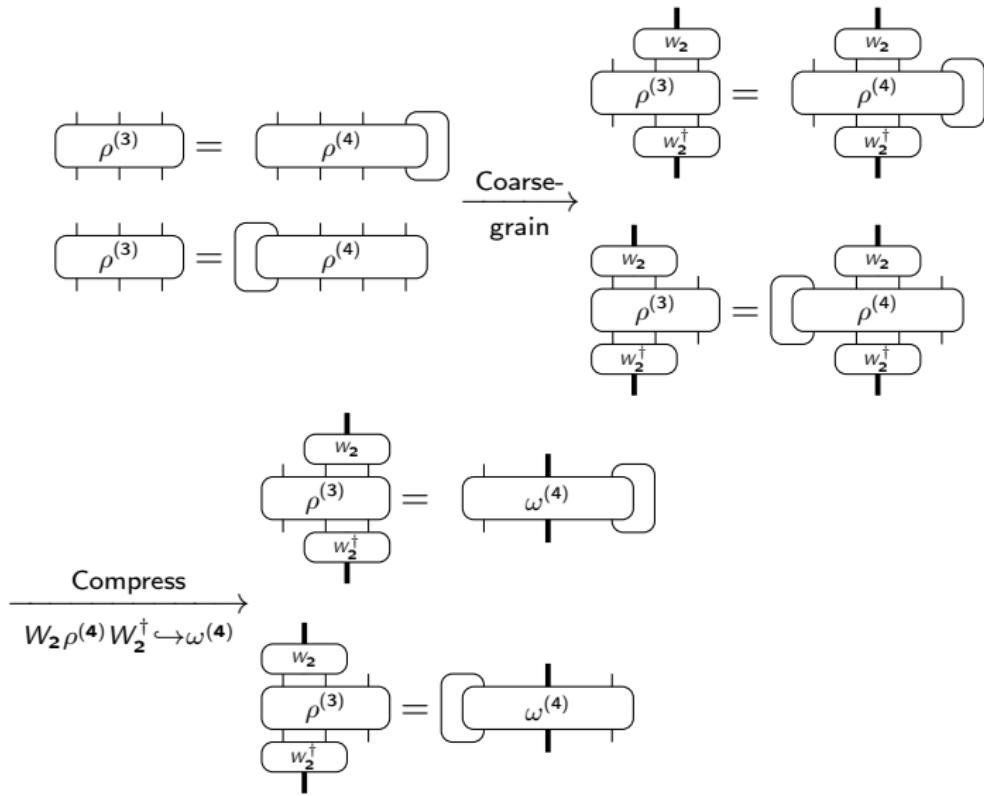
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Coarse-Graining and Compression Step

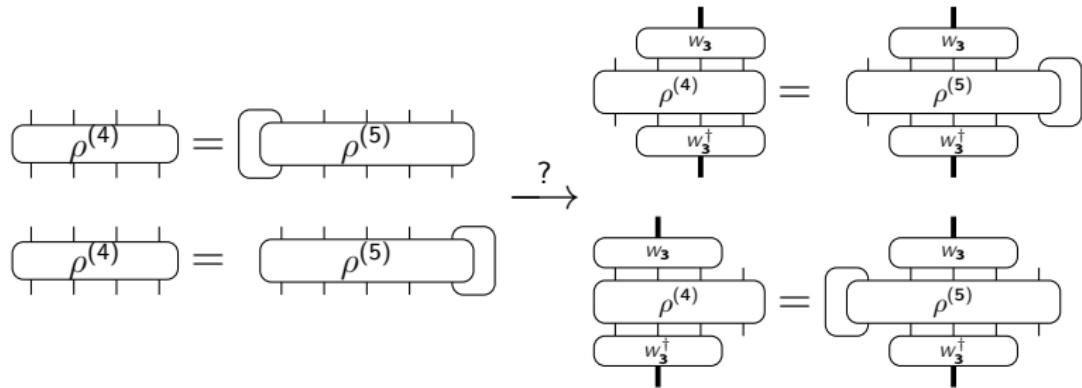


Coarse-Graining and Compression Step

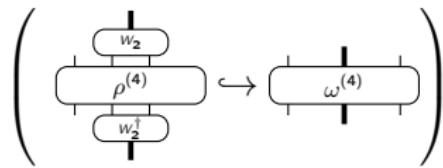


Iterative Coarse-Graining

We want:

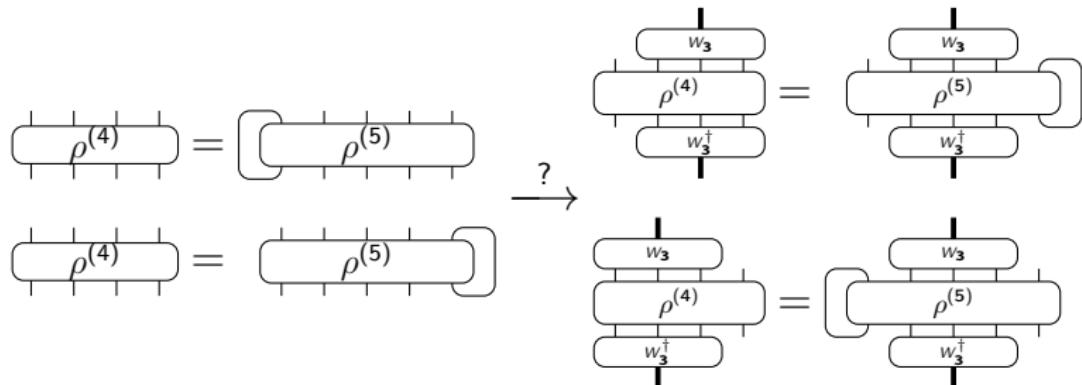


But $\rho^{(4)}$ is no longer available

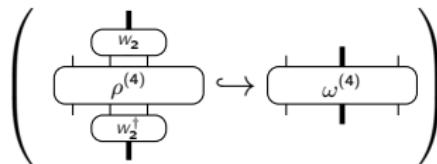


Iterative Coarse-Graining

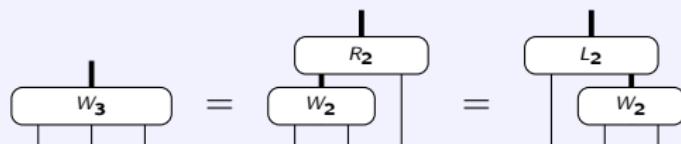
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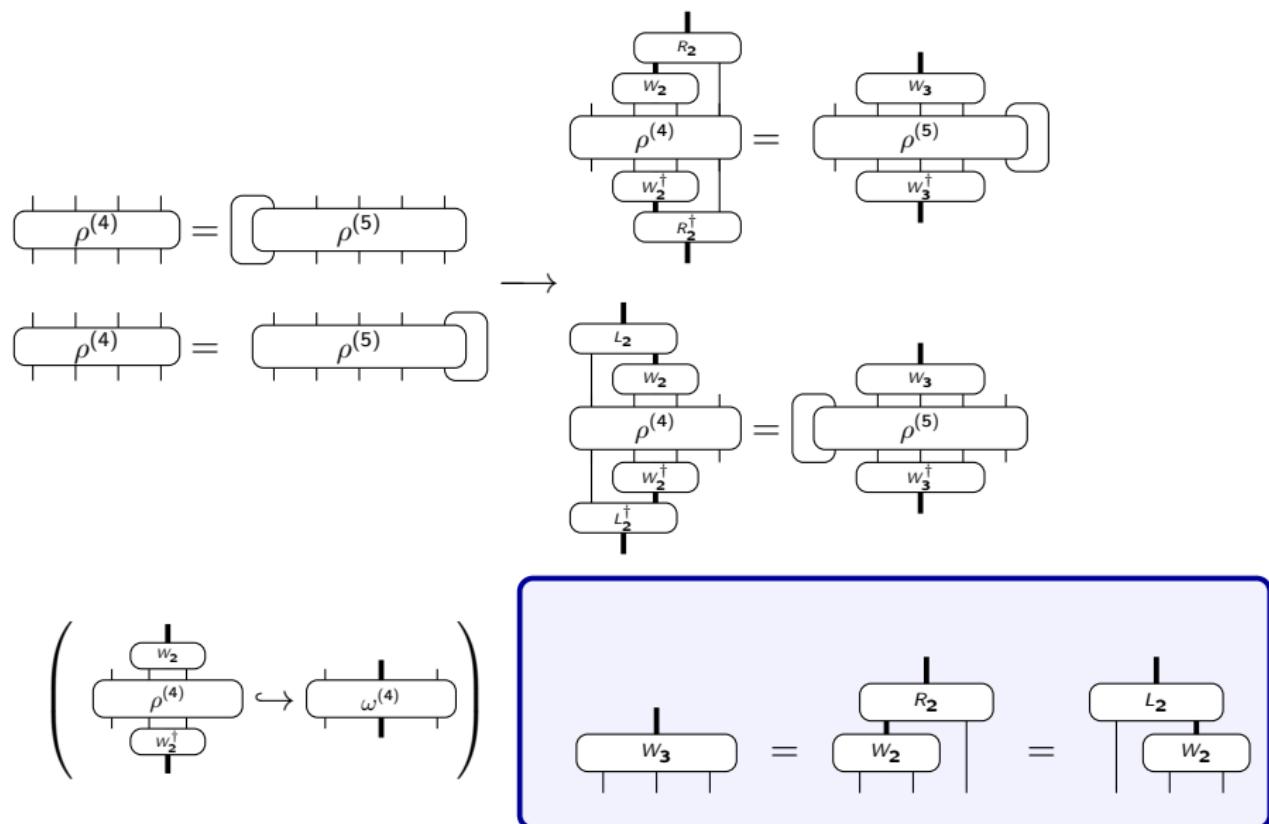
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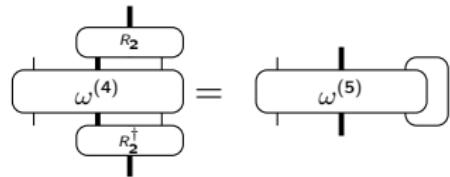
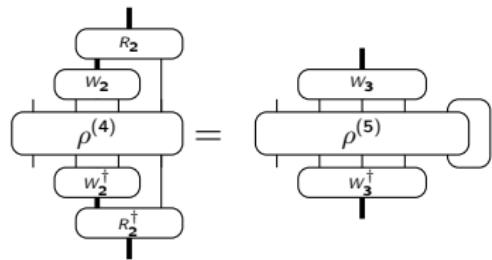
We Need:



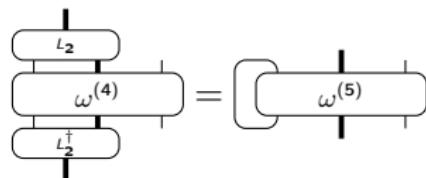
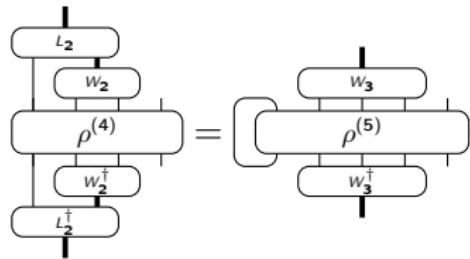
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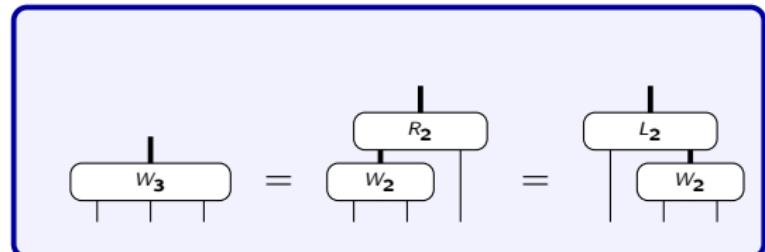
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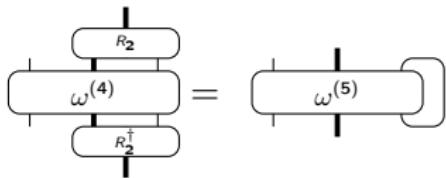
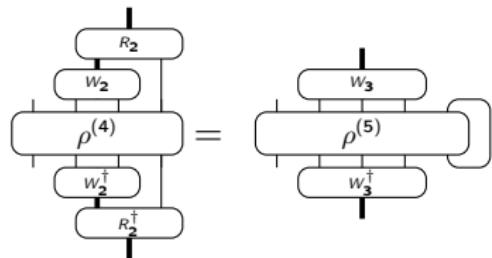
$$\frac{W_2 \rho^{(4)} W_2^\dagger \hookrightarrow \omega^{(4)}}{W_3 \rho^{(5)} W_3^\dagger \hookrightarrow \omega^{(5)}}$$



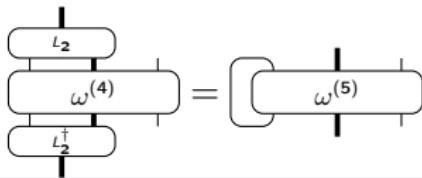
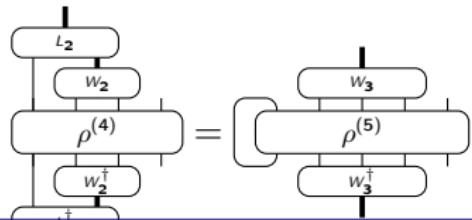
$$\left(\begin{array}{c} W_2 \\ \rho^{(4)} \\ W_2^\dagger \end{array} \right) \hookrightarrow \omega^{(4)}$$



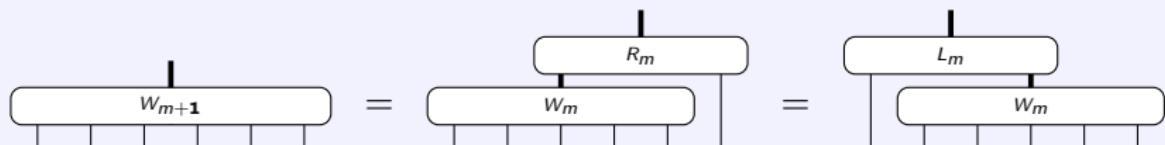
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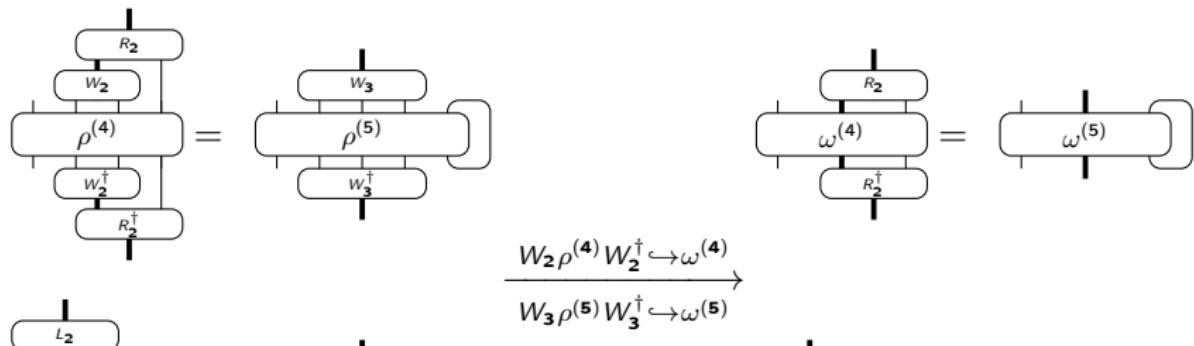
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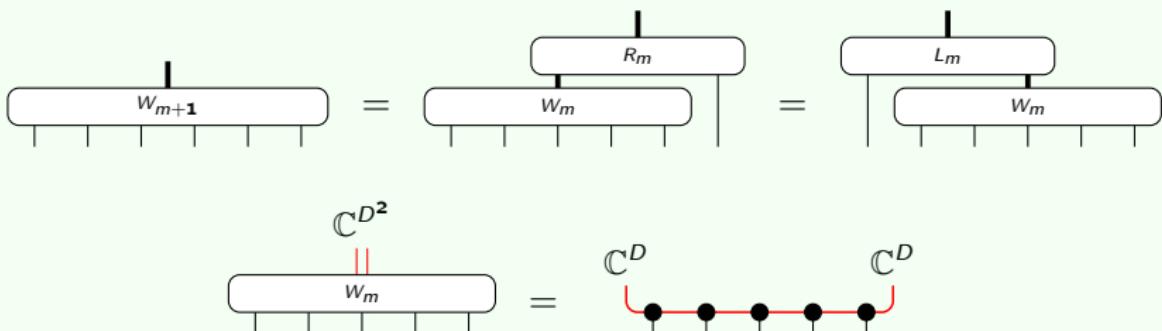
We need for all $m = 1, \dots, n - 2$



Iterative Coarse-Graining



The solution: Matrix Product States



Recap

$$E_{\text{TI}} \geq E_{\text{LTI}}(n) = \min_{\{\rho^{(m)}\}} \text{Tr} \left(h \rho^{(2)} \right)$$
$$\text{s.t. } \rho^{(2)} \leftarrow \rho^{(3)} \leftarrow \rho^{(4)} \leftarrow \rho^{(5)} \dots \leftarrow \rho^{(n)}$$

Recap

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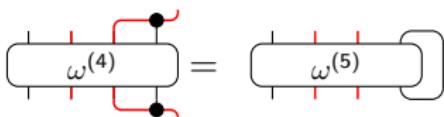
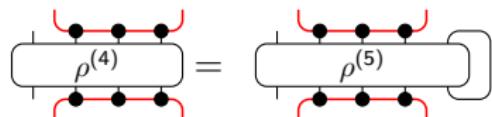
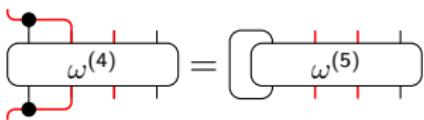
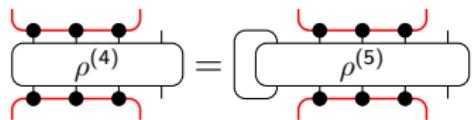
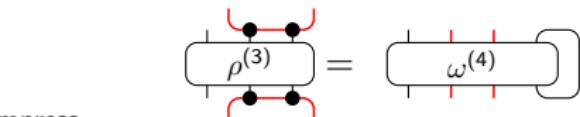
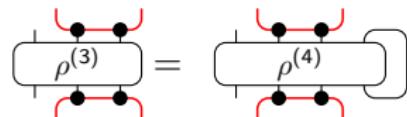
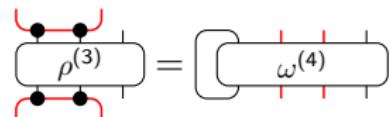
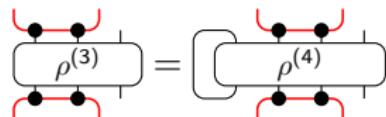
$$E_{\text{TI}} \geq E_{\text{LTI}}(n) \geq E_{\text{relax.}}(n, D) = \min_{\rho^{(2)}, \rho^{(3)}, \{\omega^{(m)}\}} \text{Tr} \left(h \rho^{(2)} \right)$$

s.t. $\rho^{(2)} \leftarrow \rho^{(3)} \rightarrow \leftarrow \omega^{(4)} \rightarrow \leftarrow \omega^{(5)} \dots \rightarrow \leftarrow \omega^{(n)}$

Recap

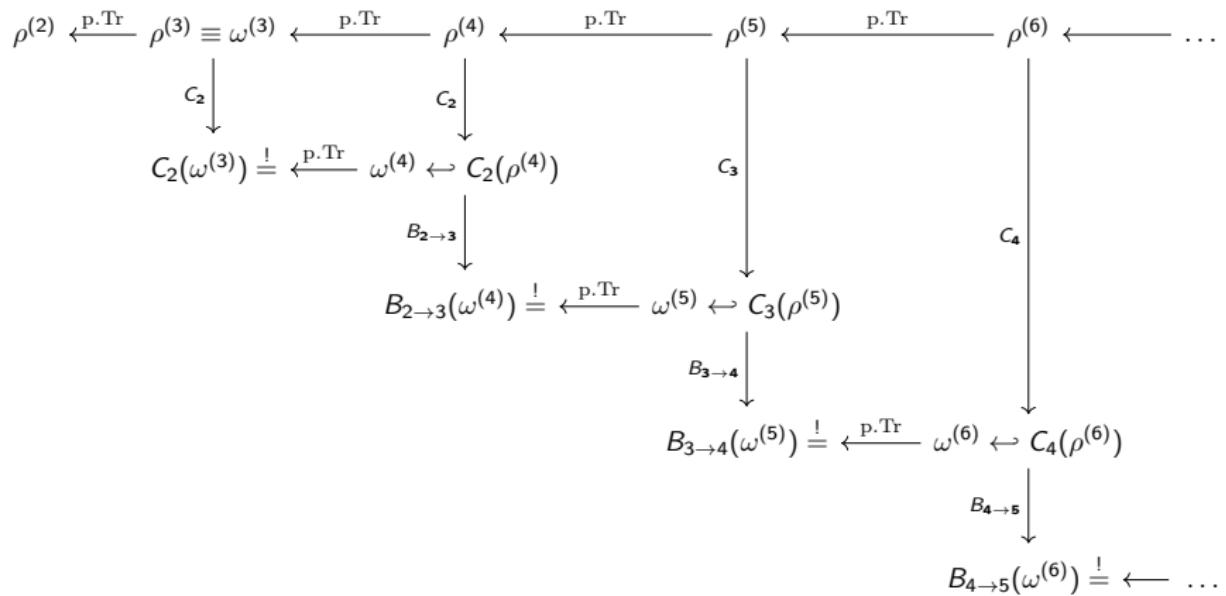
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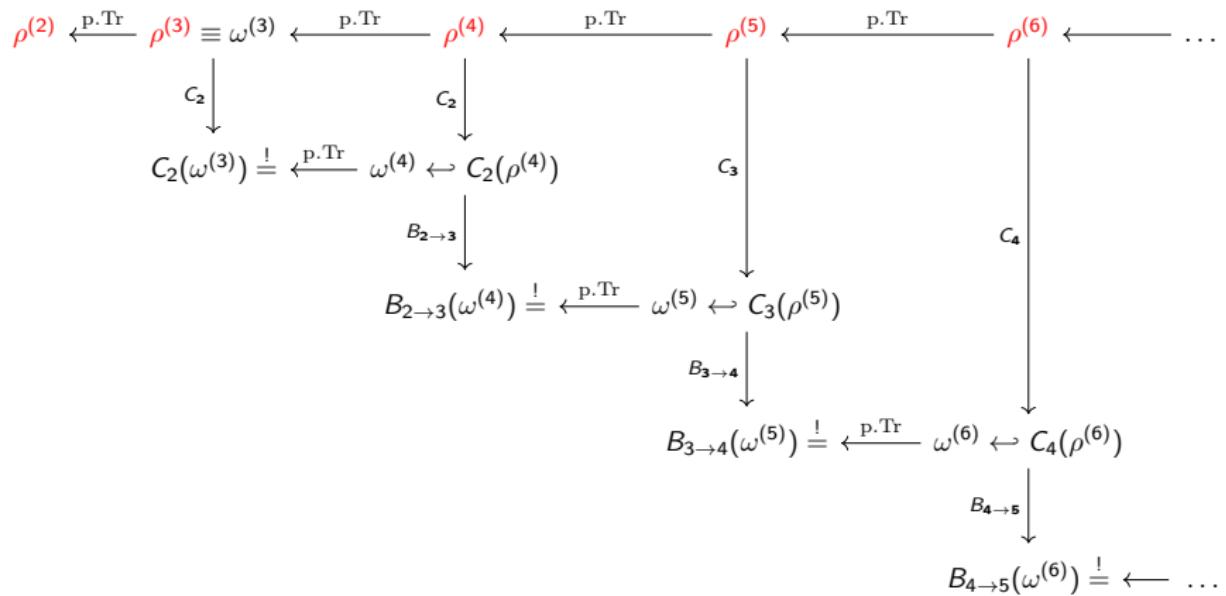


Compress

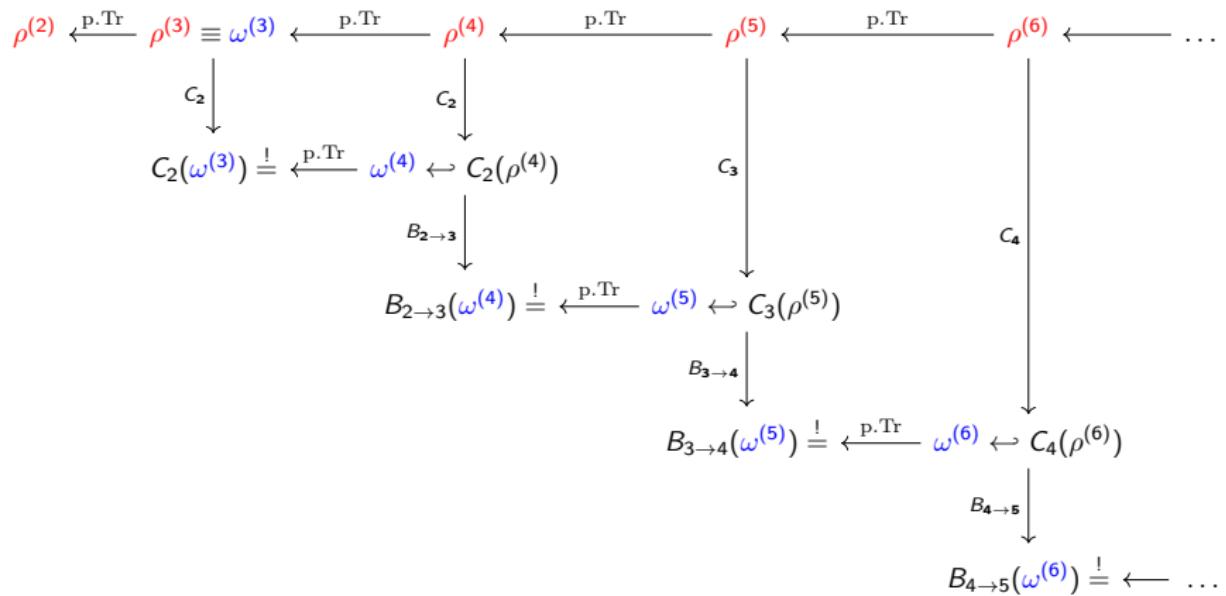
The “triangle of renormalization”



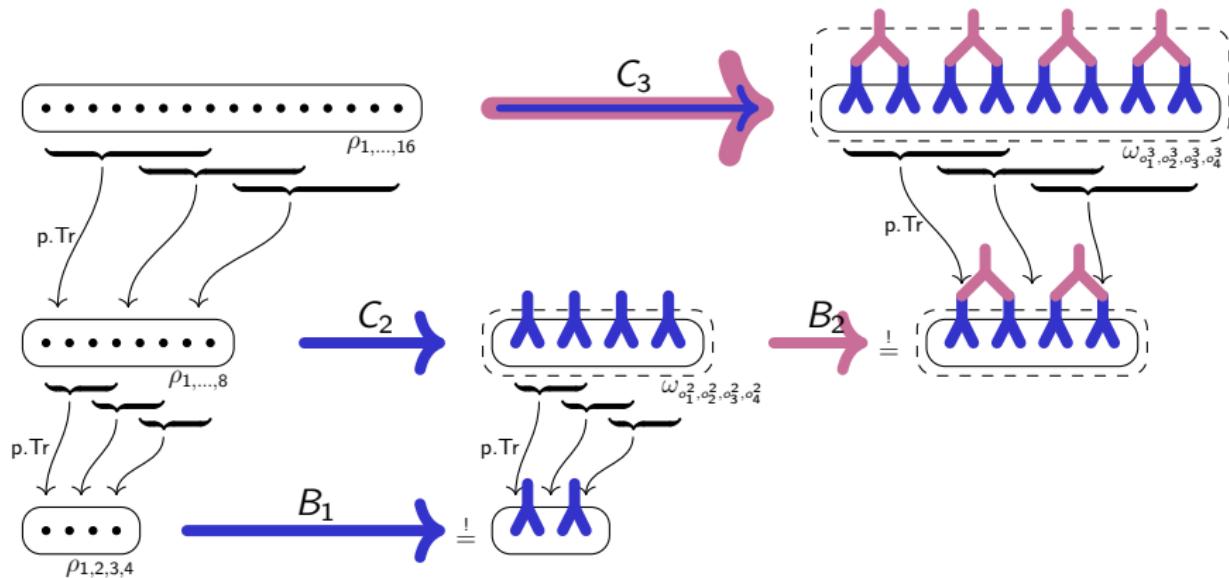
The “triangle of renormalization”



The “triangle of renormalization”



Coarse-graining using tree tensor networks



Benchmarking Results

Translation-invariant spin chains

	Model Name	Hamiltonian	Gap?
(a)	Critical Ising	$H_{\text{TFI}}(1)$	Critical
(b)	$S = 1/2$ Heisenberg	$H_{XXZ}^{1/2}(1)$	Critical
(c)	$S = 1/2$ symmetry broken XXZ	$H_{XXZ}^{1/2}(2)$	Gapped
(d)	$S = 1/2$ XX model	$H_{XXZ}^{1/2}(0)$	Critical
(e)	$S = 1$ Heisenberg	$H_{XXZ}^1(1)$	Gapped
(f)	$S = 1/2$ J_1 - J_2 Heisenberg	$H_{J_1-J_2}(4.15, 1)$	Critical

$$H_{\text{TFI}}(h_z) = - \sum_i X_i X_{i+1} - h_z \sum_i Z_i \quad \text{Transverse field Ising model}$$

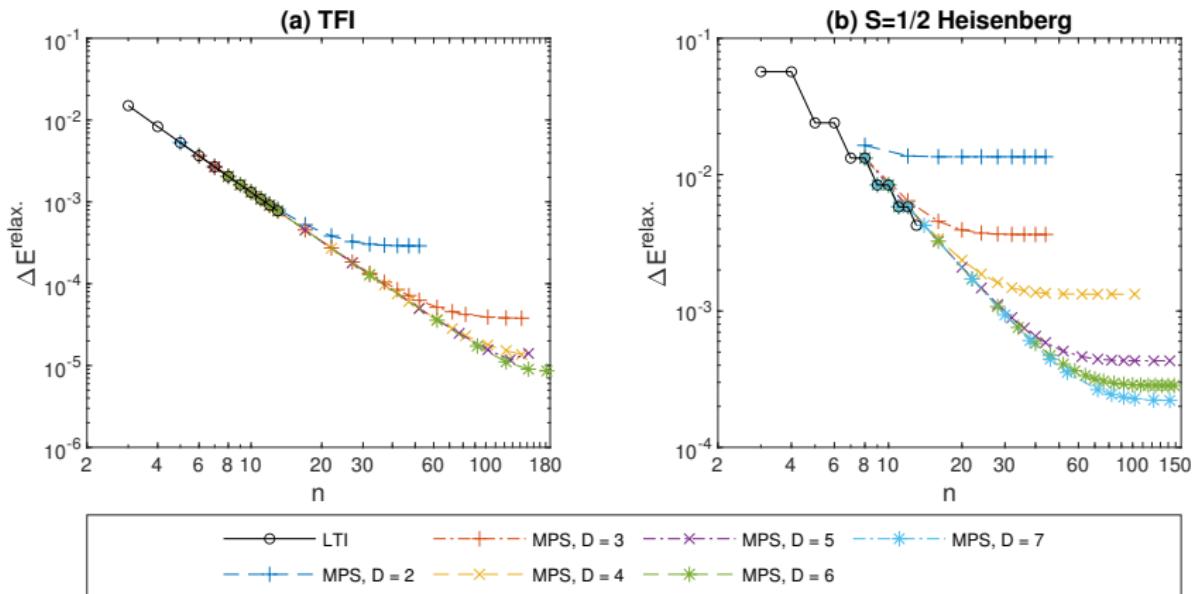
$$H_{XXZ}^S(\Delta) = \sum_i X_i X_{i+1} + Y_i Y_{i+1} + \Delta Z_i Z_{i+1} \quad \text{XXZ spin } S$$

$$H_{J_1-J_2}(J_1, J_2) = \sum_i J_1 \vec{S}_i \cdot \vec{S}_{i+1} + J_2 \vec{S}_i \cdot \vec{S}_{i+2} \quad \text{J}_1\text{-J}_2 \text{ Heisenberg spin } 1/2$$

Results: TFI & $S=1/2$ Heisenberg model

(a) Critical TFI: $H_{\text{TFI}}(1)$

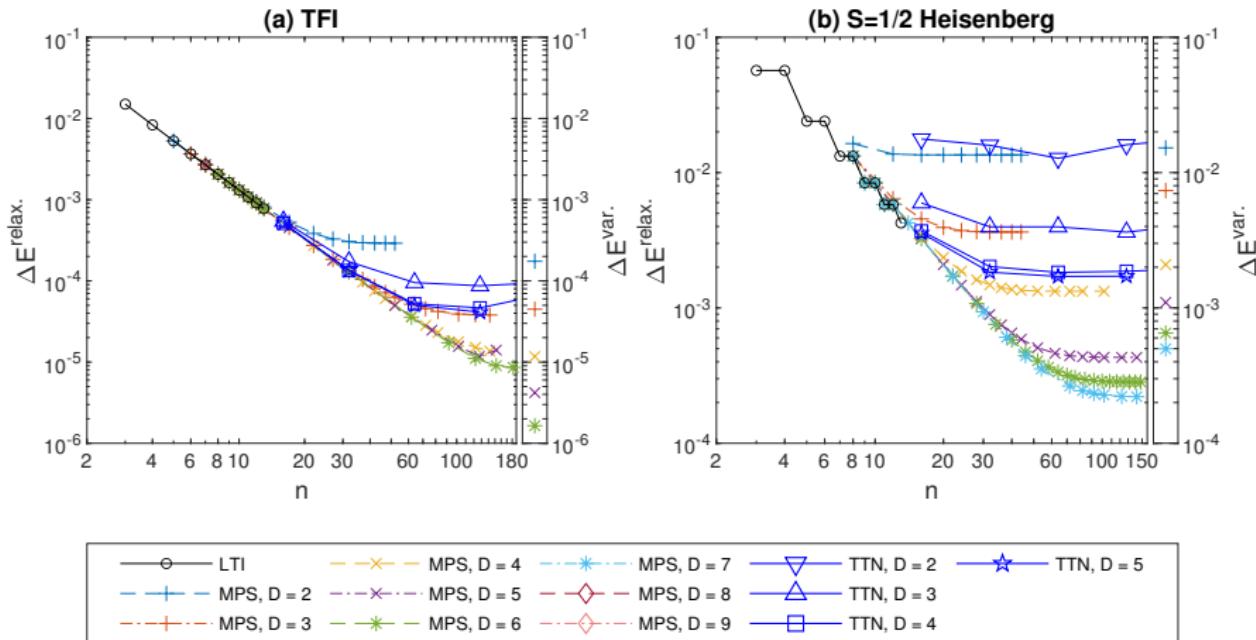
(b) Isotropic antiferromagnetic $S = 1/2$ Heisenberg: $H_{\text{XXZ}}^{1/2}(1)$



Results: TFI & $S=1/2$ Heisenberg

(a) Critical TFI: $H_{\text{TFI}}(1)$

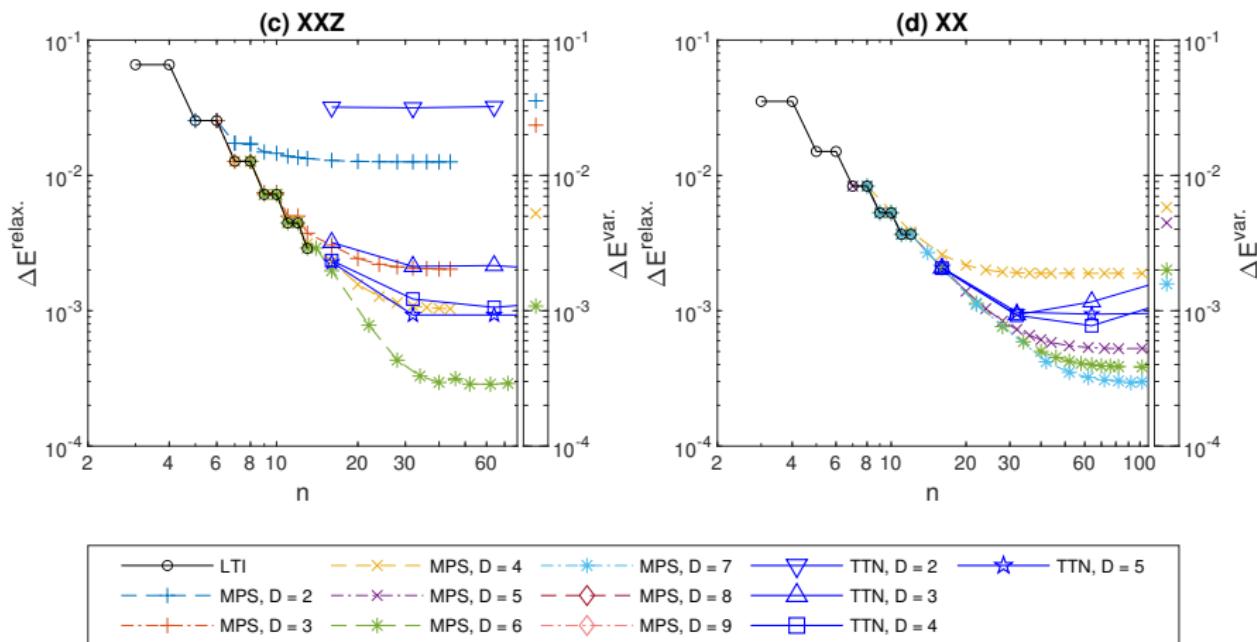
(b) Isotropic antiferromagnetic $S = 1/2$ Heisenberg: $H_{\text{XXZ}}^{1/2}(1)$



Results: XXZ & XX

(c) $S = 1/2$ symmetry broken XXZ: $H_{\text{XXZ}}^{1/2}(2)$

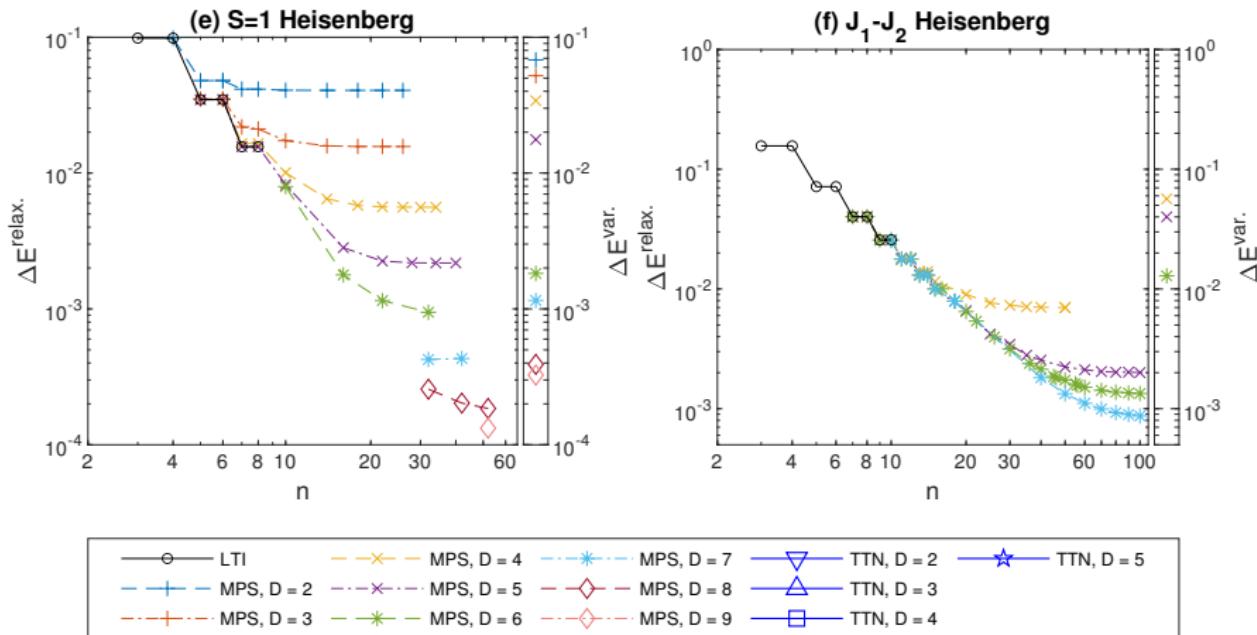
(d) $S = 1/2$ XX model: $H_{\text{XXZ}}^{1/2}(0)$



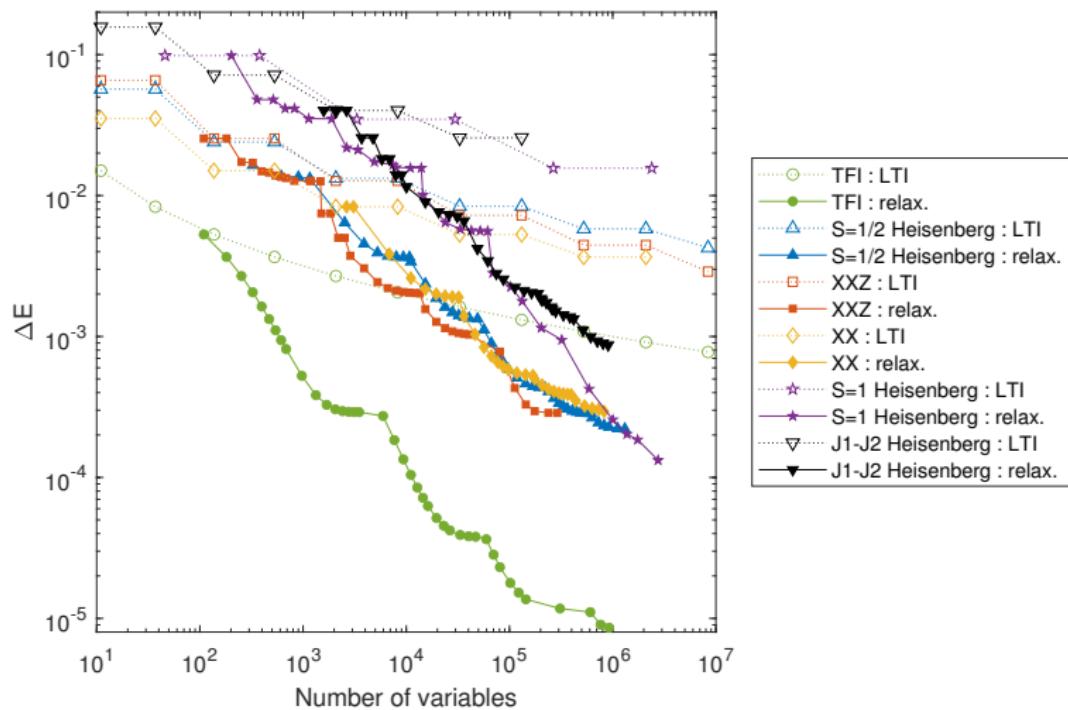
Results: S=1 Heisenberg & J₁-J₂

(e) Isotropic antiferromagnetic S = 1 Heisenberg: $H_{XXZ}^1(1)$

(f) Critical S = 1/2 J₁-J₂ Heisenberg: $H_{J_1-J_2}(4.15, 1)$

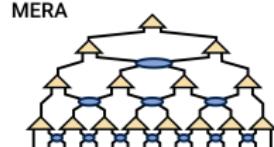
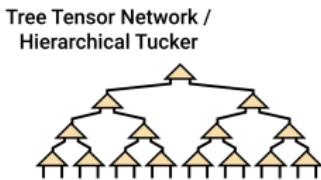
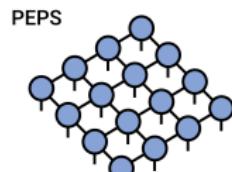
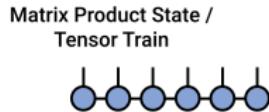


Results: Memory scaling vs. precision

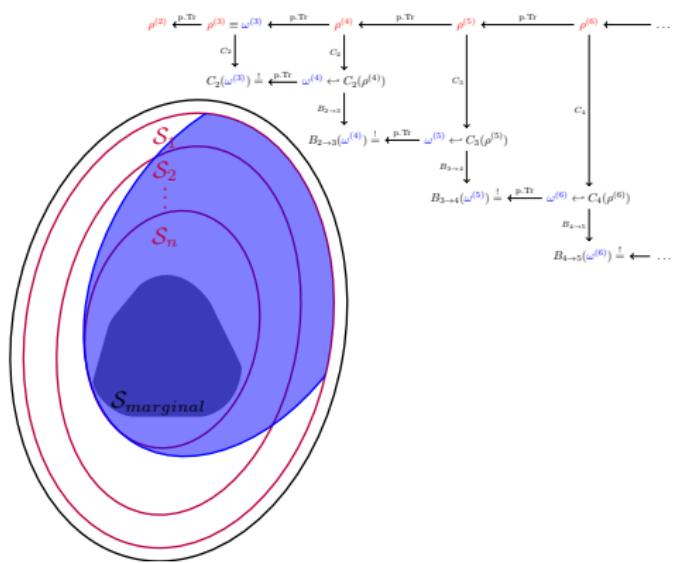


Conclusion

$$\min_{\psi \in \mathcal{C} \subset \mathcal{H}} \langle \psi | H | \psi \rangle \geq \min_{\psi \in \mathcal{H}} \langle \psi | H | \psi \rangle \geq E_{\text{relax.}}(\{C_i\})$$



[tensornetwork.org]
[Cirac and Verstraete, J.Phys.A (2009)]



Results: Relaxation precision vs. VUMPS precision

