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Detecting a $Z_2$ topologically ordered phase from unbiased infinite projected entangled-pair state simulations

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We present an approach to identify topological order based on unbiased infinite projected entangled-pair states simulations, i.e., where we do not impose a virtual symmetry on the tensors during the optimization of the tensor network ansatz. As an example we consider the ground state of the toric code model in a magnetic field exhibiting $Z_2$ topological order. The optimization is done by an efficient energy minimization approach based on a summation of tensor environments to compute the gradient. We show that the optimized tensors, when brought into the right gauge, are approximately $Z_2$ symmetric, and they can be fully symmetrized a posteriori to generate a stable topologically ordered state, yielding the correct topological entanglement entropy and modular $S$ and $U$ matrices. To compute the latter we develop a variant of the corner-transfer matrix method, which is computationally more efficient than previous approaches based on the tensor renormalization group.

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I. INTRODUCTION

Since the discovery of the fractional quantum Hall effect [1], the understanding of topologically ordered phases has been a central subject in many-body physics. These phases do not fall under the standard paradigm of Landau symmetry-breaking theory and therefore cannot be characterized in terms of a local order parameter [2]. They exhibit remarkable properties, including a ground-state degeneracy depending on the topology of the system, emergent anyonic excitations [3,4], and robustness against local perturbations, which makes them a promising platform for quantum computing [5]. However, in general it has proven to be challenging to determine, starting from a microscopic Hamiltonian $\hat{H}$, whether the ground state is in a topologically ordered phase.

In recent years substantial progress in studying and classifying these phases has been achieved on the basis of tensor networks. Many studies based on matrix product states (MPS) have shown that topologically ordered phases can be identified [6–9], including the characterization of their emerging anyonic excitations (see, e.g., Refs. [10–17]). However, due to the one-dimensional nature of the MPS ansatz, these studies are limited to cylinders up to a certain width. Projected entangled-pair states (PEPS) [18–20], which are a generalization of MPS to two dimensions, provides a more natural framework for the study of two-dimensional ($2D$) topologically ordered systems [21,22]. There exist a wide range of (nonchiral [23]) topologically ordered states which have an exact and simple PEPS representation, such as, e.g., the ground states of the toric code model [20], string-net models [24,25], or resonating valence-bond states [26]. It has been shown that the topological order is encoded locally in the PEPS tensors by respecting a certain symmetry on their virtual degrees of freedom [22], depending on the type of topological order. The characterization of these virtual symmetries and their associated topologically ordered phases has been under active development in recent years [27–33].

However, it has been shown that already a weak violation of the virtual symmetry destroys the associated topological order [34,35]. Thus, in practical calculations, when performing an optimization starting from random initial tensors, one may expect that already small numerical errors in the optimization of the tensor network ansatz will result in tensors which are not perfectly symmetric, and thus it seems challenging to correctly identify a topologically ordered phase. A way to circumvent this problem is to impose the virtual symmetry on the tensors during the optimization [36], but this requires knowledge of the virtual symmetry beforehand. Without a priori knowledge of the ground state of a given Hamiltonian, one would need to run many simulations, starting from tensors with different virtual symmetries, in order to identify the true ground state (given by the state with lowest variational energy). Since the optimization of the tensors is the computationally most expensive part in a tensor network calculation, it would be desirable to be able to start from unbiased simulations (i.e., without imposing a virtual symmetry), and to identify the topological order a posteriori. However, due to the sensitivity to perturbations, it has so far been unclear whether this is actually possible in practice (up until very recently [37]; see comment below).

In this paper we demonstrate that the study and identification of topologically ordered phases with infinite PEPS (iPEPS) [38] is indeed feasible even without imposing the virtual symmetry on the tensors during the optimization. As an example we consider the toric code model with an external magnetic field, where the tensors are known to exhibit a $Z_2$ virtual symmetry in the topologically ordered phase. We show that the resulting tensors (after a suitable gauge
The resulting state exhibits the relevant features of the topologically ordered state, including the correct topological entanglement entropy and modular $S$ and $U$ matrices characterizing the mutual and self-statistics of the emergent anyonic excitations.

Following ideas from Ref. [39], the modular matrices are obtained from the computation of wave-function overlaps of a complete set of ground states on a torus with minimum entanglement entropy, where each state has a well-defined anyonic flux through the torus. This approach has already successfully been applied based on matrix product states on cylinders [10], and also with iPEPS using the tensor renormalization-group (TRG) method [36,40–43]. In this paper we introduce another scheme based on the corner transfer-matrix (CTM) renormalization-group method [44], which is more efficient than TRG to compute wave-function overlaps.

We note that during completion of this paper a different approach to study topological order based on iPEPS without imposing a virtual symmetry was presented in Ref. [37]. Instead of recovering the virtual symmetry of the local tensors, the approach in Ref. [37] is based on projectors onto the ground states with different anyonic fluxes represented by matrix product operators (MPOs), which are found by an optimization procedure.

A technical challenge when simulating the square lattice toric code model with iPEPS is that the Hamiltonian consists of four-body operators. While previous energy minimization algorithms [45,46] are in principle not restricted to nearest-neighbor models [47], their generalization to more complicated Hamiltonians is rather tedious and computationally expensive. In the present paper we have developed an alternative energy minimization algorithm which, besides simple nearest-neighbor terms, can treat more general Hamiltonian operators in a simpler and more efficient way. (We note that recently another scheme based on automatic differentiation [48] has been introduced with a similar computational cost.)

This paper is organized as follows. In Sec. II, the toric code model is introduced as an example of a system exhibiting a $Z_2$ topological order, together with basic notions and concepts including the topological entanglement entropy (TEE), minimum entropy state (MES), and modular matrices. Then, an introduction to iPEPS and the standard CTM algorithm is given in Secs. III A and III B. In Sec. III C we explain the gradient-based optimization scheme developed in this paper. In Sec. IV, we present our approach to identify a topologically ordered phase with iPEPS, with the $Z_2$ topological order as an example. This includes the scheme to recover the $Z_2$ virtual symmetry in the tensors explained in Sec. IV C, and the extension of the CTM method to compute wave-function overlaps of the MESs to determine the modular matrices in Sec. IV E. In Sec. V we present results for the toric code model obtained with our approach, and we end with our conclusions in Sec. VI.

II. TORIC CODE MODEL

In this paper we consider the square lattice toric code model [3] as a simple example of a system exhibiting a $Z_2$ topologically ordered ground state. The model consists of

\[ \hat{H}_{\text{TC}} = - \sum_v A_v - \sum_p B_p \]  

where $v$ denotes the vertices of the lattice and $p$ denotes the plaquettes. The operator $A_v = \prod_{i \in v} \sigma_i^z$ is given as a product of Pauli matrices $\sigma_i^z$ acting on spin $i$ adjacent to a vertex $v$, and similarly $B_p = \prod_{i \in p} \sigma_i^z$ is a product of Pauli matrices $\sigma_i^z$ acting on the spins on a plaquette $p$, as shown in Fig. 1.

The model is exactly solvable, where the solution can best be seen in the $\sigma_z$ basis where the spin-up configuration is associated with the presence of a line going through the spin. All the plaquette and vertex terms commute with each other, so the ground-state configurations are the states where all plaquette and vertex terms act in a single line. On a vertex this requires that the number of up spins and down spins is even, which in the line representation means that if a line enters a vertex it also has to leave the vertex. Because lines cannot end at a vertex, the ground state can only contain configurations of closed loops of lines of up spins. The plaquette term in this basis flips all the spins around the plaquette, which allows this term to locally connect states with different closed-loop configurations. The ground state, being an eigenstate of both terms simultaneously, therefore has to consist of an equally weighted superposition of all the possible closed-loop configurations.

The ground state exhibits a degeneracy which depends on the topology of the lattice. This is because the plaquette term can only transform a closed-loop configuration locally, but it is not able to remove a noncontractible loop which winds around periodic boundaries. For example, when considering the model on a cylinder geometry, a single loop going around the cylinder cannot be removed by the plaquette operator (see Fig. 2). This results in two different ground-state sectors which can be labeled by the parity of the number of loops winding around the cylinder. The ground-state manifold is spanned by $\{ |\Psi_e\rangle, |\Psi_o\rangle \}$ where $e$ and $o$ denote the even and odd sectors, respectively. On a torus the ground-state degeneracy is fourfold, and the ground states can be labeled
FIG. 2. The ground state $|\Psi_g\rangle$ cannot be turned into $|\Psi_e\rangle$ under the action of the Hamiltonian, resulting in two distinct degenerate ground states. The two ground states can be distinguished by the non-trivial action of the Hamiltonian, resulting in two distinct degenerate ground states. The two ground states can be distinguished by the non-trivial action of the Hamiltonian, resulting in two distinct degenerate ground states.

as $\{|\Psi_{ee}\rangle, |\Psi_{eo}\rangle, |\Psi_{oe}\rangle, |\Psi_{oo}\rangle\}$ according to the parity in the horizontal and vertical direction.

Besides the ground state, also the elementary anyonic excitations in the toric code model are well understood. They correspond to a violation of a vertex term (electric excitation) or a plaquette term (magnetic excitation) and they always occur in pairs. Acting with a $\sigma^x_i$ operator on the ground state violates both vertex terms including site $i$, i.e., it creates two electric ($e$) particles on the adjacent vertices. More generally, the path operator $W^{(\gamma)}(\gamma) = \prod_{i\in\gamma} \sigma^x_i$ where $\gamma$ is an open path on the lattice creates two electric particles at its end points. Similarly, a $\sigma^y_i$ operator on the ground state violates both plaquette terms including site $i$, corresponding to a creation of two magnetic particles located on the two adjacent plaquettes, and the path operator $W^{(\bar{\gamma})}(\bar{\gamma}) = \prod_{i\in\bar{\gamma}} \sigma^y_i$ where $\bar{\gamma}$ is an open path on the dual lattice creates a pair of particles at the end points of this path. While both excitations have trivial (bosonic) self-statistics, they exhibit nontrivial mutual statistics since a phase factor of $-1$ is acquired upon braiding an $e$ and an $m$ particle. One can further identify the composite particle $e$ corresponding to a combination of an $e$ and $m$ particle, which has fermionic self-statistics, and the trivial identity particle $1$.

A. Topological entanglement entropy and minimal entanglement states

Topologically ordered states are known to exhibit a universal correction $\gamma$ to the area law of entanglement [49,50], $S_A(L) \sim a L - \gamma$, where $S_A(L)$ is the entanglement entropy between a disk-shaped region $A$ with a smooth boundary of length $L$ and the rest of the system [51]. The universal constant $\gamma$ is called the topological entanglement entropy and is equal to $\log D$, where $D = \sum_k \sqrt{d_k}$ is the total quantum dimension and $d_k$ denotes the quantum dimensions of the $k$th particle type of the underlying topological theory. For Abelian anyons, $d_k = 1/k$, and therefore $\gamma = \log 2$ for the toric code ground state [50,52]. A finite TEE is a characteristic feature of a topologically ordered phase and can be thus used to detect such a phase [53]. Special care must be taken, however, if the region $A$ has a nontrivial topology. If the boundary of $A$ is noncontractible, the value obtained for $\gamma$ depends on the specific ground state, and only the so-called minimum entropy states (MESs) yield the maximal, universal value for $\gamma$ [39].

Consider a torus cut into two cylindric (i.e., noncontractible) regions $A$ and $B$. The MESs correspond to ground states with a well-defined anyonic flux through the entanglement cut between $A$ and $B$ [39]. For the toric code the four different MESs for this bipartition can be identified with the fluxes of anyons $\{e, m, e, m\}$, corresponding to eigenstates of the loop operators $W^{(\gamma)}(C)$ and $W^{(\bar{\gamma})}(C)$ acting along the cut $C$ which detect the presence of an electric and magnetic flux, respectively. One can easily show [39] that with respect to a cut in the vertical direction the MESs are given by

$$|\Psi_{e/m}\rangle = \frac{1}{\sqrt{2}} (|\Psi_{ee}\rangle \pm |\Psi_{eo}\rangle),$$

$$|\Psi_{e/e}\rangle = \frac{1}{\sqrt{2}} (|\Psi_{oe}\rangle \pm |\Psi_{oo}\rangle).$$

Any of the MESs will yield the universal constant $\gamma = \log 2$ when computing the TEE of region $A$. We will show in Sec. IV E how we can obtain the MESs based on iPEPS.

B. Modular $S$ and $U$ matrices

Because the MESs describes states with different anyonic fluxes, they can be used to compute the modular $S$ and $U$ matrices which characterize the nontrivial braiding and self-statistics of the anyonic excitations, as shown in Ref. [39]. For Abelian anyons, the $S_{ij}$ matrix describes the phase a particle $i$ obtains when encircling particle $j$ (divided by the total quantum dimension). The $U_{ij}$ matrix, which is diagonal, describes the phase a particle $i$ obtains when exchanged with another particle of type $j$.

The $S$ matrix, which on a square geometry acts as a $\pi/2$ rotation on the MES basis, can be calculated as [2]

$$S_{ij} = \frac{1}{D} \langle \Psi_j^{\gamma} | \Psi_i^{\gamma} \rangle$$

where $D$ is the total quantum dimension, and the states $|\Psi_j^{\gamma}\rangle$ and $|\Psi_i^{\gamma}\rangle$ denote the MES with anyonic flux $j$ in the $x$ direction and anyonic flux $j$ in the perpendicular $y$ direction on the torus, respectively.

The $U$ matrix describes the action of a Dehn twist on the torus which can be viewed as cutting the torus along the $\hat{y}$ direction to create a cylinder, rotating one of the cuts by $2\pi r$, and gluing the cuts back together to get back a torus geometry. If an anyonic flux is going perpendicularly through the cut, this one gets wrapped around the $\hat{y}$ direction of the cut. Therefore, the $U$ matrix can be viewed as an operation which adds the flux present in the $\hat{x}$ direction to the flux along the $\hat{y}$ direction (see also Ref. [36]). Thus, on a state with parity $p_x$ and $p_y$ in the $x$ and $y$ direction, respectively, one obtains $U \hat{U} |\Psi_{p_x,p_y}\rangle = |\Psi_{p_x,p_y}\rangle$, and thus one finds for the MESs of the toric code

$$\hat{U} |\Psi_{e/m}\rangle = \frac{1}{\sqrt{2}} (|\Psi_{ee}\rangle \pm |\Psi_{eo}\rangle),$$

$$\hat{U} |\Psi_{e/e}\rangle = \frac{1}{\sqrt{2}} (|\Psi_{oe}\rangle \pm |\Psi_{oo}\rangle) = \pm |\Psi_{e/e}\rangle.$$
where $h_z$ and $h_x$ are the magnetic field strengths in the $z$ and $x$ direction, respectively, and we will set $J = 1/2$ in the following. This model is no longer exactly solvable, but has been studied in previous works by series expansion [54,55], Monte Carlo method [56], and iPEPS [55,57]. The topological phase extends to a finite value of the magnetic field where a second-order transition at $h_z = h_x = 0.170(1)$ [56].

III. iPEPS

A. iPEPS ansatz

An infinite projected entangled-pair state [18,19,38] is a tensor network ansatz which can systematically approximate ground states of two-dimensional lattice models in the thermodynamic limit. The ansatz exploits the area law of entanglement of gapped local Hamiltonians [60], which a PEPS reproduces by construction. On a square lattice, an iPEPS consists of a periodically repeated unit cell made up of tensors with five indices (legs), as shown in Fig. 3. Each tensor has a single physical leg representing the local Hilbert space of one or more lattice sites, and four auxiliary legs which connect to the neighboring tensors on a square lattice. The accuracy of the ansatz is systematically controlled by the dimension of the auxiliary indices called the bond dimension $D$.

In this paper, the lattice of the toric code model is mapped onto a tensor network with a unit cell consisting of two tensors $A_{ijklm}^i$ and $B_{ijklm}^i$, arranged in a checkerboard pattern [see Fig. 3(a)]. The $A$ ($B$) tensors represent the spins on the horizontal (vertical) bonds. Due to the symmetry of the lattice we choose the tensors to exhibit mirror symmetries, $A_{ijklm}^i = A_{kjmli}^i$ and $A_{ijklm}^i = A_{mklji}^i$, where the two tensors are related as $B_{ijklm}^i = A_{klmj}^i$.

The ground state of the toric code (without magnetic field) can be exactly represented with $D = 2$ tensors as defined in Fig. 3(b), where the $Q$ and $δ$ tensors are zero except for the elements

\[
Q_{111} = Q_{221} = Q_{212} = Q_{122} = 1, \tag{6}
\]

\[
δ_{11} = δ_{22} = 1. \tag{7}
\]

From the figure one can see that the two tensors $A$ and $B$ are related by a rotation of $90^\circ$ [61].

In the following we discuss the main ingredients of the iPEPS algorithm, including the contraction of the tensor network and the optimization of the tensors, i.e., finding the best variational parameters of the tensors which minimize the variational energy $E = \langle \Psi | \hat{H} | \Psi \rangle / \langle \Psi | \Psi \rangle$ where $|\Psi\rangle$ is the iPEPS wave function.

B. Corner transfer-matrix algorithm

To compute an expectation value of an observable with iPEPS, the corresponding operator is placed between the bra and ket layer and the resulting two-dimensional network is contracted. The contraction of the two-dimensional network cannot be done in an exact way but only approximately. In this paper we use the Corner Transfer Matrix renormalization-group (CTM) [44,62,63] algorithm to contract the network. The CTM algorithm approximates the entire lattice surrounding a bulk tensor by an environment, consisting of four corner tensors $C$ and four edge tensors $T$, as shown in Figs. 4(a)–4(c). The accuracy is systematically controlled by the boundary bond dimension $\chi$ of these tensors. In the present paper, due to the mirror symmetries of the bulk tensors, only a single $T$ tensor and two different $C$ tensors are needed. The latter two, which we label as $\{C_0, C_x\}$, differ by the orientation of the last absorbed bulk tensor, as defined in Fig. 4(e). Tensor $T$ is not mirror symmetric upon exchanging the boundary legs. Therefore, in order to keep track of its orientation, an oval in the shape of the last absorbed bulk tensor is added inside the depicted tensor in Fig. 4(e). All the boundary tensors are labeled by a superscript, which indicates the CTM iteration.

The initial boundary tensors $C_{x/\bar{y}}^{(0)}$, $T^{(0)}$ are constructed from the bulk $A$ and $B$ tensors by projecting the open boundary legs in the bra and ket layer onto a single state, e.g., onto the first element of each index [see Fig. 4(d)]. In each iteration of the CTM algorithm, two edge tensors and a bulk tensor are absorbed in the corner and a bulk tensor is absorbed in the edge tensor as shown in Figs. 4(f) and 4(g), thereby effectively growing the number of sites each boundary tensor represents. Both absorptions increase the boundary dimension from $\chi$ to $\chi \times D^2$, which is truncated back to the original boundary dimension $\chi$. The truncation is performed by two $\delta$ tensors, which are obtained from a singular value decomposition (SVD) of the corresponding corner tensors, shown in Fig. 4(f) [44]. The algorithm is run for multiple iterations until the singular values $s$ of the corner matrix...
FIG. 4. Tensor network diagrams describing the (symmetric) CTM algorithm used in this paper. (a) Infinite tensor network consisting of the checkerboard pattern of tensors \( a \) and \( b \), constructed by combining the iPEPS tensors \( A \) with \( A^\dagger \) and \( B \) with \( B^\dagger \), respectively, as shown in (c). The infinite tensor network surrounding a bulk tensor \( b \) is effectively represented by an environment consisting of four corner tensors \( C \times C \) and four edge tensors \( T \) as depicted in (b). The thin and thick lines have a dimension \( D^2 \) and \( \chi \), respectively. (d) The initial boundary tensors are constructed from the iPEPS bulk tensors by projecting the legs in both the bra and ket layer onto the first element, depicted by the black dots. (e) Definition of the two different corner tensors and orientation of the \( T \) tensor. (f), (g) Growth and renormalization step in the CTM algorithm (see text).

are converged. The total number of iterations needed we call \( N \).

Using the converged CTM environment tensors, the energy of the toric code model can be evaluated as shown in Fig. 5. We rewrite the Hamiltonian in Eq. (5) as \( \hat{H} = \sum_i \hat{H}_i + \sum_p \hat{H}_2 \), where \( \hat{H}_i \) and \( \hat{H}_2 \) correspond to the vertex and plaquette terms, respectively, where the on-site magnetic terms have been split evenly among both terms. Each of the two terms are then evaluated by making use of the corresponding CTM environment (see Fig. 5). We note that the two tensors \( \hat{H}_1 \) and \( \hat{H}_2 \) defined in Fig. 5 are never explicitly constructed since this would be computationally inefficient.

FIG. 5. The tensor network diagrams to evaluate the toric code Hamiltonian using the CTM environment tensors. The plaquette contribution is labeled as \( \hat{H}_1 \) and the vertex contribution \( \hat{H}_2 \).

C. Optimization

Recently, methods to optimize the iPEPS tensors based on an energy minimization have been introduced \([45,46]\). The goal is to find the optimal parameters in the tensors which minimize the variational energy,

\[
E = \frac{\langle \Psi | \hat{H} | \Psi \rangle}{\langle \Psi | \Psi \rangle},
\]

in order to obtain the best approximation to the exact ground state of a Hamiltonian \( \hat{H} \) for a given bond dimension \( D \). In Refs. \([45,46]\) different approaches have been proposed to calculate the derivative of Eq. (8) with respect to the tensors, which is then used either to perform a conjugate-gradient optimization \([46]\) or to solve a generalized eigenvalue problem \([45]\) to lower the energy in an iterative way. The derivative \( \partial_{\Omega_{\mathbf{k}}} (\langle \Psi | \hat{H} | \Psi \rangle) \) can be written as a double infinite sum where one sum goes over all Hamiltonian terms and the other sum goes over the locations of the “hole” created by taking the derivative with respect to \( \partial_{\Omega_{\mathbf{k}}} \) (note that the derivative of a tensor network with respect to a tensor \( X \) is given by the network with tensor \( X \) being removed). Due to translation invariance only the relative distances between the Hamiltonian terms and the holes matter, such that in practice only one of the two sums needs to be performed.

While in Refs. \([45,46]\) the summation was done over all Hamiltonian terms (with the hole kept fixed in the center), here we propose an alternative approach where we sum over all possible locations of the hole, with the Hamiltonian term(s) kept fixed in the center. This has the advantage that more complicated models, e.g., with longer-ranged interactions and/or multisite interactions (such as in the toric code model), can be treated more easily and computationally more efficiently, since no summation over these Hamiltonian terms is required, and the systematic summation over the hole position is model independent.
FIG. 6. Tensor network diagrams to represent the gradient of the energy with respect to tensor $A^\dagger$ (only the contributions from $\hat{H}$ are shown). (a) The derivative of the bulk and boundary tensors with respect to $\partial A^\dagger$ are represented by the green tensors, e.g., $a \otimes = \partial A^\dagger a$. The $\otimes$ symbols in the tensor shapes mark the different locations in the lattice at which the derivative is taken. In (b) the total derivative of $\langle \hat{H} \rangle$ is represented in terms of a summation over the different tensors introduced in (a). The first term can be directly evaluated and its contribution can be added to the total gradient $G$. The other contributions are rewritten in terms of the environments $\Gamma_{C_s}^{(N)}$, $\Gamma_{T}^{(N)}$, and $C^{(N)}_s$, $T^{(N)}_s$ which are evaluated recursively as shown in Fig. 7.

Following Ref. [46], we treat $A$ and $A^\dagger$ as independent and write the gradient as

$$\partial A \left[ \frac{\langle |\hat{H}| \psi \rangle}{\langle \psi | \psi \rangle} \right] = \partial A \left[ \frac{\langle |\hat{H}| \psi \rangle}{\langle \psi | \psi \rangle} \right] - \frac{\langle |\hat{H}| \psi \rangle}{\langle \psi | \psi \rangle^2} \partial A \langle |\psi | \psi \rangle.$$  \hspace{1cm} (9)

The expression is simplified by shifting the Hamiltonian $\hat{H} \rightarrow \hat{H} - \langle \Psi | \hat{H} | \Psi \rangle$, such that only the first term in Eq. (9) remains.

In the following we explain how to compute the gradient for the toric code model with two different four-site Hamiltonian terms $\hat{H}_1$ and $\hat{H}_2$, but we stress that the approach can also be easily applied to other types of Hamiltonians. The contributions to the gradient with respect to $\hat{H}_1$ are presented in Fig. 6(b): The location of the $\hat{H}_1$ operator is fixed in the center and the sum is taken over all possible locations of the hole which is located in either one of the corner tensors, one of the edge tensors, or one of the four sites in the center. Each green tensor in Fig. 6(b) denotes a sum over all possible hole locations on the sites that the tensor is effectively representing, as defined in Fig. 6(a), and we label the corresponding tensors with an $\otimes$ symbol. We note that the explicit construction of the green tensors is computationally inefficient. Instead, we use a recursive scheme based on the CTM method to sum up all contributions as described in the following.

We start by applying the CTM method (Sec. III B) until convergence is reached after $N$ iterations. All corner, edge, and isometry tensors at each CTM iteration are saved, and labeled by a superscript indicating the CTM iteration. Using the converged tensors at the final iteration $N$, the first term in Fig. 6(b) can directly be evaluated and added to the total gradient $\mathcal{G}$. The idea is now to construct the remaining terms of the summation recursively in order to recover all contributions to the gradient in a systematic way. We define the environment tensors $\Gamma_{C_s}^{(N)}$ and $\Gamma_{T}^{(N)}$ corresponding to the tensor
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Fig. 7. Recursive procedure to sum up all contributions to the gradient (only the contributions from $\hat{H}_1$ are shown). (a) The corner gradient tensor $C_{\text{on}}^{(i)}$ is expanded in its four contributions. The first term can be directly evaluated and its contribution added to the total gradient $\mathcal{G}$. The second term, which is taken with a prefactor 2 due to the presence of two $T$ tensors which are equivalent by symmetry, becomes a contribution $\Gamma_{\text{on}}^{(i-1)}$ to the edge environment for the next iteration, and the last term generates the corner environment tensors for the next iteration, $\Gamma_{\text{on}}^{(i-1)}C_{\text{on}}^{\times}$. Similarly, in (b) the edge tensor $\Gamma_{\text{on}}^{(i)}$ is expanded into two terms, where the first term can be directly evaluated and added to the total gradient $\mathcal{G}$, and the second term yields a contribution $\Gamma_{\text{on}}^{(i-1)}C_{\text{on}}^{\times}$ to the edge environment for the next iteration. (Note that in the second term the orientation of the boundary legs flips.) After each iteration the different $(i-1)$th edge environment contributions obtained in the corner and the edge update step (from both $\hat{H}_1$ and $\hat{H}_2$) are added to obtain the edge environment $\Gamma_{\text{on}}^{(i-1)}$ for the next iteration.

network surrounding the $C_{\text{on}}^{(N)}$ and $T_{\text{on}}^{(N)}$ tensors in Fig. 6(b), respectively. Starting from these environments at step $i = N$ we propagate backwards in the CTM steps and iteratively compute the respective environments at the $(i-1)$th step, and at the same time sum up all the contributions to the gradient, as described in Fig. 7. At each iteration, the $i$th corner (edge) gradient tensor contracted with its respective environment is rewritten in terms of the corner (edge) gradient tensor of the $(i-1)$th step plus a one-site gradient contribution coming from a bulk tensor which is added to the total gradient $\mathcal{G}$. A similar recursion is done (simultaneously) for the other Hamiltonian term $\hat{H}_2$, involving the environment tensor $\Gamma_{\text{on}}^{(i)}$ and corner gradient tensor $C_{\text{on}}^{(i)}$. The procedure is repeated iteratively until either all the $N$ CTM steps have been recovered or until the total gradient $\mathcal{G}$ is converged within a certain tolerance.

The total gradient can then be used in combination with a gradient-based minimization algorithm to optimize the tensors. In this paper we used the Broyden-Fletcher-Goldfarb-Shanno quasi-Newton method [64–67]. The method is converged when either the norm of the gradient becomes smaller than a certain tolerance or when it is no longer possible to find a suitable step size which would decrease the energy. In practice, in order to prevent convergence to a local minimum, we run the optimization starting from several random initial tensors and keep the state with the lowest variational energy.

This method has several advantages over the optimization schemes based on the summation of Hamiltonian terms [45,46]. First, the iterative part of the algorithm where the environments are updated scales in the same way as the normal CTM method, $O(\chi^3 D^4) + O(\chi^2 D^6)$. The dominant scaling of the algorithm lies in the calculation of the initial environments $\Gamma_{C_{\text{on}}/C_{\text{on}}}^{(N)}$, which, due to the four-body Hamiltonian, scales as $O(\chi^3 D^6)$, but it only needs to be done once per gradient computation. This is in contrast to previous approaches [45,46] involving the summation of four-body contributions at each CTM iteration which is computationally less efficient. Second, the current scheme can be very easily extended to other types of Hamiltonians, including longer-ranged and multisite interactions, by initializing the initial environments accordingly (i.e., without performing a summation of more complicated Hamiltonian terms).

We note that recently an alternative optimization method based on automatic differentiation (AD) has been introduced.
[48]. This method performs a similar back-propagation of environments as in our CTM approach in an automatized fashion. One difference is that the AD approach also includes the gradient contribution with respect to the SVD step in the CTM method in contrast to our approach. It would be interesting to compare the performance of the two approaches in detail, which we leave for future work.

IV. DETECTION OF A TOPOLOGICALLY ORDERED PHASE WITH iPEPS

A. Virtual symmetry

The advantage of using iPEPS to study (nonchiral) topologically ordered phases is that there exists a powerful framework on how these phases can be represented with this ansatz [22]. The distinguishing properties of a topologically ordered phase occur at a global level, however it has been shown that these properties can be translated to necessary symmetry requirements on the virtual indices of the iPEPS tensors, namely, that the tensors are invariant under the action of a certain symmetry group $G$ on the virtual indices of the tensors, which is called a virtual symmetry. More specifically, the tensor remains invariant under the simultaneous action of $U_g$ on all the virtual indices, where $U_g$ is a unitary representation of an element $g$ of the group $G$, as shown in Fig. 8(a). This idea has also been generalized to symmetries represented by matrix-product operators [28,29] obeying a similar pulling-through condition as shown in Fig. 8(b), i.e., pulling an MPO (or in our case a product of operators on a string) through a tensor leaves the tensor invariant.

Here we will focus on the $Z_2$ topologically ordered phase of the toric code model where the tensors are invariant under the simultaneous action of a unitary representation of the $Z_2$ symmetry group on the virtual legs. In the $D = 2$ case, we use the standard representation $U_g = \{I, \sigma^x\}$, with a straightforward generalization to larger $D$. To each state in the virtual space we can assign a parity label of even ($e$) or odd ($o$) and the symmetry condition with $U_g = \sigma^x$ implies that elements with a total odd parity in a tensor are vanishing, leading to a block structure of the tensors.

This block structure can be easily seen in the exact representation of the TC ground state, by identifying the first and second element of the virtual indices with the even and odd sectors, respectively. It naturally arises from the constraint of having closed loops in the ground state; i.e., whenever a loop (odd parity) enters a tensor, it has to exit it again, and all tensor elements corresponding to an odd total parity are zero.

B. Challenges in practical simulations

We have seen in the previous section that the iPEPS tensors representing the exact TC ground state exhibit a virtual $Z_2$ symmetry, i.e., a block structure where half of the tensor elements are zero due to this symmetry. However, in practice, when performing an optimization of the tensors for the toric code Hamiltonian starting from random initial tensors, the resulting tensors will in general not be $Z_2$ symmetric for two reasons. First, the symmetry may not be apparent due to the gauge freedom in the tensor network; i.e., between each bond two $D \times D$ size matrices $q$ and $q^{-1}$ can be inserted, acting as a basis transformation on the virtual level, which does not change the state. After the optimization the tensors are in an arbitrary gauge, i.e., not necessarily in the basis in which they are $Z_2$ symmetric.

The second reason is that the optimization may fail to yield perfectly symmetric tensors due to round-off and truncation errors and nonperfect convergence. Even if these errors are small they may be problematic since it has been shown that already small errors which violate the $Z_2$ virtual symmetry lead to a loss of topological order [34,35]. One way to overcome this problem is to enforce the virtual symmetry during the optimization [35,36], which, however, requires knowledge of the correct virtual symmetry beforehand. (If it is not known, one would need to run separate simulations, testing different virtual symmetries, which is not efficient since the optimization is computationally the most expensive part.)

In order to overcome these issues, in the following we present a scheme to recover the virtual symmetry starting from an unconstrained optimization. This allows us to perform an unbiased iPEPS optimization, i.e., without a priori imposing a virtual symmetry, and we will show that it is possible to correctly identify the topological order.

C. Restoring the virtual symmetry

In this section we present a scheme to restore the $Z_2$ virtual symmetry of the tensors obtained from an unconstrained optimization (i.e., where we do not impose the virtual symmetry during the optimization). We start by fixing the gauge freedom $qq^{-1}$ between all the tensors such that in the final basis the tensor is closest to a $Z_2$ symmetric tensor. Due to the mirror symmetries on our $A$ and $B$ tensors, $q$ reduces to a $D \times D$ unitary matrix. To fix the gauge we have to pick a basis in which we define the $Z_2$ symmetry. In practice, for a bond dimension $D$, we choose a basis where we associate the first $D/2$ entries of a leg with the even sector, and the rest with the odd sector. When $D$ is odd, we round the dimension up (down) to an integer for the even (odd) sector. The initial $q$ is taken as a random unitary matrix. Then, the optimal $q$ is found by minimizing the norm difference $\delta_Z$ as shown in Fig. 9, which measures the deviation of the tensor from a $Z_2$ symmetric one. If $\delta_Z$ is zero, it implies that the tensor fulfills the condition in Fig. 8(a), i.e., that it is perfectly $Z_2$ symmetric. The
\[
\delta Z_2 = \min_q \left\| \sigma_q \right\|^2
\]

FIG. 9. The PEPS tensor is brought into an approximate \(Z_2\) symmetric form by performing a gauge change using a unitary matrix \(q\) which minimizes \(\delta Z_2\).

optimization of \(q\) is done by a quasi-Newton minimization. Once convergence is reached, we transform the tensor using the \(q\) matrices, yielding a new tensor which is approximately \(Z_2\) symmetric.

After fixing the gauge, the tensor \(A\) (and \(B\)) exhibits a block structure, in which the elements lying outside the allowed \(Z_2\) symmetry blocks are small. By setting these small elements to zero, the tensor becomes fully \(Z_2\) symmetric, leading to a state with a robust topological order (if we consider a state within the topologically ordered phase). This procedure should not alter the state in any significant way, e.g., it should only lead to a very small difference in energy. We will show in the results section Sec. V that this is indeed the case.

D. Computing the TEE

We obtain the TEE from the second Rényi entropy \([68]\), 
\[
S_2(\rho_L) = -\log[\text{Tr}(\rho_L^2)],
\]

between two halves \(L\) and \(R\) of an infinite cylinder which can be efficiently computed with iPEPS based on ideas from Ref. \([69]\). It has been shown that there exists an exact mapping between the physical degrees of freedom of a region and the virtual degrees of freedom connecting to this region. Specifically, as shown in Ref. \([69]\), the reduced density matrix \(\rho_L\) can be represented as

\[
\rho_L = U \sqrt{\sigma_L^T \sigma_R} \sqrt{\sigma_L^T U^\dagger}
\]

where \(\sigma_L\) and \(\sigma_R\) are the left and right reduced density operators defined in the virtual space along the cut, respectively. That is, \(\sigma_L\) (\(\sigma_R\)) is obtained by contracting the double-layer tensor network in the region \(L\) (\(R\)), keeping the virtual indices at the boundary open. \(U\) is an isometry defining the mapping between the physical and virtual space. With Eq. (10) the second Rényi entropy can be written as

\[
S_2(\rho_L) = -\log \left[ \text{Tr}(\sigma_L^T \sigma_R \sigma_L^T \sigma_R) / N^2 \right],
\]

where we have introduced a normalization factor \(N = \text{Tr}[\rho_L] = \text{Tr}[\sigma_L \sigma_R^T]\) for the case that \(\rho_L\) is not normalized.

Using the CTM algorithm, \(\sigma_L\) (and \(\sigma_R\)) of an infinite cylinder of circumference \(L\) can simply be represented by a periodic chain of edge tensors \(T\) as shown in Fig. 10(a), since each edge tensor represents an infinite row of bulk tensors contracted with its complex conjugate. The trace in Eq. (11) can then be obtained by contracting the tensor network shown in Fig. 10(b), made of a \(4 \times L\) periodic network of \(T\) tensors (taking into account the orientation of the \(T\) tensors), and the normalization factor is represented in Fig. 10(c). A similar approach was also used in Refs. \([70,71]\). For large cylinders, it is beneficial to diagonalize the matrices represented by the two rows of edge tensors shown in Fig. 10(d), and then obtain the second Rényi entropy as

\[
S_2(L) = -\log[\text{Tr}(\Lambda^{L/2})]/[\text{Tr}(\Lambda^{L/2})]^2, \quad \text{where } \Lambda \text{ and } \Lambda' \text{ are the corresponding diagonal matrices.}
\]

The TEE entanglement entropy can then be determined from the intersection of a linear fit to \(S_2(L)\) for sufficiently large \(L\) with the \(y\) axis at \(L = 0\) (see Fig. 12 in Sec. V for an example). \(S_2(L)\) is computed from one of the MESs, which we obtain as explained in the following section.

E. Calculation of the modular \(S\) and \(U\) matrices with CTM

In Ref. \([36]\) an approach based on TRG was introduced for the calculation of wave-function overlaps to determine the modular \(S\) and \(U\) matrices. The idea is to preserve the virtual symmetry in the TRG coarse-graining process in both the bra and ket layer, such that the resulting coarse-grained tensor \(T\) representing the infinite 2D system on a torus exhibits the same virtual symmetry. From this tensor all the MESs can be obtained by acting with operators on the virtual legs, and overlaps between them can be efficiently computed.
FIG. 11. Variant of the CTM algorithm where the parity on the
bra and ket level on each boundary can be controlled. The red
lines are of dimension 2, with one state in the even (e) and odd
(o) sector, respectively, and the green triangles denote a projection
onto this space. A black dot corresponds to a projection onto the
even-parity sector. (a) Initialization of the boundary tensors, where
the boundaries of the corner tensor are projected onto the even-
even sector, and the red parity legs of the $T$ tensor are kept open.
(b) Coarse-graining step, in which the corner tensors are kept in
the even-even boundary sector, and the red parity legs of the $T$
remain open. (c) Contraction of the network on the left yields the
double-layer tensor $T$ representing an infinite plane (or infinite torus
when connecting the legs in a periodic way), where the parity in each
layer on each boundary can be controlled via the red legs.

Here we introduce an alternative scheme based on the
CTM, which is computationally more efficient than TRG. For
simplicity we discuss it here for the case of a $Z_2$ topologi-
cal order, and we consider bulk tensors with a $Z_2$ virtual
symmetry (e.g., after applying the symmetrization procedure
described in Sec. IV C). The main idea is to keep track of
the parity sectors on the open boundaries during the CTM
iterations, such that the different ground states can be indi-
vividually selected on the bra and ket level, similarly as in the
TRG approach [36].

The following adaptations are made to the standard CTM
algorithm (Sec. III B): First, instead of initializing the open
boundary of the edge tensor $T$ by a projection onto a single
state we keep two states, one with even and one with odd
parity, on both the bra and ket level, as shown in Fig. 11(b).
These extra legs are kept open in each CTM renormalization
step (Fig. 11). Second, we initialize the corner tensor to have
an even-even parity on the two open boundaries, and we keep
it in this sector in each renormalization step by projecting all
of the open parity legs of the absorbed $T$ tensors onto the
even-even sector [Fig. 11(b)]. The isometries to perform the
renormalization are found in a similar way as in the standard
approach.

Once the CTM has converged we construct the network
in Fig. 11(d) representing the infinite 2D system. Since the
parity on the boundary of the corner tensors is fixed to the
even-even sector, the total parity on the boundaries of the bra
and ket level can be fully controlled by the extra legs of the
double-layer tensor $\tilde{T}$ representing the infinite plane.

FIG. 12. (a) Second Rényi entropy of the toric code model as
a function of width $L$ of an infinite cylinder, obtained from the
optimized tensors ($O$), combined with a gauge transformation to
bring the tensors into an approximate $Z_2$ symmetric form ($O + GT$),
and symmetrized to recover the $Z_2$ virtual symmetry ($O + GT + S$).
The TEE $\gamma$ is obtained from the intersection of a linear fit (dashed
lines) with the $y$ axis. At short distances the correct value of TEE
is obtained after performing the gauge transformation even without
symmetrizing the tensors. (b) The TEE $\gamma$ obtained from linear fits
to the data between $L$ and $L + 100$ on large cylinders. At large
distances, only the symmetrized tensors yield the correct value for $\gamma$.
V. Results

A. Toric code model without a magnetic field

We start by testing our approach for the toric code model without magnetic field ($h_i = h_c = 0$, with $J = 1/2$). The optimization is done for $D = 2$ and $\chi = 40$. Since there exists an exact $D = 2$ representation, we expect to be able to reproduce the ground state accurately. The CTM contraction is terminated when the difference in the spectrum of the subsequent corner tensors $\left|\|g^{(i+1)} - g^{(i)}\|\right| < 10^{-12}$. The minimization is done using the gradient-based energy minimization approach described in Sec. III C. At each iteration, the calculation of the gradient is completed either when all the CTM steps are recovered or when the total gradient has converged, $\left|\|\delta g^{(i)} - \delta g^{(i-1)}\|\right| < 10^{-12}$. The optimization is terminated when either the difference in energy between two iterations has become smaller than $10^{-14}$ or when it is no longer possible to find a new direction which lowers the energy. We run several independent optimizations with different random initial states and take the state with lowest variational energy. The optimization scheme yields an iPEPS with only a very small difference in energy compared to the exact result, $|E - E_{\text{exact}}| = 2.4 \times 10^{-9}$.

Next, we apply the schemes presented in Sec. IV to see whether we can detect the $Z_2$ topological order. We first determine the gauge in which the tensors are closest to a $Z_2$ symmetric form using the method described in Sec. IV C. We find that, after performing the gauge transformation, the tensor elements which lie outside the $Z_2$ virtual symmetry blocks are small. We quantify the deviation from a $Z_2$ symmetric tensor by $\delta g_i$ (see Fig. 9) divided by the norm of the tensor. Here we obtain $\delta g_i/\|A\|^2 = 1.6 \times 10^{-5}$, demonstrating that the tensors are very close to a $Z_2$ symmetric form, but they are not perfectly symmetric. After performing the symmetrization, i.e., setting the small elements which violate the $Z_2$ symmetry to zero (see Sec. IV C), we obtain a state which has essentially the same energy, $|E - E_{\text{exact}}| = 4.1 \times 10^{-10}$, showing that the symmetrization step has only a minor effect on the ground-state energy.

In Fig. 12(a) we present results for the second Rényi entropy as a function of the width $L$ of an infinite cylinder. To compute the TEE $\gamma$ we determine the intersection of a linear fit to the data with the $y$ axis. Using the tensors directly obtained from the optimization (labeled $O$), the TEE is not correctly reproduced, which is expected since the iPEPS is in an arbitrary gauge, i.e., it does not represent an MES. After performing the gauge transformation (labeled $O + GT$) we obtain the correct TEE with a deviation of only $1.7 \times 10^{-7}$ if we take the linear fit up to $L = 40$. However, when increasing $L$ the result for the TEE starts to deviate from the exact result [see Fig. 12(b)], because they are not perfectly $Z_2$ symmetric. This is consistent with previous observations [34] where a breakdown of the topological order was found by manually adding small perturbations to the tensors which violate the $Z_2$ virtual symmetry. In contrast, after symmetrization (labeled $O + GT + S$) the TEE is correctly reproduced even on very large cylinders [72].

Another difference between the $O + GT$ and the $O + GT + S$ simulations can be identified by comparing the eigenvalue spectrum of the transfer matrix, i.e., the eigenvalues $N$

| $E$ | $|E - E_{\text{exact}}|$ |
|-----|-----------------|
| $\gamma - \log 2$ | $1.1 \times 10^{-13}$ |

TABLE I. Summary of the results obtained for the toric code model (without magnetic field) for $D = 2$ (see main text for discussion).

shown in Fig. 10. For a $Z_2$ topologically ordered state we expect a twofold degeneracy [27]. While in the symmetrized case the degeneracy is accurate up to machine precision, without symmetrization the difference between the two leading eigenvalues is $2.8 \times 10^{-5}$, i.e., very close but not a perfect degeneracy, which causes the loss of topological order at long distances.

Last, we calculate the modular $S$ and $U$ matrices based on the procedure explained in Sec. IV E using the symmetrized tensors. We find values, presented in Table I, which are in perfect agreement with the exact results (see e.g., Ref [39]), i.e., the anyonic particles $\Id, \epsilon$, and $m$ have bosonic self-statistics, whereas the $e$ particle has fermionic self-statistics, and braiding an $e$ particle with an $m$ particle (or with an $\epsilon = em$ particle) yields a phase of $\pi$ (i.e., they are mutually semions).

B. Toric code model in a magnetic field

Having established that the correct features of the topologically ordered phase can be extracted in the unperturbed toric code model, we now test our approach for more challenging cases which are no longer exactly solvable. We first consider two examples with a magnetic field applied in the $z$ direction for values $h_i = [0.1, 0.18]$ which lie inside and outside the topologically ordered phase, respectively. [The location of the critical point is $h_i = 0.164 237(2)$ [56].] The simulations and analysis are done in a similar way as in the previous case, with the results summarized in Tables II and III, for $D = 2$ and $D = 3$, respectively.

In the topological phase ($h_i = 0.1$) we can make similar observations as at the exactly solvable point. We again obtain tensors which, after a suitable gauge change, are approximately $Z_2$ symmetric, and after symmetrizing them we are able to successfully extract the correct TEE and modular matrices with a very high accuracy. Interestingly, after the gauge change the deviation from a $Z_2$ symmetric tensor is larger for $D = 3$ than for $D = 2$, probably because there is more freedom in these tensors to add off-diagonal elements which do not affect the energy in a significant way. We note also that the change in energy from $D = 2$ to $D = 3$ is very
TABLE II. $D = 2, \chi = 40$ results for the toric code model with a magnetic field in the $z$ direction. The data in the left and right column correspond to a state in and outside of the topologically ordered phase, respectively. In the topologically ordered phase the same $S$ and $U$ matrices are obtained as in Table I, with an accuracy which is close to machine precision (see Fig. 13).

| $h_z$ | $E$ | $|E - E_{Z_2}|$ | $\gamma - \log 2$ | $\delta_{E_z}/||A||^2$ |
|-------|-----|----------------|------------------|------------------|
| 0.10  | $-1.01041$ | $1.7 \times 10^{-56}$ | $7.3 \times 10^{-15}$ | $6.3 \times 10^{-2}$ |
| 0.18  | $-1.04220$ | $1.5 \times 10^{-5}$ | $0.6931$ | $4.8 \times 10^{-2}$ |

small here. In Fig. 13 we show the error in the modular matrices with respect to the exact results as a function of CTM iteration, showing that after sufficiently many steps (i.e., for sufficiently large system sizes depending on the correlation length in the system) the $U$ and $S$ matrices are accurately reproduced.

Outside the topological phase ($h_z = 0.18$) we consistently find a vanishing TEE and a nondegenerate transfer matrix spectrum, which allows us to correctly identify the trivial phase. Since the ground state is no longer degenerate, the CTM approach only yields a state with a finite norm in the even-even sector, whereas the other sectors exhibit a vanishingly small norm, such that the (trivial) modular matrices cannot be computed in a meaningful way here. We further note that we do find a gauge in which the tensors are close to being $Z_2$ symmetric; however, the $Z_2$ virtual symmetry alone does not automatically imply a topologically ordered phase.

In Table IV we present data of two additional examples for $D = 3$ ($\chi = 80$) in and outside the topological phase along the self-dual line, $(h_z, h_z) = (h, h)$ with $h = [0.1, 0.18]$ for which we find similar results as in the above case. [The critical point is located at $h_z = 0.170(1)$ [56].]

C. Phase transition as a function of $h_z$

Finally we test the approach for magnetic fields close to the phase transition as a function of $h_z$ (with $h_z = 0$) to locate the critical point. In Fig. 14(a) we present the $D = 2$ results for the TEE $\gamma$ [obtained from linear fits to $S_2(L)$ with $L$ between 200 and 400], which exhibits a clear jump from $\log(2)$ down to zero at $h_z = 0.1675(5)$, which is close but not equal to the Monte Carlo result $h_z = 0.164237(2)$ [56] due to finite $D$ effects. In Fig. 14(a) we also show the lowest few eigenvalues of the transfer-matrix spectrum $N$ where we clearly find a degeneracy within the topologically ordered phase up to the $D = 2$ critical point as expected. At the phase transition we consistently find a peak in the correlation length $\xi$.

TABLE III. Same as in Table II, here for $D = 3, \chi = 60$.

| $h_z$ | $E$ | $|E - E_{Z_2}|$ | $\gamma - \log 2$ | $\delta_{E_z}/||A||^2$ |
|-------|-----|----------------|------------------|------------------|
| 0.10  | $-1.01042$ | $1.7 \times 10^{-56}$ | $7.3 \times 10^{-15}$ | $6.3 \times 10^{-2}$ |
| 0.18  | $-1.04220$ | $1.5 \times 10^{-5}$ | $0.6931$ | $4.8 \times 10^{-2}$ |

[Fig. 14(b)], computed from the two leading (nondegenerate) eigenvalues of the transfer matrix of the gauge-transformed state. With increasing bond dimension the location of the phase transition approaches the Monte Carlo result, as shown in Figs. 14(c) and 14(d) with a transition value $h_z = 0.1665(5)$ for $D = 3$.

These results demonstrate that our approach is applicable also close to a phase transition (with an accuracy on the location of the critical point depending on $D$ as in conventional phase transitions [73,74]).

VI. SUMMARY AND DISCUSSION

In this paper we have demonstrated that it is possible to correctly identify a topologically ordered phase using unbiased iPEPS simulations, i.e., where we start the optimization from random initial tensors without imposing the corresponding virtual symmetry on the tensors. As an example we considered the toric code model in a magnetic field where, within the topologically ordered phase, the tensors should exhibit a virtual $Z_2$ symmetry. We found that, after a suitable gauge change, the resulting tensors are approximately $Z_2$ symmetric, and they can be fully symmetrized \textit{a posteriori} to generate a stable topologically ordered state, exhibiting the correct topological entanglement entropy and modular $S$ and $U$ matrices.

What are the implications of our findings? So far, a common belief was that the virtual symmetry needs to be
FIG. 14. (a) The TEE $\gamma$ and the dominant eigenvalues of the transfer matrix $\mathcal{N}$ across the phase transition as a function of the magnetic field $h_z$, obtained for $D = 2$ ($\chi = 30$). The dashed line shows the Monte Carlo result for the location of the phase transition [56]. (b) Correlation length $\xi$ as a function of $h_z$ across the phase transition for different values of the boundary dimension $\chi$, exhibiting a peak at the $D = 2$ critical point. (c), (d) Same as in (a) and (b) for bond dimension $D = 3$ ($\chi = 60$ in (a)).

imposed on the tensors in order to obtain and identify the correct phase, which is not a problem if the type of topologically order is known beforehand. However, if it is not known one would need to run a separate simulation for each possible virtual symmetry (each corresponding to another topologically ordered phase), which is not efficient since the optimization of the tensors is the computationally most expensive part. In our approach, in contrast, the idea is to perform a single unbiased simulation using unconstrained tensors, and determine the (approximate) virtual symmetry a posteriori, as a part of the analysis of the state. This is computationally cheaper and more in the spirit of unbiased tensor network calculations. We note that during completion of this paper a different approach to study topological phases based on unbiased iPEPS simulations was presented in Ref. [37].

In this paper we have also developed a variant of the CTM method where the parity on the boundary can be controlled, which is computationally more efficient than schemes based on TRG [36] to compute the modular $S$ and $U$ matrices in the topological phase. Furthermore, we have introduced a gradient-based energy minimization algorithm based on a summation of tensor environments, which is simpler and more efficient than approaches based on a summation of Hamiltonian terms [45,46], especially for models with interactions beyond nearest-neighbor sites and multisite interactions.

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[23] We note that there has also been interesting progress in the study of chiral topologically ordered states with PEPS recently [75, 76].
[51] We note that the value is independent of the Rényi entropy index.
[53] We note that the TEE does not uniquely identify a topologically ordered phase, since two different topologically ordered phases can have the same value for $\gamma$.

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