#### Scalable variational embedding for quantum many-body problems

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Electronic structure methods: quantum magnetism and numerical approaches beyond density functional Theory

**Michael Lindsey** Collaborators: Yuehaw Khoo, Lin Lin

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# The ground-state eigenvalue problem

- The basic quantum many-body problem is the *ground state eigenvalue problem* 
  - Find lowest eigenvalue  $E_0$  of an operator  $\hat{H}$  on a Hilbert space Q of exponentially high dimension
  - Variational formulation

$$E_0 = \min_{\phi \in \mathcal{Q} : \phi^* \phi = 1} \phi^* \hat{H} \phi$$

- Wide-reaching applications in chemistry, physics, and materials science
  - Include equilibrium geometry of molecules, *ab initio* molecular dynamics
  - Moreover, expectations of the form  $\phi^* \hat{O} \phi$  predict physically observable quantities within the ground state

#### Quantum spin systems

- Consider a model consisting of M sites, indexed by  $i \in [M] := \{1, \dots, M\}$
- Classically, each site can assume a binary state  $s_i \in \{0,1\}$
- The classical states are then binary strings

$$\mathbf{s} = (s_1, \ldots, s_M) \in \{0, 1\}^M$$

• We will consider the 'quantum analog' of classical combinatorial (0-1) optimization problems

$$\min_{\mathbf{s}\in\{0,1\}^M}f(\mathbf{s})$$

- e.g., MaxCut:  $f(\mathbf{s}) = \mathbf{s}^{\top} A \mathbf{s}$ , where A is an adjacency matrix for a graph on [M]
- .Quantum wavefunctions are complex functions  $\phi : \{0,1\}^M \to \mathbb{C}$ 
  - Can be viewed as tensors in  $\mathcal{Q}:=\mathbb{C}^2\otimes\cdots\otimes\mathbb{C}^2$  (*M* times) via

$$\phi_{s_1 s_2 \cdots s_M} = \phi(\mathbf{s})$$

- This is the Hilbert space for quantum spin- $\frac{1}{2}$  systems
  - Also the Hilbert space for electronic structure problems after passing to the second-quantized fermionic formalism

- $\bullet\,$  Let  ${\mathcal A}$  be the algebra of operators on  ${\mathcal Q}$ 
  - $\bullet$  Classical analog is functions on  $\{0,1\}^{M}$

• For any subset  $S \subset [M]$ , we have a subalgebra  $\mathcal{A}_S$  of *local* operators

- Classical analog is functions that depend only on a subset of variables
- For quantum spin systems,  $A_S$  consists of operators obtained by tensoring with the identity operator on sites in  $[M] \setminus S$
- For fermions,  $\mathcal{A}_S$  is generated by the creation/annihilation operators  $a_i^\dagger, a_i$  for  $i \in S$

# Algebras of operators



• Then given a partition of [M] into disjoint clusters  $C_{\gamma}$ , assume our Hamiltonian operator can be written

$$\hat{H} = \sum_{\gamma} \hat{H}_{\gamma} + \sum_{\gamma\delta} \hat{H}_{\gamma\delta},$$

where  $\hat{H}_{\gamma} \in \mathcal{A}_{\gamma} := \mathcal{A}_{\mathcal{C}_{\gamma}}$  and  $\hat{H}_{\gamma\delta} \in \mathcal{A}_{\gamma\delta} := \mathcal{A}_{\mathcal{C}_{\gamma} \cup \mathcal{C}_{\delta}}$  are Hermitian

• True of many physical problems, including suitable discretizations of electronic structure problems

## State formulation

- Call a linear functional  $\omega : \mathcal{A} \to \mathbb{C}$  a state on  $\mathcal{A}$  if  $\omega(\hat{A}^*) = \omega(\hat{A})^*$ ,  $\omega(\hat{A}^*\hat{A}) \ge 0$  for all  $\hat{A} \in \mathcal{A}$ , and  $\omega(\mathrm{Id}) = 1$ 
  - $\omega(\hat{A}) = \operatorname{Tr} \left[ \hat{A} \rho \right]$  for some density operator  $\rho$  on  $\mathcal{Q}$  ( $\rho \succeq 0$ ,  $\operatorname{Tr}[\rho] = 1$ )
  - Let  $\Omega$ ,  $\Omega_{\gamma}$ ,  $\Omega_{\gamma\delta}$  be the (convex) sets of states on  $\mathcal{A}$ ,  $\mathcal{A}_{\gamma}$ , and  $\mathcal{A}_{\gamma\delta}$
- Ground state eigenvalue problem can be rephrased as

$$E_0 = \min_{\omega \in \Omega} \omega(\hat{H})$$

• The optimizer  $\omega$  is given by  $\omega(\hat{A}) = \text{Tr} \left[\hat{A}\phi\phi^*\right] = \phi^*\hat{A}\phi$ , where  $\phi$  is the ground-state eigenvector

• Note that we can rewrite

$$E_{\mathbf{0}} = \inf \left\{ \sum_{\gamma} \omega_{\gamma}(\hat{H}_{\gamma}) + \sum_{\gamma \delta} \omega_{\gamma \delta}(\hat{H}_{\gamma \delta}) \, : \, \omega_{\gamma} \in \Omega_{\gamma}, \, \, \omega_{\gamma \delta} \in \Omega_{\gamma \delta} \, \, \text{are jointly representable} 
ight\}$$

- Joint representability means that the  $\omega_{\gamma} \in \Omega_{\gamma}$ ,  $\omega_{\gamma\delta} \in \Omega_{\gamma\delta}$  all could have come from the same  $\omega \in \Omega$  by restriction
- We have changed exponential size of optimization space for exponential complexity of constraints

# Semidefinite relaxation

- We aim to relax the joint representability constraint to get a lower bound
- That is, enforce some necessary (but not sufficient) constraints for joint representability:
  - **3** State:  $\omega_{\gamma\delta} \in \Omega_{\gamma\delta}$ 
    - Yields independent semidefinite constraints for each pair  $(\gamma, \delta)$
  - **3** Local consistency:  $\omega_{\gamma\delta}(\hat{A}) = \omega_{\gamma}(\hat{A})$  for  $\hat{A} \in \mathcal{A}_{\gamma}$ ,  $\omega_{\gamma\delta}(\hat{A}) = \omega_{\delta}(\hat{A})$  for  $\hat{A} \in \mathcal{A}_{\delta}$ 
    - Yields linear equality constraints coupling overlapping pairs of clusters
  - $\textbf{O} \quad \textbf{Global consistency:} \ \omega \left[ \left( \sum_{\gamma} \hat{A}_{\gamma} \right)^{\dagger} \left( \sum_{\gamma} \hat{A}_{\gamma} \right) \right] \geq 0 \text{ for any } \hat{A}_{\gamma} \in \mathcal{A}_{\gamma}$

 $\bullet\,$  Yields global semidefinite constraint coupling all pairs  $(\gamma,\delta)$ 

## Semidefinite relaxation

• Concretely one obtains

$$\begin{array}{ll} \underset{\{\rho_i\}, \{\rho_{ij}\}_{i < j}}{\text{minimize}} & \sum_{i} \operatorname{Tr} \left[H_i \rho_i\right] + \sum_{i < j} \operatorname{Tr} \left[H_{ij} \rho_{ij}\right] \\ \text{subject to} & \rho_{ij} \succeq 0, \quad 1 \le i < j \le M, \\ & \rho_i = A_1[\rho_{ij}], \ \rho_j = A_2[\rho_{ij}], \quad 1 \le i < j \le M, \\ & \operatorname{Tr}[\rho_i] = 1, \quad i = 1, \dots, M, \\ & G[\{\rho_i\}, \{\rho_{ij}\}_{i \le j}] \succeq 0. \end{array}$$

for suitable matrices  $H_i$ ,  $H_{ij}$ 

- For simplicity now use *i*, *j* to index clusters, not sites
- Call it the 2-marginal relaxation, optimal value  $E_0^{(2)}$ 
  - Analogy of local states to marginals in classical probability

## Partial duality

• Dualize only the global semidefinite constraint to obtain

$$\mathsf{E}_0^{(2)} = \sup_{X \succeq 0} \ \mathcal{F}[X],$$

where  $\mathcal{F}[X]$  is optimal value of 'effective problem'

$$\begin{array}{ll} \underset{\{\rho_i\}, \{\rho_{ij}\}_{i < j}}{\text{minimize}} & \sum_{i} \operatorname{Tr} \left[ H_i(X)\rho_i \right] + \sum_{i < j} \operatorname{Tr} \left[ H_{ij}(X)\rho_{ij} \right] \\ \text{subject to} & \rho_{ij} \succeq 0, \quad 1 \le i < j \le M, \\ & \rho_i = A_1[\rho_{ij}], \ \rho_j = A_2[\rho_{ij}], \quad 1 \le i < j \le M, \\ & \operatorname{Tr}[\rho_i] = 1, \quad i = 1, \dots, M \end{array}$$

• Effective problem has similar structure to original SDP

- $\bullet\,$  But global semidefinite constraints omitted, exchanged for effective contribution dependent on X
- Gives the interpretation of a *quantum embedding theory*

## Partial dual gradient ascent approach

- Want to perform projected gradient ascent on  $\mathcal{F}[X]$  over  $X \succeq 0$
- Alternate between:
  - Obtain {\(\rho\_i\)}\), {\(\rho\_{ij}\)}\_{i < j}\) by solving the effective problem (holding X fixed)</li>

Supdate  $X \leftarrow \prod_{\geq 0} (X - \varepsilon G[\{\rho_i\}, \{\rho_{ij}\}_{i < j}])$ 

- In practice, we replace step (1) with a single iteration of an augmented Lagrangian-type solver
- Translation-invariance can be exploited for a per-iteration cost scaling *linearly* in number K of clusters
  - Bottleneck: K full matrix diagonalizations
    - These are decoupled and can be perfectly parallelized
- Otherwise the global semidefinite constraint is generally cubic in K
- Scaling is exponential in cluster size  $L = |C_{\gamma}|$ 
  - In our experiments,  $L \leq 4$
  - Ongoing: further (local) relaxation may enable larger clusters for *ab initio* quantum chemistry

• I'll present results for the transverse-field Ising (TFI) and anti-ferromagnetic Heisenberg (AFH) model

$$\begin{split} \mathcal{H}_{\mathrm{TFI}} &= -h \sum_{i} \sigma_{i}^{\mathrm{x}} - \sum_{\langle i,j \rangle} \sigma_{i}^{z} \sigma_{j}^{z} \\ \mathcal{H}_{\mathrm{AFH}} &= \sum_{\langle i,j \rangle} \left[ \sigma_{i}^{\mathrm{x}} \sigma_{j}^{\mathrm{x}} + \sigma_{i}^{\mathrm{y}} \sigma_{j}^{\mathrm{y}} + \sigma_{i}^{z} \sigma_{j}^{z} \right] \end{split}$$

#### TFI exact benchmark



$1 \times 1$ clusters	2  imes 1 clusters	4  imes 1 clusters
0.5383	0.0521	0.0034

Table: Relaxation error per site for the AFH model on a  $20 \times 1$  periodic lattice for various cluster sizes.

$1 \times 1$ clusters	$2 \times 1$ clusters	$2 \times 2$ clusters
0.6634	0.1851	0.0034

Table: Relaxation error per site for the AFH model on a  $4 \times 4$  periodic lattice for various cluster sizes.

	1  imes 1 clusters	$2 \times 1$ clusters	$1 \times 3$ clusters
With global constraints	1.0439	0.3937	0.0410
W/o global constraints	3.5439	2.1897	0.8773

Table: Relaxation error per site for the AFH model on a  $4\times 3$  periodic lattice for various cluster sizes

#### Dependence of convergence on system size



## Dependence of convergence on cluster size



## References and related work

- References for the talk
  - L. Lin and M.L., Variational embedding for quantum many-body problems, *CPAM* (2021)
  - Y. Khoo and M.L., Scalable semidefinite programming approach to variational embedding for quantum many-body problems (*in preparation*)
- Classical 0-1 optimization, graphical models
  - [Goemans and Williamson (1995)] (MaxCut)
  - [Jordan and Wainwright (2008)] (Variational inference, marginal polytope relaxations)
- 2-RDM theory (electronic structure)
  - e.g., many papers of D. Mazziotti et al and M. Nakata et al
- Other quantum marginal-type relaxations
  - [Ferris and Poulin (2013)] (quantum belief propagation)
  - [Barthel and Hübener (2012)]
- Classical analogs of the relaxation in this talk
  - [Khoo, Lin, M.L., Ying (2020)] (multi-marginal optimal transport)
  - [Chen, Khoo, M.L. (2020)] (marginal relaxation for global optimization)