

Semi-definite programming for solving the local Hamiltonian problem

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The power of semidefinite programmes (SDPs) for solving quantum ground state problems has been hardly explored so far. This is a veritable deficiency given that they have great potential to outperform and complement existing methods. SDPs, in contrast to any other known method, give lower bounds on the ground state energy, rather than upper bounds. Thereby they yield benchmarks for other successful methods such as those based on tensor networks. What is more, SDPs, unlike quantum Monte-Carlo techniques, do not suffer from the sign problem that prevents application in many interesting situations.

Convex optimization

- A general **convex optimization problem** is of the form:
With f, g_i convex and h_i affine

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^m} f(\mathbf{x}) \\ \text{such that } \forall i : g_i(\mathbf{x}) \geq 0 \\ \forall i : h_i(\mathbf{x}) = 0 \end{aligned}$$

- A simple **example**:

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^2} x_1 x_2 \\ \text{such that } -x_2^2 + x_2 + 1/2 \geq 0 \\ x_1^2 - x_1 = 0 \end{aligned}$$

Semi-definite relaxation

We can **relax** any **convex problem** to a **semi-definite** one. The latter is **easier to solve** and yields an approximate solution of the former. There is a **hierarchy** (Lasserre hierarchy) of relaxations that yield better and better approximations [1].

- A general **semi-definite optimization problem** is of the form:
With a cost vector $\mathbf{c} \in \mathbb{R}^m$ and constrain matrices $(F_i)_{i=0}^m$

$$\begin{aligned} \min_{\mathbf{y} \in \mathbb{R}^m} \mathbf{c} \cdot \mathbf{y} \\ \text{such that } \sum_{i=1}^m F_i y_i - F_0 \geq 0 \end{aligned}$$

- Level 2 relaxation** of the example:

- Construct all order 2 monomials: $\mathbf{x} := (1, x_1, x_2, x_1 x_2, x_1^2, x_2^2)$
- Introduce SDP variables y_1, y_2, \dots, y_8 for all distinct elements in $\mathbf{x} \mathbf{x}^T$ after applying the substitution rule $x_i^2 \mapsto x_i$.
- Rewrite everything in terms of the SDP variables:

$$\begin{aligned} \min y_3 \\ \text{such that } \begin{pmatrix} 1 & y_1 & y_2 & y_3 & y_4 \\ y_1 & y_1 & y_3 & y_3 & y_5 \\ y_2 & y_3 & y_4 & y_5 & y_6 \\ y_3 & y_3 & y_5 & y_5 & y_7 \\ y_4 & y_5 & y_6 & y_7 & y_8 \end{pmatrix} \geq 0 \\ \begin{pmatrix} -y_4 + y_2 + 1/2 & -y_5 + y_3 + y_1/2 & -y_6 + y_4 + y_2/2 \\ -y_5 + y_3 + y_1/2 & -y_5 + y_3 + y_1/2 & -y_7 + y_5 + y_3/2 \\ -y_6 + y_4 + y_2/2 & -y_7 + y_5 + y_3/2 & -y_8 + y_6 + y_4/2 \end{pmatrix} \geq 0 \end{aligned}$$

Non-commutative generalizations

In a **non-commutative** convex optimization problem the vector \mathbf{x} of numbers is replaced by a vector of **operators**.

- A similar hierarchy of semi-definite relaxations still exists [2]
- SDP variable become **expectation values of local observables**
- The hierarchy can be **proven to converge** if the constraints force the operators to be **bounded** (satisfied for fermions and spins) [3]

Benchmarking with the Fermi-Hubbard model

The Fermi-Hubbard model is the **paradigmatic model** of condensed matter physics. It describes electrons hopping on a lattice.

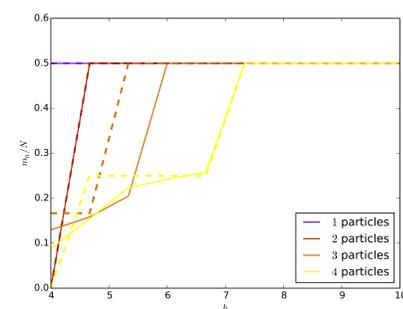
$$H := \underbrace{-t \sum_{\langle j,k \rangle, \sigma \in \{\uparrow, \downarrow\}} f_{j,\sigma}^\dagger f_{k,\sigma}}_{\text{hopping}} + \underbrace{U \sum_j f_{j,\uparrow}^\dagger f_{j,\downarrow}^\dagger f_{j,\downarrow} f_{j,\uparrow}}_{\text{on-site interaction}} - \underbrace{\frac{h}{2} \sum_j f_{j,\uparrow}^\dagger f_{j,\uparrow} - f_{j,\downarrow}^\dagger f_{j,\downarrow}}_{\text{magnetic field}}$$

- Exactly solvable in 1D by Bethe ansatz
- Difficult to solve in 2D
 - No analytic solution
 - Hilbert space grows fast \Rightarrow exact diagonalization only for $\lesssim 12$ sites
 - Fermions \Rightarrow sign problem \Rightarrow no Monte Carlo
 - 2D \Rightarrow tensor networks not efficiently contactable

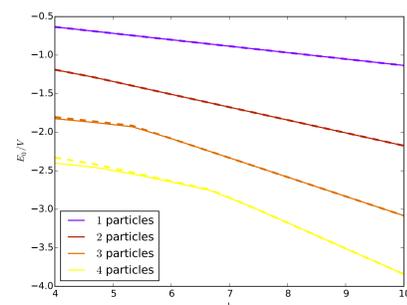
Work in progress: very first preliminary numerical results

We are implementing a pair of **python libraries** [4, 5] that make formulating and solving semi-definite relaxations of ground-state problems of local Hamiltonians **easy**.

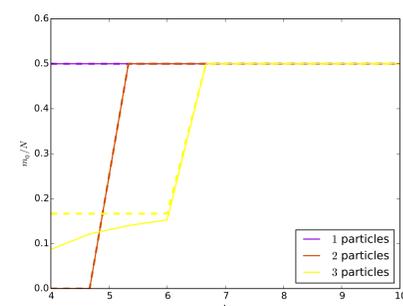
- Magnetization $m_0 := \langle \sum_j a_{\uparrow,j}^\dagger a_{\uparrow,j} - a_{\downarrow,j}^\dagger a_{\downarrow,j} \rangle_0$



- Energy $E_0 := \langle H \rangle_0$



- Energy $E_0 := \langle H \rangle_0$ in 2D



[1] J. B. Lasserre, "Global Optimization with Polynomials and the Problem of Moments", *SIAM Journal on Optimization*, 11.3 (2001), 796–817.

[2] M. Navascués, S. Pironio, and A. Acín, "Bounding the Set of Quantum Correlations", *Physical Review Letters*, 98.1 (2007), 10401.

[3] S. Pironio, M. Navascués, and A. Acín, "Convergent Relaxations of Polynomial Optimization Problems with Noncommuting Variables", *SIAM Journal on Optimization*, 20.5 (2010), 2157–2180.

[4] ncpol2sdpa, URL: <https://github.com/peterwittek/ncpol2sdpa>.

[5] sdpqpy, URL: <https://github.com/cgogolin/sdpqpy>.