Locality optimization for parent Hamiltonians of tensor networks

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Tensor network states form a powerful framework for both the analytical and numerical study of strongly correlated phases. Vital to their analytical utility is that they appear as the exact ground states of associated parent Hamiltonians, where canonical proof techniques guarantee a controlled ground space structure. Yet, while those Hamiltonians are local by construction, the known techniques often yield complex Hamiltonians which act on a rather large number of spins. In this paper, we present an algorithm to systematically simplify parent Hamiltonians, breaking them down into any given basis of elementary interaction terms. The underlying optimization problem is a semidefinite program, and thus the optimal solution can be found efficiently. Our method exploits a degree of freedom in the construction of parent Hamiltonians-the excitation spectrum of the local terms-over which it optimizes such as to obtain the best possible approximation. We benchmark our method on the AKLT model and the toric code model, where we show that the canonical parent Hamiltonians (acting on 3 or 4 and 12 sites, respectively) can be broken down to the known optimal two-body and four-body terms. We then apply our method to the paradigmatic resonating valence bond (RVB) model on the kagome lattice. Here, the simplest previously known parent Hamiltonian acts on all the 12 spins on one kagome star. With our optimization algorithm, we obtain a vastly simpler Hamiltonian: we find that the RVB model is the ground state of a parent Hamiltonian whose terms are all products of at most four Heisenberg interactions, and whose range can be further constrained, providing a major improvement over the previously known 12-body Hamiltonian.

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I. INTRODUCTION

Tensor network states (TNS), in particular onedimensional matrix product states (MPS) and higherdimensional projected entangled pair states (PEPS), form a powerful framework for the study of strongly correlated quantum many-body systems. They describe complex many-body wave functions by associating local tensors to individual sites which are then correlated locally, based upon the understanding that the entanglement follows the locality of the interactions. MPS and PEPS therefore provide a faithful approximation of low-energy states of local Hamiltonians, which—together with efficient algorithms for their variational optimization—makes these states the basis of a wide variety of numerical methods [1–7].

At the same time, MPS and PEPS also provide an effective toolkit for the *analytical* study of quantum many-body systems [8], for several reasons: first, tensor networks allow to model global symmetries locally, enabling one to understand their action on the entanglement [9–11]; second, many interesting wave functions have an exact MPS or PEPS representation (such as topological fixed point models or Anderson's resonating valence Bond wave function) [12–15]; and third, any MPS or PEPS appears as the exact ground state of some associated local *parent Hamiltonian* [11,16–21]. Importantly, these Hamiltonians inherit the symmetries encoded in the tensor, and techniques for lower bounding their gaps have been established, which makes them perfectly suited for the analytical characterization of the physics of strongly correlated phases. The most well-known among such models is arguably the AKLT (Affleck-Kennedy-Lieb-Tasaki) model [22,23], which provided the first example of a spin-1 model with SO(3) symmetry for which the Haldane gap could be rigorously proven, and which historically gave rise to the development of MPS as an analytical tool [8].

In constructing parent Hamiltonians, two main desiderata must be met: They should have a well-behaved ground space, and they should be simple. By construction, they enforce local consistency of the ground space with the tensor network description, and a variety of conditions has been derived which guarantee that this local consistency implies global consistency, that is, a global ground space which is either unique or has a controlled degeneracy (such as for topological phases) [11,17–21]. At the same time, the construction principle of enforcing local consistency automatically yields a Hamiltonian which is local, i.e., a sum of local terms. However, though local, these Hamiltonians can still act on fairly large clusters, in particular in 2D. This is due to the fact that generically,

nontrivial constraints only arise once the degrees of freedom in the bulk exceed those at the boundary, which requires larger regions in higher dimensions.

Remarkably, however, in many cases of practical interest much smaller Hamiltonians than those derived through canonical (i.e., generally applicable) proof techniques suffice. For instance, in the 1D AKLT model, the canonical three-body Hamiltonian can be broken down into two-body Hamiltonians, as well as in the 2D honeycomb AKLT model (ten-body to two-body) or in the toric code model (12-body to four-body). Clearly, obtaining such significantly simplified Hamiltonians is highly desirable. Unfortunately, no systematic procedure for breaking down canonically constructed parent Hamiltonians is known: to start with, this requires a suitable guess for a simpler Hamiltonian, which is not always available, and said guess must subsequently be shown to have the same ground space as the canonical Hamiltonian, which must either be done brute-force on a computer, or by proving it by hand on a case-by-case basis.

In this paper, we present a systematic method to break down any given parent Hamiltonian into sums of simple elementary Hamiltonian terms. The method can be combined with any canonical tensor network parent Hamiltonian construction from the literature, and for any given target set of desired simple Hamiltonian terms. Our method gives rise to an optimization problem which can be solved systematically and efficiently; formally, the optimization can be rephrased as a semidefinite program (SDP) and is thus converging provably efficiently. A central ingredient which we exploit in this optimization is an ambiguity in the canonical construction of parent Hamiltonians, relating to the spectrum of local excitations.

To test our method, we first benchmark it on two wellstudied models. First, the 1D AKLT model, where we show that it allows to reduce the four-body parent Hamiltonian obtained through the most fundamental canonical technique to a sum of nearest-neighbor two-body interactions. Second, the toric code model, where more refined canonical techniques allow to construct a 12-body Hamiltonian, which our method breaks down to the well-known sum of four-body vertex and plaquette terms.

We then apply our algorithm to one of the most paradigmatic models of a topological spin liquid, the spin-1/2 resonating valence bond (RVB) state on the kagome lattice. There, our method allows us to obtain a vastly simpler parent Hamiltonian than the ones previously known. Namely, for this model, refined canonical techniques gave rise to a 19-body Hamiltonian (acting on two overlapping stars) [13], which had subsequently be shown to be a sum of two 12-body terms (each acting on one star) by a suitable guess, confirmed by brute-force numerical analysis [24]. Applying our method to systematically study the possibility to further decompose this 12-body interaction, we find that it can be decomposed into terms containing product of at most four Heisenberg interactions each, thereby significantly reducing the complexity of the parent Hamiltonian of the kagome RVB model.

II. TENSOR NETWORKS AND PARENT HAMILTONIANS

We start by introducing tensor network states and parent Hamiltonians, and in particular canonical constructions for the



FIG. 1. Tensor networks (shown for a 2D square lattice) are constucted from a local tensors $A^i_{\alpha\beta\gamma\delta}$ (with physical index *i* and virtual indices α, β, \ldots); they are arranged on a lattice and the virtual indices are contracted to yield the expansion coefficient $|\Psi\rangle = \sum c_{i_1,\ldots,i_N} |i_1,\ldots,i_N\rangle$.

latter. Tensor networks [1,2,8] are constructed by associating a tensor $A^i_{\alpha\beta\cdots}$ to each site (we focus on translational invariant systems, where all tensors are the same), where i = 1, ..., dis the *physical index* and $\alpha, \beta, \ldots = 1, \ldots, D$ are *virtual in*dices or entanglement indices with bond dimension D. The tensors are then arranged on a lattice and the virtual indices contracted (that is, identified and summed over) with those of the adjacent tensors. Two particularly important classes are MPS, where the tensors are arranged on a 1D line, and PEPS, where the tensors are placed on some 2D lattice. After contracting all indices with suitable boundary conditions (for the purpose of this work, the boundary conditions are irrelevant, but take periodic), one is left with a multi-index tensor $c_{i_1,...,i_N}$ which depends on the physical index i of all N spins in the system, and which provides an MPS or PEPS description of its wave function $|\Psi\rangle$ by virtue of $|\Psi\rangle = \sum c_{i_1,\dots,i_N} |i_1,\dots,i_N\rangle$ [Fig. 1]. The construction can be modified to include more than one type of tensors, such as tensors which only carry virtual indices and which are placed in between the tensors with physical indices; we will encounter such an instance later in Sec. IV B.

We now turn to parent Hamiltonians. To this end, consider blocking the tensors in a contiguous region R, and consider the resulting object as a linear map $\mathcal{P} : (\mathbb{C}^D)^{\otimes |\partial R|} \to (\mathbb{C}^d)^{\otimes |R|}$ from the virtual indices at the boundary ∂R of the region to the physical indices in the bulk R [Fig. 2(a)]. Since the volume |R| grows faster than the boundary $|\partial R|$ as the size of R is (uniformly) increased, one quickly reaches a point where Im \mathcal{P} is not full rank [Fig. 2(b)]. Then, we can define a parent Hamiltonian $h \ge 0$ such that $h|_{\text{Im }\mathcal{P}} = 0$ and $h|_{\mathcal{K}} > 0$ on $\mathcal{K} = (\text{Im }\mathcal{P})^{\perp}$ [Fig. 2(c)]; canonically, one chooses h to be the projector onto \mathcal{K} . This Hamiltonian satisfies $h|\Psi\rangle = 0$. If we now take such an h_i centered around every site i in the lattice, then $H = \sum h_i \ge 0$, and $H|\Psi\rangle = \sum h_i|\Psi\rangle = 0$, that is, the MPS or PEPS $|\Psi\rangle$ is a ground state of the *parent Hamiltonian* H.

A key question is when this construction gives Hamiltonians with a well-behaved ground space. For this, the concept of *injectivity* and the associated *injectivity length* is of central importance [8]. A tensor network is said to be *normal* if under blocking sufficiently many sites, \mathcal{P} becomes an injective map. The blocked tensor network is then said to be injective, and the size L of the block (in 1D) is the associated injectivity length. In 2D, the characterization of the smallest injective



FIG. 2. Schematics of the parent Hamiltonian construction. (a) The elementary tensors are blocked on a region *R*. The resulting tensor is interpreted as a linear map \mathcal{P} from the virtual to the physical system. (b) If the blocked region is sufficiently large, the image of the map \mathcal{P} , and thus the support of the overall tensor network wave function, will be only a subspace of the full physical Hilbert space \mathcal{H} , leaving an orthogonal complement $(\operatorname{Im} \mathcal{P})^{\perp}$ orthogonal to the state. (c) A parent Hamiltonian for the state is obtained by taking an arbitrary positive semi-definite operator *h* which is zero of $\operatorname{Im} \mathcal{P}$ and strictly positive on $(\operatorname{Im} \mathcal{P})^{\perp}$.

block will depend on the lattice and might not be unique, for the following, we consider a square region of size $H \times V$ which is injective. The relevance of injectivity lies in the fact that an injective \mathcal{P} can be inverted on Im \mathcal{P} by acting on the physical system. Thus an injective tensor network is equivalent up to local (nonunitary) transformations to maximally entangled states between nearest neighbors. The latter is the unique ground state of a two-body Hamiltonian, and by conjugating this Hamiltonian with the inverse of \mathcal{P} , one arrives at the conclusion that the two-body parent Hamiltonian constructed from the blocked tensor network has a unique ground state [8,25,26]. Thus the parent Hamiltonian acting on 2L sites (in 1D) or $2H \times V$ and $H \times 2V$ sites (in 2D) has a unique ground state. This result can be improved by using more refined proof techniques, which allow to show that the parent Hamiltonian constructed on L + 1 (in 1D) or $(H+1) \times V$ and $H \times (V+1)$ sites (in 2D) has a unique ground state [8,11,17].

Similar results have been derived for 2D systems exhibiting topological order. There, the individual PEPS tensors must exhibit an entanglement symmetry, characterized by a group action or set of matrix product operators (MPOs), and the relevant scale is set by the block size at which \mathcal{P} becomes injective on the symmetric subspace of the entanglement degrees of freedom (*G* injectivity or MPO injectivity) [19–21]. Again, this implies that the PEPS is equivalent to a topological fixed point model up to local transformations, which can be used to obtain a parent Hamiltonian with a topological ground space (e.g., acting on $2H \times 2V$ sites for Kitaev's double models on the square lattice) [13]; yet again, this can be improved using more refined techniques to smaller regions, such as $(H + 1) \times (V + 1)$ on the square lattice, where the size and shape of the minimal region will depend on the structure of the tensor network at hand (but not on the specific model) [8].

In the construction of parent Hamiltonians, there remains a degree of freedom, beyond the size of the region (the \mathcal{P}) from which the Hamiltonian is constructed. Namely, we require $h|_{\mathcal{K}} > 0$ exactly on $\mathcal{K} = (\operatorname{Im} \mathcal{P})^{\perp}$, but any such h will be a valid choice (and moreover, since any two such operators are relatively bounded, this choice does not affect gappedness of the total Hamiltonian). This degree of freedom can be utilized to find an h which can be broken down into a sum of simpler terms. One example where this happens is the AKLT model, where the sum of two two-body Hamiltonian terms has the same ground space as the canonical three-body Hamiltonian with a suitably chosen $h|_{\mathcal{K}} > 0$. In this specific case, a guess for the two-body Hamiltonian can be obtained from the parent Hamiltonian construction, applied to two sites. However, this need not generally be the case; rather, we generally expect a parent Hamiltonian to be decomposable into a sum of smaller terms which by themselves are not parent Hamiltonians (i.e., positive semi-definite operators which annihilate the state). Thus the question arises whether and how the degree of freedom $h|_{\mathcal{K}} > 0$ can be systematically exploited to break down parent Hamiltonians into a sum of simpler terms. This is precisely the question which we will address in the remainder of the paper.

III. LOCALITY OPTIMIZATION ALGORITHM

As we have just discussed, the parent Hamiltonian, constructed on any given patch, is highly nonunique, as any operator $h \ge 0$ with $h|_{\mathcal{K}} > 0$ on $\mathcal{K} = (\operatorname{Im} \mathcal{P})^{\perp}$, and zero otherwise, will serve that purpose. In the following, we will devise an algorithm which uses this degree of freedom to break down the parent Hamiltonian into a sum of simpler "target" terms.

A. Algorithm

The goal is to expand the target Hamiltonian in some given set of (local) operators O_a ,

$$h = \sum_{a} c_a O_a, \quad O_a = (O_a)^{\dagger}. \tag{1}$$

In order to keep the dimension of $\{O_a\}$ as small as possible, one can e.g. use that the parent Hamiltonian inherits all the symmetries of the TN state, and restrict to a set which shares the same symmetries. The goal of the algorithm is to find a set of coefficients $\{c_a\}$ such that *h* is a parent Hamiltonian, that is, *h* vanishes on Im \mathcal{P} and it is strictly positive on its orthogonal complement. To achieve this, we choose a basis $\{|\varphi_a\rangle\}$ of $(\text{Im }\mathcal{P})^{\perp}$ and expand

$$h = \sum_{\alpha,\beta} P_{\alpha\beta} |\varphi_{\alpha}\rangle \langle \varphi_{\beta} | , \qquad (2)$$

where we impose that $P \equiv (P_{\alpha\beta})$ is a strictly positive matrix; this guarantees that *h* has the required property. Finding a decomposition $\{c_a\}$ of the parent Hamiltonian now amounts to finding zeros of the cost function

$$F(c_a, P_{\alpha\beta}) = \left\| \sum_a c_a O_a - \sum_{\alpha, \beta} P_{\alpha\beta} |\varphi_{\alpha}\rangle \langle \varphi_{\beta}| \right\|_2^2$$
$$= X_A^* \mathcal{M}_{AB} X_B \stackrel{!}{=} 0, \qquad (3)$$

where the vector $X_A = (-c_a, P_{\alpha\beta})$ contains both the information on the expansion coefficients $\{c_a\}$ of h in the operator basis $\{O_a\}$, and on the positive matrix $P_{\alpha\beta}$ representing has a strictly positive operator on $(\operatorname{Im}\mathcal{P})^{\perp}$. The norm $\|\cdot\|_2$ is the Frobenius norm. If we choose the basis $\{|\varphi_{\alpha}\rangle\}$ to be orthonormal, $\langle\varphi_{\alpha}|\varphi_{\beta}\rangle = \delta_{\alpha\beta}$, the matrix \mathcal{M} becomes sparse and it reads

$$\mathcal{M}_{AB} = \left(\begin{array}{c|c} M_{ab} & R_{a,\alpha\beta} \\ \hline R_{\gamma\delta,b} & \delta_{\alpha\gamma}\delta_{\beta\delta} \end{array} \right),$$
$$M_{ab} = \operatorname{Tr}(O_a O_b), \quad R_{a,\alpha\beta} = \langle \varphi_{\alpha} | O_a | \varphi_{\beta} \rangle. \tag{4}$$

By solving the eigenvalue problem (3), we thus obtain solutions $h = \sum c_a O_a$ which satisfy (2) and thus vanish on Im \mathcal{P} . However, *h* is not positive (or even non-negative) on $(\text{Im }\mathcal{P})^{\perp}$ unless we impose P > 0. To obtain solutions which additionally satisfy this positivity condition, we thus choose to minimize the quadratic form in Eq. (3) on the convex space of positive definite matrices $P_{\alpha\beta}$. Hence, the optimized Hamiltonian is given by the coefficients c_a such that

$$(c_a, P_{\alpha\beta}) = \operatorname{ArgMin} F(c_a, P_{\alpha\beta}), \qquad P \ge \mathbb{1}, \qquad (5)$$

where we imposed that the eigenvalues of *P* are ≥ 1 without loss of generality (as this only amounts to a rescaling).

A possible route to solve this optimization problem is to apply a gradient descent algorithm to the cost function F, and project onto the desired space at each step. We start from an initial point such that the Hamiltonian h vanishes, i.e., $c_a = 0$ for all a, and the initial $P_{\alpha\beta}$ is the identity. Since the cost function is a quadratic form, the gradient can be efficiently computed by simple matrix multiplication. The algorithm is thus as follows:

(1)
$$X_{in} = (0, 1),$$

- (2) $X' = X \eta \nabla F(X) = X 2\eta \mathcal{M} \cdot X$,
- (3) $X'' = \Pi(X')$,
- (4) Repeat from 2. until convergence.

The projection Π in 3. has to enforce the condition $P \ge 1$. This can be achieved by taking the $P_{\alpha\beta}$ components of the vector *X*, and setting to one all the diagonal elements of the triangular part of its Schur decomposition which dropped below one at step 2., i.e.,

$$\Pi(X) = \begin{pmatrix} c \\ \Pi(P) \end{pmatrix} = \begin{pmatrix} c \\ Z^{\dagger} \Pi(T) Z \end{pmatrix} = \begin{pmatrix} c \\ Z^{\dagger} \widetilde{T} Z \end{pmatrix}, \quad (6)$$

where the triangular matrix \tilde{T} is the same as T, but all the diagonal entries that are smaller than one in T are set to one in \tilde{T} .

To monitor the status of the convergence during the minimization we directly compute the cost function at each step. To speed up the convergence we employ an adaptive step size $\eta_n = \langle \delta X_n, \delta X_n \rangle / \langle \delta X_n, \delta G_n \rangle$, where $\delta X_n = X_n - X_{n-1}$ and $\delta G_n = G_n - G_{n-1}$ are the point and gradient displacements at the *n*th step of the optimization [27].

From a numerical perspective, the main bottleneck of the algorithm comes from the calculation of the basis $\{|\varphi_{\alpha}\rangle\}$ of $(\text{Im }\mathcal{P})^{\perp}$. To this end, we employed linear algebra routines that compute the null space of the operator \mathcal{P}^t , stored in a dense format. We note that the matrix representation of the operator \mathcal{P} is sparse, and its columns provide a sparse basis $\{|\psi_{\alpha}\rangle\}$ of Im \mathcal{P} . Starting from this sparse basis, it might be possible to build a sparse basis of $(\text{Im }\mathcal{P})^{\perp}$, thus improving the efficiency of this step and allowing to treat larger patches.

Once the basis $\{|\varphi_{\alpha}\rangle\}$ is at hand, the computation of the matrix elements $M_{a,b}$ and $R_{a,\alpha\beta}$ in Eq. (4) can be carried out efficiently by storing the local operator basis in a sparse format and performing sparse linear algebra matrix-matrix and matrix-vector multiplication. As we explain below, exploiting the symmetries of the TNS and the operator basis can further reduce the computational cost of determining $|\varphi_{\alpha}\rangle$ and *R*.

B. Symmetries

The dimension of the parameter space of the optimization algorithm is $D_O + (\dim (\operatorname{Im} \mathcal{P})^{\perp})^2$, where D_O is the dimension of the operator basis $\{O_a\}$. The number of parameters can be reduced if the target Hamiltonian is invariant under some symmetry. In this case, the matrix $P_{\alpha\beta}$ can be decomposed into blocks labeled by the eigenvalues of the symmetry generator:

$$P = \bigoplus_{\lambda} P^{\lambda} . \tag{7}$$

Upon proper choice of the basis of local operators, the same is true for all the O_a s. The dimension of the parameter space is thus reduced to $D_O + \sum_{\lambda} (D_{\lambda})^2$, where D_{λ} is the dimension of the intersection of $(\text{Im } \mathcal{P})^{\perp}$ with each eigenspace (irrep) of the symmetry generators. The cost function in Eq. (3) becomes

$$F(c_{a}, P_{\alpha\beta}^{\lambda}) = \sum_{\lambda} \left\| \sum_{a} c_{a} O_{a}^{\lambda} - \sum_{\alpha, \beta} P_{\alpha\beta}^{\lambda} |\varphi_{\alpha}^{\lambda}\rangle \langle\varphi_{\beta}^{\lambda}| \right\|^{2}$$
$$= c_{a} \operatorname{Tr}(O_{a}O_{b}) c_{b} - 2c_{a} \sum_{\lambda} \langle\varphi_{\alpha}^{\lambda}|O_{a}^{\lambda}|\varphi_{\beta}^{\lambda}\rangle P_{\alpha\beta}^{\lambda}$$
$$+ \sum_{\lambda} (P_{\alpha\beta}^{\lambda})^{2} = X_{A}^{*} \mathcal{M}_{AB} X_{B}, \qquad (8)$$

where O^{λ} is the restriction of *O* to the symmetry sector labeled by λ , and we used the fact that O_a is block diagonal for all *a*. The vector $X_A = (-c_a, P_{\alpha\beta}^{\lambda_1}, P_{\alpha\beta}^{\lambda_2}, \dots)$, and the matrix \mathcal{M} reads

$$\mathcal{M}_{AB} = \begin{pmatrix} M & R_{\lambda_1} & 0 & 0 \\ R_{\lambda_1}^{\dagger} & \mathbb{1}_{\lambda_1} & R_{\lambda_2} & 0 \\ 0 & R_{\lambda_2}^{\dagger} & \mathbb{1}_{\lambda_2} & \ddots \\ 0 & 0 & \ddots & \ddots \end{pmatrix},$$
$$M_{ab} = \operatorname{Tr}(O_a O_b), \quad R_{a,\alpha\beta} = \langle \varphi_\alpha | O_a | \varphi_\beta \rangle. \tag{9}$$

An example which we will use in the following is SU(2) symmetry. In this case, we take $\lambda = s, S^z$, where S^z is the eigenvalue of the *z* component of the total angular momentum on the chosen patch of physical sites, $S^z = \sum_i S_i^z$, and *s* is the quantum number of the total squared angular momentum

 $S^2 = (S^x)^2 + (S^y)^2 + (S^z)^2$ (with eigenvalues s(s + 1)). Note that, thanks to SU(2) invariance, the $P_{\alpha\beta}$ blocks are independent of the eigenvalue S^z , and the cost function in Eq. (8) becomes

$$F(c_a, P^{\wedge}_{\alpha\beta}) = c_a \operatorname{Tr}(O_a O_b) c_b$$

- $2c_a \sum_s (2s+1) \langle \varphi^{s,0}_{\alpha} | O^{s,0}_a | \varphi^{s,0}_{\beta} \rangle P^s_{\alpha\beta}$
+ $\sum_s (2s+1) (P^s_{\alpha\beta})^2,$ (10)

where the factor (2s + 1) takes into account the multiplicity of the eigenvalue S^z for a given *s*, so that the generators $|\varphi_{\alpha}^{s,S^z}\rangle$ need to be computed only in the $S^z = 0$ sector. This further reduces the number of variational parameters down to $D_O + \sum_s (D_s)^2$, where D_s is the dimension of the simultaneous eigenspace of S^2 and S^z in the $S^z = 0$ sector.

C. Formulation as a semidefinite program

The minimization problem in Eq. (5) can be formulated as a semi-definite program (SDP). This implies that the problem can be systematically and efficiently solved using a suitable SDP solver, and makes clear why the gradient method chosen for the optimization in this work performs so well on the problem.

To start with, let us consider a variation where we define the cost function F using the operator norm $\|\cdot\|_{\infty}$ rather than the Frobenius norm squared. $\|X\|_{\infty}$ can be obtained by minimizing λ subject to $-\lambda \leq X \leq \lambda$. Thus minimizing Famounts to solving the SDP

$$\begin{array}{l} \underset{\{c_a\},P,\lambda}{\text{minimize }\lambda} \\ \text{subject to } -\lambda \leqslant \sum_{a} c_a O_a - \sum_{\alpha,\beta} P_{\alpha\beta} |\varphi_{\alpha}\rangle \langle \varphi_{\beta}| \leqslant \lambda \\ P \geqslant 0. \end{array}$$

In case the cost function is defined using the Frobenius norm, as in Eq. (3), one can rewrite $||X||_2$ as the minimum of λ subject to

$$egin{pmatrix} \lambda \mathbb{1} & X \ X^\dagger & \lambda \mathbb{1} \end{pmatrix} \geqslant 0 \ ,$$

to yet again re-express the minimization of F as an SDP.

IV. BENCHMARKS

Let us now benchmark our method with two well-studied models. The 1D AKLT model and the Toric Code on the square lattice.

A. The AKLT state on a chain

The AKLT state [22,23] is constructed by placing spin-1/2 singlets on a chain and projecting every two adjacent spin-1/2's to the joint spin-1 (symmetric) subspace. The resulting spin-1 chain is rotationally [i.e., SO(3)] invariant and can be described as an MPS with the tensors given in Fig. 3(a). This MPS is normal and becomes injective upon blocking L = 2 sites. Thus the elementary proof technique of inverting \mathcal{P} on



FIG. 3. (a) Tensor Network representation of the AKLT state. (b) Blocking three sites of the MPS tensors yields the map \mathcal{P} , starting point of the parent Hamiltonian construction. In this case the number of blocked sites N_s is larger than the minimum N_s required to build the most local parent Hamiltonian Eq. (11).

the injective block results in a parent Hamiltonian acting on $N_s = 4$ sites, while the more refined techniques allow to prove well-behavedness of the parent Hamiltonian on $N_s = 3$ sites. On the other hand, it is well known that the AKLT state is the unique ground state of the two-body Hamiltonian

$$H = \sum_{i=1}^{L} \left[\frac{1}{2} \vec{S}_i \cdot \vec{S}_{i+1} + \frac{1}{6} (\vec{S}_i \cdot \vec{S}_{i+1})^2 + \frac{1}{3} \right], \quad (11)$$

where $\overline{S} = (S^x, S^y, S^z)$ is the spin-1 representation of SU(2). Indeed, the Hamiltonian (11) can be obtained by applying the parent Hamiltonian construction to two sites. Once one has obtained a guess for a two-body Hamiltonian in this way, it is straightforward to verify that it is indeed a well-behaved parent Hamiltonian, by checking that its ground space on three (or four) sites is just the same as that of the three-site (or foursite) parent Hamiltonian and thus, it has a unique ground state and a gap.

For this approach, however, a prior guess for a suitable two-body Hamiltonian is required. We will now demonstrate that our method can be used to break up the canonical three-body or four-body AKLT parent Hamiltonian into the two-body Hamiltonian (11) without any such prior knowledge. Since we aim to decompose the Hamiltonian into SU(2)-invariant two-body nearest neighbor terms, we only need to include nearest-neighbor Heisenberg interactions in the operator basis $\{O_a\}$; for instance, with $N_s = 3$ the operator basis is {Id, $\vec{S}_1 \cdot \vec{S}_2$, $\vec{S}_2 \cdot \vec{S}_3$, $(\vec{S}_1 \cdot \vec{S}_2)^2$, $(\vec{S}_2 \cdot \vec{S}_3)^2$ }. Fig. 4 shows the results of the minimization procedure. In Figs. 4(a) and 4(b), we plot the cost function during the (projected) gradient descent optimization for $N_s = 3$ and $N_s = 4$, respectively. We compare a constant step $\eta = 0.2$ (green line) and an adaptive step $\eta = \text{Min}\{\langle \delta X_n, \delta X_n \rangle / \langle \delta X_n, \delta G_n \rangle, \eta_{\text{max}}\}$ (orange and blue lines).¹ Figures 4(e) and 4(f), demonstrate that the matrix $P_{\alpha\beta}$

¹A maximum step size η_{max} is necessary to ensure stability of the optimization. $\eta = 0.2$ is the maximum size yielding a stable constant step optimization.



FIG. 4. [(a) and (b)] Cost function F, Eq. (3), during the minimization. The adaptive step-gradient descent procedure converges much faster (~200 and 1500 steps to reach $F \sim 10^{-12}$ for $N_s = 3$ and 4, respectively) than the constant step one ($\eta = 0.5$). [(c) and (d)] Coefficients of the operators $\{O_a\}$ in the optimal decomposition at convergence. The basis ordering is $\{\text{Id}, \vec{S}_i \cdot \vec{S}_{i+1}, \ldots, (\vec{S}_i \cdot \vec{S}_{i+1})^2, \ldots\}$ with $i = 1, \ldots, N_s - 1$. (e),(f) Eigenvalues of the matrix $P_{\alpha\beta}$ at convergence ($F < 10^{-12}$) for $N_s = 3$ and 4. The result shows that the locality-optimized Hamiltonian, acting on N_s sites, is not the projector onto (Im \mathcal{P})[⊥].

obtained, and thus the Hamiltonian density, is not a projector on the original $N_s = 3, 4$ sites. Rather, the Hamiltonian obtained is equal to Eq. (11) within numerical precision, as shown in Figs. 4(c) and 4(d).²

B. The toric code on the square lattice

Let us now turn to a paradigmatic two-dimensional model which exhibits topological order: Kitaev's toric code [28]. Consider an arbitrary lattice, and assign a qubit $\{|0\rangle, |1\rangle\}$ to every edge of the lattice. The toric code wave function is then given by the equal weight superposition of all configurations which obey a \mathbb{Z}_2 Gauss law around every vertex, i.e., there are an even number of $|1\rangle$ states adjacent to every vertex. Equivalently, this amounts to saying that the wave function is an equal weight superposition of all closed loop configurations on the lattice, where the states $|0\rangle$ and $|1\rangle$ correspond to



FIG. 5. (a) Vertex tensors providing the PEPS representation of the toric code ground state. The virtual index 1 signals the presence of a loop entering the vertex. $R_{\pi/2}$ stands for all the possible $\pi/2$ rotations of the vertex tensors. (b) Local projectors that map diagonally the virtual space of the vertex tensors to the physical space of a site on the tilted square lattice. (c) $N_s = 12$ sites patch used for the parent Hamiltonian construction. Physical legs are omitted in the drawing for clarity; that is, each red dot additionally carries a physical index, as in (b). The linear map \mathcal{P} obtained from this patch maps the virtual space (\mathbb{C}^2)^{$\otimes 12$}. Plaquette terms are product of *X* s on the green line, cross terms are product of *Z* s on the blue line.

the no-loop and loop state, respectively. In the following, we focus on the square lattice. A tensor network representation of the toric code can be constructed by using two types of tensors: One vertex tensor which only carries virtual indices and which enforces the \mathbb{Z}_2 Gauss law [see Fig. 5(a)] and an edge tensor which is described by a Kronecker delta and which "copies" the virtual degree of freedom to a physical qubit [see Fig. 5(b)]; these tensors are arranged to a tensor network as shown in Fig. 5(c).

Conceptually, a parent Hamiltonian constructed on a region R (with edge tensors at its boundary) ensures that for any given edge configuration, all bulk loop configurations get the same weight. It is easy to see that such a Hamiltonian will enforce the closed loop constraint, as well as enforce an equal amplitude for all loop configurations which can be coupled by local moves; this gives rise to a fourfold degeneracy on the torus, with sectors labeled by the parity of loops around the torus in either direction [19,28]. To construct a parent Hamiltonian with controlled ground space degeneracy, we start by noting that the map \mathcal{P} given by a vertex tensor surrounded by four edge tensors is injective.³ A resulting canonical parent Hamiltonian can then be constructed on the 12-sites patch shown in Fig. 5(c); it is straightforward to prove that it satisfies

²We note that for $N_s = 4$, the Hamiltonian density is not homogeneous on the four-site patch, rather the coefficients of $S_2 \cdot S_3$ and $(S_2 \cdot S_3)^2$ are larger at the center of the patch. Yet, as the density has to be summed on all sites to construct the parent Hamiltonian on a ring, the latter is translation-invariant and coincides with Eq. (11) up to multiplicative factors.

³In principle, three edge tensors are sufficient. This results in a slightly smaller parent Hamiltonian (some edge sites—but not all!— at its boundary can be omitted in Fig. 5(c), which however breaks the lattice symmetry. We thus choose to work with the given four-site injective patch.



FIG. 6. (a) Decomposition of the projector-valued parent Hamiltonian $h_{\rm Id}$ of the toric code in terms of *N*-body Pauli products; the plot shows the total weight of all *N*-body terms. Terms with all possible weights are needed in the decomposition. (b) Cost function during the minimization, using the adaptive step. Symmetric and nonsymmetric versions of the algorithm achieve convergence in the same number of steps, but the symmetric version reduces significantly the computation cost. (c) Coefficients of the operators $\{O_a\}$ in the optimal decomposition at convergence. When only cross and plaquette terms (made of *X*, *Y* or *Z*) are included in the basis, the algorithm produces the expected result within machine precision. (d) Eigenvalues of the matrix $P_{\alpha\beta}$ at convergence, divided in the two physical \mathbb{Z}_2 symmetry sectors. The most local Hamiltonian is different from the projector on (Im \mathcal{P})[⊥].

the intersection property by using the refined "invert and grow back" techniques discussed, e.g., in Refs. [8,13], and thus has the correct ground space structure.

In the following, we apply our method to analyze how one can break down the canonical parent Hamiltonian acting on patches of $N_s = 12$ sites into simpler terms, where one goal is to check whether our optimization method does indeed return the toric code Hamiltonian

$$H = -\sum_{v} \prod_{i \in v} Z_i - \sum_{p} \prod_{i \in p} X_i , \qquad (12)$$

which in known to have the toric code wave function as its unique ground state (up to topological degeneracy) [28]. Here, the terms in the first sum act on the four qubits surrounding any given vertex v, and those in the second sum on the four qubits surrounding any given plaquette p of the square lattice, as shown in Fig. 5(c), where X and Z are Pauli matrices.

To this end, we start from the $N_s = 12$ site patch depicted in Fig. 5(c), for which the dimension of $(\text{Im } \mathcal{P})^{\perp}$ is 3968. Figure 6(a) shows the decomposition of the projector-valued Hamiltonian h_{Id} [i.e., with P = Id in Eq. (2)] in an operator basis made of all the possible products of Pauli matrices on the 12 sites, grouped by the number N_s of nontrivial Pauli operators: We observe that h_{Id} contains terms all the way up to $N_s = 12$ -body terms, and thus, optimization over *P* is required to obtain a simpler decomposition of the parent Hamiltonian *h* into Pauli products.

We now apply the algorithm of Sec. III starting from an operator basis with all possible products of Pauli matrices on four-site crosses and plaquettes [see Fig. 5(c)] that respect the physical symmetries of the TN state on the chosen patch: Specifically, we used the \mathbb{Z}_2 symmetry generated by $\prod_{i=1}^{N_s} Z_i$, the fourfold $\pi/2$ -rotation $R_{\pi/2}$ around the center of the patch, and the reflection \mathcal{R} with respect to the horizontal axis that goes through the center of the patch. This basis contains 25 operators (including the identity), and the total number of variational parameters for the nonsymmetric version of the algorithm would be $25 + 3968^2 \simeq 1.6 \times 10^7$. Given the aforementioned symmetries, we can apply the symmetric version of the algorithm to reduce the number of parameters. In particular, we consider the algorithm where we use the global \mathbb{Z}_2 symmetry only, which reduces the dimension of the variational space down to $25 + 2 \cdot (3968/2)^2 \simeq 7.9 \times 10^6$, and the fully symmetric version where we exploit \mathbb{Z}_2 , $R_{\pi/2}$, and \mathcal{R} symmetries, yielding a variational space dimension of $1466797 \simeq 1.5 \times 10^6$, and compare their performance. The cost function during the minimization is plotted in Fig. 6(b). Both versions of the algorithm converge to the minimum in \sim 1000 steps. However, it is computationally much cheaper for the most symmetric version to perform a single step, reducing the CPU time by a factor proportional to the ratio between the variational space dimensions without and with symmetries. Although in this example, exploiting the full symmetry group of the patch is not indispensable, it will be crucial in the next example, where even a single optimization step would be prohibitive without it.

Considering the optimum found by the algorithm, we find that our method indeed yields a parent Hamiltonian identical to the one in Eq. (12), except for a different relative weight of the two types of terms (this is possible as the terms commute, and the Hamiltonian is frustration-free, i.e., the ground state minimizes each of the four-body terms individually). In Fig. 6(c), we show the coefficients of the Hamiltonian density on the 12-site patch at convergence. In Fig. 6(d), we plot the eigenvalues of the optimized parent Hamiltonian h on the 12-site patch, demonstrating that the most local h is indeed not a projector.

V. THE SU(2) RESONATING VALENCE BOND STATE ON THE KAGOME LATTICE

In the following, we apply our method to find an optimally local parent Hamiltonian for the paradigmatic resonating valence bond state on the kagome lattice, which is a prime example of a topological spin liquid. For this model, the simplest hitherto known parent Hamiltonian was a general operator acting on a whole kagome star, that is, $N_s = 12$ sites. Applying our method, we arrive at a much simpler Hamiltonian, where each term is a product of at most four Heisenberg interactions. This represents a significant simplification over the original 12-body interaction and demonstrates the power of our method.

Resonating valence bond (RVB) states are constructed as equal-weight superpositions of all possible singlet coverings between nearest neighbors on a given 2D lattice. In quantum dimer models, singlets are replaced by orthogonal dimers, facilitating their analysis. On frustrated lattices, such dimer models have been known for a long time to be simple representatives of topologically ordered phases [29]; in particular, the dimer model on the kagome lattice is a topological \mathbb{Z}_2 fixed point model, locally unitarily equivalent to the toric code. When orthogonal dimer coverings are replaced by spin-1/2 singlets, one obtains the RVB state, which is a good candidate for describing the physics of frustrated magnets. On the kagome lattice, it was shown to be in the same \mathbb{Z}_2 spin liquid phase as the kagome dimer model [13], and demonstrated to be even more stable against perturbation [30]. Starting from the SU(2) singlets RVB state, simple ansatze for the ground state wave functions of physically relevant models have been devised [31]. RVB states have a natural PEPS representation [12,19] with a \mathbb{Z}_2 entanglement symmetry; they become \mathbb{Z}_2 -injective upon blocking and thus, their canonical parent Hamiltonians exhibit a fourfold degenerate ground space on the torus, as required for a \mathbb{Z}_2 topological spin liquid. However, the hitherto known parent Hamiltonians for the kagome RVB state are rather complicated: the simplest parent Hamiltonian which has been obtained using canonical techniques is constructed on two overlapping stars, that is, 19 sites [13]; later, it has been shown by brute-force numerical checking that the parent Hamiltonian constructed from the map \mathcal{P} on a single star, that is, $N_s = 12$ sites, has the same ground space as the 19-site two-star Hamiltonian when applied on both of the stars [24]. In the following, we will apply our algorithm to analyze whether, and how, this one-star Hamiltonian can be broken down into simpler terms.

We use the TN representation of the SU(2) RVB state introduced in Ref. [13], which is given in Fig. 7. Figure 7(c) shows the patch of 12 sites that we consider in what follows, where the virtual space consists of six bonds with dimension D = 3. The PEPS is \mathbb{Z}_2 -injective, with the virtual symmetry generator g = diag(1, 1, -1). Since the dimension of $(\text{Im } \mathcal{P})^{\perp}$ is 3731, the number of variational parameters which arise from the matrix P [Eq. (2)] would be more than 10⁷. As this is prohibitive, we will need to use the symmetries of the system at hand.

Let us start with the symmetries of the parent Hamiltonian h, that is, the basis $\{|\varphi_{\alpha}\rangle\}$ on which h is supported and the associated positive matrix P > 0, $h = \sum P_{\alpha\beta} |\varphi_{\alpha}\rangle \langle \varphi_{\beta}|$. The map \mathcal{P} of the one-star tensor network in Fig. 7(c) (which maps the virtual to the physical system) commutes with both SU(2)and a 60° rotation; by choosing boundary conditions with well-defined quantum numbers on the virtual system, we can thus ensure that we obtain $|\varphi_{\alpha}\rangle$ with well-defined SU(2) and angular momentum quantum numbers. In addition, we can ensure that the $|\varphi_{\alpha}\rangle$ transform nicely under reflection about the vertical axis: Such a reflection changes $\varepsilon_{abc} \rightarrow -\varepsilon_{abc}$ in Fig. 7(b), that is, the state acquires a minus sign for each blue tensor in the ε_{abc} configuration. The number of those tensors equals the number of triangles which hold a singlet (or dimer), and the number of singlets inside the star is determined by a simple counting argument from the number of spin-1/2states at the virtual boundary (i.e., boundary configurations $\{|0\rangle, |1\rangle\}$, or more precisely this number modulo 4. This quantum number of the boundary condition commutes with



FIG. 7. TN representation of the spin-1/2 RVB state. The on-site tensors (a) and triangle plaquette tensors (b)—only nonzero entries are shown—are placed and contracted on the kagome lattice as shown in (c). The states 0 and 1 carry the spin-1/2 degree of freedom, while the state 2 signals the absence of a spin 1/2. The on-site tensor selects either the left or right singlet and maps it to the physical spin 1/2; the triangle plaquette tensor can either hold no singlet (222 configuration), or exactly one singlet (ε_{abc}). Up- and downward-pointing blue tensors are related by rotation. Physical legs are omitted in the drawing for clarity; that is, each red dot additionally carries a physical index, just as in (a).

SU(2) and angular momentum, and thus, we can obtain a basis of $|\varphi_{\alpha}\rangle$ labeled by SU(2), rotation, and reflection quantum numbers. By exploiting these symmetries, as described in Sec. III B, we are able to break down the $P_{\alpha\beta}$ variational matrix into blocks, yielding 20 931 parameters.

We exploit the same symmetries to select the operators to be included in the basis $\{O_a\}$. As we show in the Appendix, every SU(2)-invariant Hamiltonians H on $(\mathbb{C}^2)^{\otimes N}$ can be decomposed as $H = \sum w_a K_a$ in a basis of SU(2)-invariant Hamiltonians $\{K_a\}$, where each K_a is a nonoverlapping product of only two types of terms: Heisenberg interactions $\vec{S}^i \cdot \vec{S}^j$, and "chiral" terms of the form $\vec{S}^i \cdot (\vec{S}^j \times \vec{S}^k)$; in addition, at most one such chiral term is needed. We use this basis of SU(2)-invariant Hamiltonians to build the basis of operators $\{O_a\}$ for our parent Hamiltonian. In this process, we make use of further symmetries: First, we observe that the RVB wave function is real, and thus, the parent Hamiltonian can be chosen real as well, $h = \bar{h}$ (more specifically, we can replace any parent Hamiltonian h by $(h + \bar{h})/2$, which changes neither the ground space nor the spectrum). Since $\vec{S}^1 \cdot (\vec{S}^2 \times \vec{S}^3)$ is purely imaginary (as it is a sum of terms with one S_v each) and it appears at most once, this implies that we can omit it altogether, and the basis $\{O_a\}$ can be chosen to be spanned solely by nonoverlapping products of Heisenberg terms. Further, we impose that the basis $\{O_a\}$ has the same lattice symmetries as used for the $\{|\varphi_{\alpha}\rangle\}$, i.e., reflection about the vertical axis and rotation by 60° .

We thus have that each operator O_a is a symmetrized product of $1 \le r \le 6$ Heisenberg interactions. As we are interested in finding the most local parent Hamiltonian, we apply our method to a growing sequence of basis sets $\mathcal{O}_r = \{O_a\}$, where a given \mathcal{O}_r consists of all operators with *up to r* Heisenberg



FIG. 8. Operator basis dimension for the two cases considered here. (a) The basis contains all products of up to r Heisenberg interactions. (b) The basis contains all products of up to r = 4 Heisenberg interactions, restricted to the spins on the central hexagon and up to N_{Δ} continguous outer spins. In both cases, the operators are symmetrized w.r.t. the sixfold rotation and twofold reflection symmetry of the star, see text.

interaction [see Fig. 8(a) for the dimension of those basis sets, going up to about 6000). Figure 9(a) shows the results obtained by our algorithm when approximating the parent Hamiltonian with the basis set \mathcal{O}_r , for r = 1, ..., 4. We find that while for $r \leq 3$, the parent Hamiltonian cannot be faithfully approximated, the basis set \mathcal{O}_4 , which contains products of up to r = 4 Heisenberg terms, provides an approximation of the exact parent Hamiltonian up to machine precision.⁴

In order to investigate whether it is possible to further simplify the Hamiltonian, we analyze the effect of restricting \mathcal{O}_r to more local Heisenberg terms; specifically, we define $\mathcal{O}_4(N_{\Delta})$ to contain all products of up to 4 Heisenberg terms which act on the six central spins and up to N_{Δ} adjacent spins at the tips of the star, as shown in Fig. 8(b) (which also provides the basis size). Figure 9(c) shows the results on the accuracy of the approximation of the parent Hamiltonian using the basis set $\mathcal{O}_4(N_{\Delta})$: While for $N_{\Delta} \leq 3$, the parent Hamiltonian is not well approximated, we can reproduce the Hamiltonian up to machine precision for $N_{\Delta} = 5$. The case $N_{\Delta} = 4$ lies in between those regimes, as it provides a very good but not perfect approximation of the parent Hamiltonian, with an error of about 10⁻⁵ in Frobenius norm squared.

In order to double-check the correctness of the Hamiltonian found by our algorithm, we performed exact diagonalization on small tori, employing periodic clusters including up to 18 sites on the kagome lattice [see Fig. 10(a)]. Figure 10 shows comparisons between the spectrum of the 12-site projector-valued parent Hamiltonian and the optimized Hamiltonians returned by our method (for r = 4 and $N_{\Delta} = 4$, respectively), for clusters of 12 and 18 sites. The expected



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FIG. 9. [(a) and (c)] Cost function *F*, Eq. (3), at the optimum as a function of *r* and N_{\triangle} , respectively (see Fig. 8). A machine-precision approximation is obtained for r = 4 and $N_{\triangle} = 5$, respectively, but $N_{\triangle} = 4$ already provides a fairly accurate approximation. [(b) and (d)] Eigenvalues of $P_{\alpha\beta}$ in the $S^z = 0$ sector vs the value s(s + 1) of the total spin S^2 , for r = 4 and N_{\triangle} , respectively, showing that the optimal solution is not a projector.

topological fourfold degeneracy of the ground space is accurate within 10^{-9} ; this is in agreement with the accurate approximation of the parent Hamiltonian whose ground space has an exact fourfold degeneracy. On the other hand, as expected, the two spectra already differ for low-lying excited states. Similarly, the spectra of the optimized parent Hamiltonians on the full star for r = 4 and $N_{\Delta} = 4$ [Figs. 9(b) and 9(d), respectively) are far from the projector-valued parent Hamiltonian. We also verified that the overlap of the RVB state with the ground-space of the optimized Hamiltonian equals 1 within numerical accuracy, for clusters up to N = 24 sites.

An interesting question is whether one can gain physical intuition on the operators that are essential to obtain a good parent Hamiltonian for the RVB state. Unfortunately, we found this to be hard in practice, due to the large number of operators in the basis (~ 3000 at least). Although certain operators contribute far more than others to the final result (e.g., the number of operators O_a with weights $|c_a/c_0| > 0.5$, where c_0 is the largest coefficient of the identity operator, are only 10), all the operators appear to be necessary to give high overlap with the RVB state, and to reproduce the correct ground state degeneracy.

VI. CONCLUSIONS

A. Summary

In this work, we have presented an algorithm to systematically simplify parent Hamiltonians for tensor network states such as MPS and PEPS. Specifically, our method allows to decompose a parent Hamiltonian into a sum of terms chosen

⁴Depending on the basis size the algorithm may take several thousands of steps to converge, and it does not always find an exact parent Hamiltonian. This is signaled by the fact that at convergence—we monitor the status of convergence by computing the norm displacement vector at consecutive steps $\epsilon = ||X_n - X_{n-1}||$, and stop the algorithm when $\epsilon < 10^{-10}$ —the cost function is not vanishing within numerical precision.



FIG. 10. (a) Periodic clusters employed for the computation of the spectrum of the parent Hamiltonian. The latter is obtained as a sum of the parent Hamiltonian *h* on all stars of the cluster. [(b)– (e)] Comparison between the low-energy spectra of the RVB parent Hamiltonian obtained as a projector on $(\text{Im }\mathcal{P})^{\perp}$ (blue empty diamonds) and the optimized Hamiltonian (orange filled diamonds) obtained on a basis of operators with up to four Heisenberg interactions [(b) and (c)] and up to four triangles around the hexagon [(d) and (e)]. Periodic clusters with N = 12 [(b) and (d)] and N = 18[(c) and (e)] are considered.

from any given set of elementary operators, while preserving the ground space as well as the gappedness of the original parent Hamiltonian. A central ingredient is the remaining degree of freedom in the parent Hamiltonian construction: while its local terms have a fixed ground space, they can have an arbitrary excitation spectrum. Our method exploits this degree of freedom to optimize for the parent Hamiltonian which is best approximated by the given basis set of elementary interactions. This results in a convex optimization problem which can formally be written as a semidefinite program, and which can thus be solved efficiently by a gradient-based algorithm to find the optimal decomposition. Additionally, our algorithm can be combined with symmetries of the parent Hamiltonian (that is, the symmetries of the tensor network state) by imposing the same symmetries on the basis set, which leads to a significant reduction in computational resources.

We have applied our method to three paradigmatic tensor network models. First, the 1D AKLT model, where we found that our algorithm allows to decompose the canonically constructed three- or four-body parent Hamiltonians into the well-known two-body AKLT Hamiltonian. Next, the toric code model, where refined canonical constructions yield a parent Hamiltonian acting on 12 sites, which our algorithm would break down into the well-known toric code Hamiltonian consisting of four-body vertex and plaquette terms.

Finally, we have studied the resonating valence bond (RVB) state on the kagome lattice, which is a prime example of a topological spin liquid, and which possesses a succinct PEPS representation. However, while the RVB state (as a \mathbb{Z}_2 -injective PEPS) is the exact fourfold degenerate ground state of a local parent Hamiltonian, the smallest canonically constructed parent Hamiltonian still acts on two overlapping stars, that is, 19 spins, which had been shown by brute force to be decomposable as the sum of two one-star, i.e., 12-body, terms. The application of our algorithms to this 12-site Hamiltonian results in a significant simplification: We find that the RVB parent Hamiltonian can be decomposed into a sum of interaction terms each of which is a product of no more than four Heisenberg interactions, a notable improvement over the general 12-body interaction; moreover, the range of those interactions can be further restricted.

Altogether, this demonstrates the ability of our algorithm to significantly simplify the locality of parent Hamiltonians beyond the abilities of existing proof techniques, and opens up the possibility to identify tensor network models which are ground states of particularly simple parent Hamiltonians.

B. Discussion

Let us conclude by discussing two aspects of our method which might raise natural questions.

The first point is that in the decomposition $h = \sum c_a O_a$ of the parent Hamiltonian [Eq. (1)], the local terms \overline{O}_a are, in general, not frustration free. This is, in fact, an advantage of our method, since this way, we can obtain decompositions with smaller local terms. In case one wants to restrict the search to decompositions into frustration free terms-as is the case for the AKLT or the Toric Code model-this is in fact a much easier task. In that case, it is sufficient to check whether the reduced density operator ρ for any subset of sites on the region supporting h is rank deficient, construct parent Hamiltonians on those patches (e.g., as projectors onto the kernel of ρ), and verify whether their joint ground space coincides with that of h. However, this method will fail to further simplify, e.g., the one-star parent Hamiltonian of the RVB model [13]. Thus, in order to obtain the most simple decomposition, one will generally want to consider decompositions into arbitrary local terms, and this is what our method achieves. Let us note that a situation akin to the RVB also arises for the Majumdar-Ghosh model [32]: The canonical parent Hamiltonian acts on four adjacent sites, which can be decomposed into a sum of two frustration-free terms acting on three adjacent sites. The latter can, yet again, be expressed as the sum of three (locally frustrated) Heisenberg interactions acting on nearest and nextnearest neighbor spins, respectively, while no decomposition of the Majumdar-Ghush Hamiltonian into frustration free twosite terms exists.

Fortunately, the decomposition into locally frustrated terms does not affect the power of the parent Hamiltonian framework, such as tools to characterize the ground space or to prove gaps: since we decompose the local, frustration free parent Hamiltonian terms h as $\sum c_a O_a$, frustration freeness is re-gained when considering that local sum, and the methods developed for parent Hamiltonian generally only depend on the correct ground space of h, but not on the precise form of the higher eigenspaces of h.

A second point of concern might be that in general, the decomposition $h = \sum c_a O_a$ obtained by our method is exact only up to numerical accuracy, unless one manages to first find an analytic expression for the c_a and then to analytically verify the exact validity of $h = \sum c_a O_a$, which seems rather challenging beyond frustration free decompositions. However, there are a few points to notice: First, the clear jump in accuracy to machine precision when reaching a large enough set of O_a 's [Figs. 9(a) and 9(c)] strongly suggests that the decomposition is exact. Second, for many practical purposes, such as the construction of simple Hamiltonians for the experimental realization of topological spin liquids in quantum simulators, one is interested in systems of comparatively small size (that is, rather $\sim 10^4$ than 10^{23} particles), in which case perturbations of the Hamiltonian on the order of machine precision will not change the ground state, as long as there is a reasonably sized gap. And finally, in order to be able to meaningfully talk about a gapped phase, the physics of a system ought to be stable against sufficiently small (but finite) local perturbations even in the thermodynamic limit, and thus, for such "reasonable" systems, a sufficiently accurate decomposition $h \approx \sum c_a O_a$ will exhibit the same physics; in fact, for frustration-free systems, and in particular tensor network states, it has been shown that such a notion of stability against small but extensive perturbation follows from a condition on the Hamiltonian called LTQO [33–36].

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APPENDIX: DECOMPOSITION OF SU(2)-INVARIANT HAMILTONIANS INTO HEISENBERG AND CHIRAL THREE-BODY TERMS

In this Appendix, we show that any SU(2)-invariant Hamiltonian can be decomposed as a sum of terms, each of which only consists of nonoverlapping products of Heisenberg interactions $\vec{S}^1 \cdot \vec{S}^2$ and (at most one) chiral three-body term $\vec{S}^1 \cdot (\vec{S}^2 \times \vec{S}^3)$.

Let us first consider a general (not necessarily hermitian) SU(2)-invariant operator X on $(\mathbb{C}^2)^{\otimes N}$. It is well-known that the space of all operators X such that $[X, u^{\otimes N}] = 0$ for all

 $u \in SU(2)$ is spanned by the canonical representation V_{π} of the permutation group $\mathfrak{S}_n \ni \pi$, where V_{π} acts by permuting the tensor components of $(\mathbb{C}^2)^{\otimes N}$ [37]. Any permutation π can be expressed as a product of 2-cycles (i, j), i.e., swapping two elements *i* and *j*. Since $V_{(i,j)} = (\vec{S}^i \cdot \vec{S}^j + 1)/2$ (where we normalize the spin operator $\vec{S}^i = (S_x^i, S_y^i, S_z^i)$ at position *i* to have eigenvalues ± 1), we find that the space of all *X* is spanned by products of Heisenberg interactions $\vec{S}^i \cdot \vec{S}^j = \sum \delta_{ab} S_a^i S_b^j$. Note that this includes cases where the different Heisenberg terms overlap.

We will now show that one can get rid of overlapping Heisenberg terms at the cost of introducing just one additional type of interaction, namely $\vec{S}^i \cdot (\vec{S}^j \times \vec{S}^k) = \sum \varepsilon_{abc} S^i_a S^j_b S^k_c$. To this end, consider the overlapping term

$$\begin{split} (\vec{S}^{1} \cdot \vec{S}^{2})(\vec{S}^{2} \cdot \vec{S}^{3}) \\ &= \sum_{abcd} \delta_{ab} S_{a}^{1} S_{b}^{2} \delta_{cd} S_{c}^{2} S_{d}^{3} \\ \stackrel{(*)}{=} \sum_{abcde} \delta_{ab} \delta_{cd} i \varepsilon_{bce} S_{a}^{1} S_{e}^{2} S_{d}^{3} + \sum_{abcd} \delta_{ab} \delta_{cd} \delta_{bc} S_{a}^{1} S_{d}^{3} \\ &= \sum_{bce} i \varepsilon_{bce} S_{a}^{1} S_{e}^{2} S_{d}^{3} + \sum_{ad} \delta_{bc} S_{a}^{1} S_{d}^{3} \\ &= i \vec{S}^{1} \cdot (\vec{S}^{2} \times \vec{S}^{3}) + \vec{S}^{1} \cdot \vec{S}^{3}, \end{split}$$

where in (*), we have used that

$$S_a S_b = \sum_c i \varepsilon_{abc} S_c + \delta_{ab} \mathbb{1}.$$
 (A1)

We have thus succeeded in rewriting a product of two overlapping Heisenberg interactions as a linear combination of elementary two- and three-body interactions $\vec{S}^i \cdot \vec{S}^j$ and $\vec{S}^i \cdot$ $(\vec{S}^j \times \vec{S}^k)$. We can now continue this analysis for all possible overlaps of those two types of terms, making use of (A1) and the summation rules for δ and ε tensors. Even without carrying out this analysis explicitly, it is easy to see that it allows us to transform arbitrary products of those two- and three-body terms into *nonoverlapping* products of the same two types of terms: Every application of (A1) gets rid of one overlap of two terms at one position while introducing an ε , and sums over δ and ε tensors always yield linear combinations of products of δ and ε tensors with no joint indices. Note that this also gives a constructive procedure to arrive at such a decomposition.

We thus find that any operator X with $[X, u^{\otimes N}] = 0$ can be expressed as a complex linear combination of nonoverlapping products of terms $\vec{S}^i \cdot \vec{S}^j$ and $\vec{S}^i \cdot (\vec{S}^j \times \vec{S}^k)$; since those are all hermitian, they also span the set of all hermitian matrices H with $[H, u^{\otimes N}] = 0$ over the real numbers. An additional simplification can be obtained by observing that a product of two ε can be replaced by a linear combination of δ 's ⁵, and thus, in the basis we require only products of Heisenberg interactions with at most one three-body term $\vec{S}^1 \cdot (\vec{S}^2 \times \vec{S}^3)$.

 $^{{}^{5}\}varepsilon_{ijk}\varepsilon_{lmn} = \delta_{il}(\delta_{jm}\delta_{kn} - \delta_{jn}\delta_{km}) - \delta_{im}(\delta_{jl}\delta_{kn} - \delta_{jn}\delta_{kl}) + \delta_{in}(\delta_{il}\delta_{km} - \delta_{im}\delta_{kl})$

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