Quantum cluster algorithm for frustrated Ising models in a transverse field

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Working within the stochastic series expansion framework, we introduce and characterize a new quantum cluster algorithm for quantum Monte Carlo simulations of transverse field Ising models with frustrated Ising exchange interactions. As a demonstration of the capabilities of this new algorithm, we show that a relatively small, ferromagnetic next-nearest neighbour coupling drives the transverse field Ising antiferromagnet on the triangular lattice from an antiferromagnetic three-sublattice ordered state at low temperature to a ferrimagnetic three-sublattice ordered state.

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I. INTRODUCTION

The stochastic series expansion quantum Monte Carlo method [1], based on a stochastic sampling of terms in the high temperature expansion of a quantum many-body system, is among the most useful computational tools for the study of bosonic many-body systems. When the model being studied has a locally conserved charge (such as boson number or a component of the total spin), nonlocal directed loop updates are possible within this scheme [2, 3]. These lead to very efficient sampling of the high-temperature expansion and allow for a study of fairly large system sizes.

When there is no locally conserved charge, a different strategy needs to be used. This has been explored previously in the context of Ising models in a transverse field [4]. In this case, a quantum cluster algorithm has been formulated and tested in some examples [4]. Our motivation here is the following: In a broad class of frustrated Ising models in which the application of a transverse field leads to interesting physics, we have found that the clusters grown by this quantum cluster algorithm do not have “nice” statistical properties in the sense that they are either too small to be of much use, or too large, leading to updates that are nearly global spin-flips of the entire configuration. This leads to a very significant deterioration of ergodicity, and affects our ability to obtain reliable and accurate data at large sizes in interesting parameter regimes.

This motivates our introduction of a new quantum cluster algorithm for Quantum Monte Carlo simulations within the stochastic series expansion framework. In this article, we describe this new algorithm and demonstrate that it allows for an efficient study of such frustrated Ising antiferromagnets in a transverse field. In order to facilitate a detailed discussion of our method, we focus on a particular two-dimensional model system in the rest of this article, although the underlying idea has much wider applicability, and is not restricted to two dimensional systems. The particular model system we use to illustrate our scheme is the transverse field Ising antiferromagnet on the triangular lattice with additional further neighbour interactions.

In this model system, we provide two illustrations of the utility of our new method. First, we study the antiferromagnetic three-sublattice ordered low-temperature phase [5] of the transverse field Ising antiferromagnet on the triangular lattice, and demonstrate that our algorithm allows us to access key low temperature properties that are out of the reach of the standard [4] quantum cluster approach. Second, we use our method to demonstrate that a rather weak additional ferromagnetic coupling between next-nearest neighbours drives a phase transition to a ferrimagnetic three-sublattice ordered state at low temperature, and study the performance of our algorithm in this challenging regime of parameter space in which the nature of the ground state depends crucially on the balance between quantum fluctuations, which favour an antiferromagnetic three-sublattice ordering pattern, and ferromagnetic next-nearest neighbour interactions, which favour a ferrimagnetic three-sublattice ordering pattern.

The rest of this paper is organized as follows: In Section II, we provide an introduction to the physics of the transverse field Ising antiferromagnet on the triangular lattice, and summarize our understanding of the phase diagram of this model system. In Section III, we review the standard quantum cluster algorithm [4] and then describe our new algorithm. In Section IV, we first introduce the physical observables of interest to us, as well as the measures we use to characterize the performance of our algorithm. We then use these measures to characterize the performance of our algorithm. Additionally, we display our results for the low temperature transition between antiferromagnetic and ferrimagnetic three-sublattice order in this model system. We close with a brief discussion of connections to previous work and possible extensions in Section V.

II. MODEL

The transverse field Ising antiferromagnet on the triangular lattice, with Hamiltonian

\[ H_{\text{TFIM}} = J_1 \sum_{\langle rr' \rangle} \sigma_r^x \sigma_{r'}^x - \Gamma \sum_{r} \sigma_r^z, \]

where the sum over \( \langle rr' \rangle \) runs over links of the triangular lattice, has been studied earlier as a simple, paradigmatic
example of the interplay between quantum fluctuations and frustrating antiferromagnetic exchange couplings [5–7]. It provides a simple example of “order-by-disorder” effects in the following sense: when $\Gamma = 0$, the classical Ising antiferromagnet has a macroscopic degeneracy of minimum exchange-energy configurations that give rise to a non-zero entropy-density in the $T \to 0$ limit [8,9]. In this classical limit, the Ising spins remain disordered all the way down to $T = 0$ [8,10], albeit with a diverging correlation length, providing a simple example of a classical spin liquid. At $T = 0$, the spins have power-law correlations at the three-sublattice wave vector [10]. The transverse field $\Gamma$ induces quantum mechanical transitions that flip the individual $\sigma^z$. In spite of inducing these quantum fluctuations, a small transverse field immediately stabilizes long-range three-sublattice order of $\sigma^z$ [5,7].

This long-range order is of the antiferromagnetic variety (with no accompanying net magnetization) [5], and should be contrasted with the ferrimagnetic three-sublattice order exhibited at low temperature by the classical Ising antiferromagnet with ferromagnetic next-nearest neighbour couplings [11]. The distinction between the two kinds of three-sublattice order is best summarized by the following caricature. When the ordering is of the antiferromagnetic variety, the system spontaneously chooses one sublattice (out of the three sublattices corresponding to the natural tripartite decomposition of the triangular lattice) on which the spins are polarized along the $+\hat{z}$ direction in response to the transverse field. From the other two sublattices, it spontaneously chooses one sublattice on which the spins become polarized along the $+\hat{z}$ direction, while the spins on the third sublattice all point in the $-\hat{z}$ direction in spin space. On the other hand, when the ordering is of the ferrimagnetic variety, the system spontaneously chooses one sublattice on which the spins all point along the $+\hat{z}$ direction ($-\hat{z}$ direction), while the spins on the other two sublattices all point along the $-\hat{z}$ direction ($+\hat{z}$ direction).

At low temperature in the presence of an additional next-nearest neighbour ferromagnetic exchange coupling $J_2$ between the $\sigma^z$, one thus expects antiferromagnetic three-sublattice order for small values of $J_2/\Gamma$ [5], and ferrimagnetic three-sublattice order for large values of $J_2/\Gamma$ [11]. The vicinity of this low temperature transition is expected to be a challenging regime for any algorithm, since the ultimate fate of the system in the vicinity of this transition depends sensitively on the competition between quantum fluctuations induced by $\Gamma$ and the energetic preferences introduced by the presence of a nonzero $J_2$. Nevertheless, we find that our scheme does not suffer from any significant deterioration in ergodicity or efficiency in this regime. As a result, we are able to use it to pinpoint the value of $J_2$ at which the antiferromagnetic three-sublattice order gives way to ferrimagnetic three-sublattice order when we fix $T = 0.1$ and $\Gamma = 0.8$ in units of $J_1$.

### III. ALGORITHM

We begin with a brief review of the conventional quantum cluster algorithm for transverse field Ising models [4], focusing attention on models with a nearest-neighbour Ising exchange interaction. In this approach, the Hamiltonian is written as a sum of transverse field terms living on sites, and Ising exchange interactions living on links of the lattice:

$$H_{\text{TFIM}} = - \sum_{a,i} H_{a,i}$$

$$H_{1,i} = |J_1| - J_1 \sigma^z_{1(i)} \sigma^z_{2(i)}$$

$$H_{2,i} = \Gamma \mathbf{1}_i$$

$$H_{3,i} = \Gamma \sigma^z_i$$

In the above, the sum over $a$ ranges from $a = 1$ to $a = 3$. When $a = 1$, the sum over $i$ corresponds to a sum over links of the lattice. On the other hand, when $a = 2$ or $a = 3$, the sum over $i$ corresponds to a sum over sites of the lattice. The composite labels $1(i)$ and $2(i)$ refer to the two sites connected by link $i$, and $\mathbf{1}_i$ denotes the identity operator acting in the two-dimensional Hilbert space of site $i$. These constant terms $H_{2,i}$ are added to the Hamiltonian to facilitate the quantum cluster updates used in this approach. Additionally, $H_{1,i}$, which is diagonal in the basis of eigenstates of $\sigma^z$, is shifted by a constant in order to make all matrix elements positive, which is essential for sign-problem-free Monte Carlo sampling. The operators $H_{a,i}$ acting on any basis state $|\alpha \rangle$ consisting of a tensor product of eigenstates of $\sigma^z$ at various sites $i$ give rise to a constant multiple of some other basis state $|\beta \rangle$, not a linear superposition of basis states:

$$H_{a,i} |\alpha \rangle \propto |\beta \rangle .$$

One now expands $Z = \text{Tr}[\exp(-\beta H)]$ in a power series in the inverse temperature $\beta$:

$$Z = \sum_{n=0}^{\infty} \beta^n n! \sum_{\alpha_k} \prod_{k=1}^{n} \langle \alpha_{k+1} | H_{a(k),i(k)} | \alpha_k \rangle ,$$

where the sum over operator strings $\alpha_n$ runs over all possible sequences of $n$ operators $H_{a(k),i(k)}$ and basis states $\alpha_k$ ($k = 1, 2 \ldots n$, with the convention that $|\alpha_{n+1} \rangle \equiv |\alpha_1\rangle$). The basic idea now is to perform a Monte Carlo sampling of this sum over operator strings, with the weight of each operator string set by the product of matrix elements, along with the factor of $\beta^n/n!$. For completeness, we detail below how this proceeds at an operational level in the conventional algorithm [4].

To change the value of $n$, one uses a diagonal update. To facilitate discussion of this, it is useful to rewrite each term in the expansion Eq. (4) using a fixed length representation for operator strings. To do this, one introduces a fixed upper cut-off $L$ on the length $n$ of operator strings allowed during the simulation, and converts each operator string $\alpha_n$ to a length $L$ object by introducing $L - n$
identity operators denoted by $H_{0,0}$ in all possible ways. There are thus $\binom{\mathcal{L}}{n}$ fixed length operator strings corresponding to a single operator string of length $n$. In the fixed-length representation, the analog of Eq. (4) is thus

$$Z = \frac{1}{\mathcal{L}!} \sum_{S_{\mathcal{L}}} \beta^n (\mathcal{L} - n)! \prod_{k=1}^{\mathcal{L}} \langle \alpha_{k+1} | H_{a(k),i(k)} | \alpha_k \rangle$$

with the convention that $|\alpha_{\mathcal{L}+1} \rangle = |\alpha_1 \rangle$. The diagonal update is now implemented as a sweep through the operator string $S_{\mathcal{L}}$, in which one attempts to replace each diagonal operator by an identity operator $H_{0,0}$ and vice versa. The acceptance probabilities are fixed by the requirement that they satisfy the detailed balance condition with respect to the weight (Eq. (5)) of each operator string:

$$P(H_{0,0} \rightarrow \text{diag. operator}) = \text{Min} \left( 1, \frac{\beta(\mathcal{N} + 2|J_1|N_{\text{links}})}{\mathcal{L} - n} \right),$$

$$P(\text{diag. operator} \rightarrow H_{0,0}) = \text{Min} \left( 1, \frac{\beta(\mathcal{N} + 2|J_1|N_{\text{links}})}{\mathcal{L} - n + 1} \right).$$

Here, $N$ is the number of lattice sites, and $N_{\text{links}}$ is the number of links in the lattice. On deciding to replace an identity operator by a diagonal operator, the type of diagonal operator is decided such that each $H_{2,i}$ is chosen with a probability $\Gamma/(\mathcal{N} + 2|J_1|N_{\text{links}})$, while each Ising exchange operator $H_{1,i}$ is chosen with a probability $2|J_1|/(\mathcal{N} + 2|J_1|N_{\text{links}})$. If a particular $H_{2,i}$ is chosen, but not allowed due to the matrix element corresponding to $H_{2,i}$ in Eq. (5) being 0, then no change is made at that step of the sweep through the operator string.

The other crucial ingredient is the quantum cluster update. This attempts large scale non-local changes in the operator string. To implement this, one maps the operator string and the states $|\alpha_k \rangle$ appearing in Eq. (5) to a linked vertex list representation. A vertex corresponds to the non-zero matrix element of a non-identity operator $H_{a,i}$ (i.e., $a \neq 0$), with legs denoting the spin states of relevant sites in the bra and the ket. The vertices that appear in the construction of clusters in the conventional quantum cluster algorithm are shown in Fig. 1 for $H_{\text{TFIM}}$ (Eq. (1)). The Ising-exchange operators correspond to vertices with four legs each, while the single-site operators correspond to vertices with two legs. Links connect a leg of a vertex to corresponding legs of the next or previous vertex (operator) which acts on that site. In this representation, the conventional quantum cluster update proceeds as follows: One starts the cluster construction by choosing a random leg in the operator string. This is the initial entrance leg. If an entrance leg belongs to an Ising-exchange vertex, all the four legs belonging to this vertex are added to the cluster. If an entrance leg belongs to a two-leg vertex corresponding to a single-site operator, only that entrance leg is added to the cluster. The cluster building process then proceeds by following the links of all legs added to the cluster (as shown by the outgoing arrows in Fig. 2). The legs reached by following these links are added to a stack to be processed by the cluster building routine. After this, one repeatedly takes a new leg from the stack and checks if it is already part of the cluster being built. If not, this leg is used as an entrance leg and the same rules are followed. The process ends when there are no legs left on the stack. Once a cluster has been made, it can be flipped by flipping the spin state of all legs that are part of the cluster. Thus, a cluster flip effects operator substitutions of the form $H_{2,i} \rightarrow H_{3,i}$ and $H_{3,i} \rightarrow H_{2,i}$. The constant operators $H_{2,i}$ have the same weight as $H_{3,i}$ in Eq. (6), so flipping a cluster does not change the weight of the operator string. With this in mind, the conventional quantum cluster algorithm first decomposes the operator string into a set of non-overlapping clusters using this cluster-building prescription, and then flips each of these clusters with probability 1/2. One Monte Carlo step of this conventional quantum cluster algorithm is defined as a diagonal update followed by one quantum cluster update.

To appreciate the challenge faced by this conventional quantum cluster algorithm when dealing with a frustrated transverse field Ising model, we note that the low temperature physics in such magnets is dominated by the properties of configurations with minimum Ising exchange energy. In a frustrated Ising model, the number of such minimally frustrated configurations increases exponentially with the system size, corresponding to a non-zero $T = 0$ limit of the entropy density in the thermodynamic limit of the classical system. In the specific case of the triangular lattice antiferromagnet, these minimally frustrated configurations are characterized by the requirement that there is exactly one frustrated bond (linking parallel Ising spins) on each triangular plaquette of the lattice. For $H_{\text{TFIM}}$ on a triangular lattice,
the nature of the low-temperature order is determined by the interplay of thermal and quantum fluctuations acting within this subspace of minimally frustrated configurations \[3, 7\].

As we will see below, the conventional quantum cluster algorithm presented above fails to efficiently capture the effects of these fluctuations within the subspace of minimally frustrated configurations. To understand why, we note that the conventional cluster building procedure, being blind to the distinction between minimally frustrated triangular plaquettes, and fully frustrated plaquettes (with all Ising spins pointing in the same direction), either produces very small clusters or branches out along unfrustrated bonds of the lattice to produce a cluster that more often than not spans a very large fraction of the triangular lattice. The small clusters produced by the conventional algorithm accurately capture the short-distance physics of flippable spins, which have zero net Ising-exchange field acting on them, allowing them to flip freely in response to the transverse field \( \Gamma \). However, and this is key, the large clusters produced by the conventional algorithm do not capture the important effects of large-scale fluctuations within the minimally frustrated subspace, since these large-scale fluctuations do not comprise of simple spin-flips of large parts of the system.

In order to overcome these difficulties with the conventional quantum cluster algorithm, one needs a way of incorporating in our cluster construction procedure the distinction between minimally frustrated triangular plaquettes and fully frustrated triangular plaquettes [12, 13]. Following previous work on classical and quantum frustrated systems [14, 19], we achieve this by decomposing the antiferromagnetic nearest-neighbour Ising-exchange part of \( H_{\text{TFIM}} \) as a sum of operators acting on elementary triangular plaquettes. Although this can be done for any frustrated Ising antiferromagnet on a lattice with triangular plaquettes (such as the triangular, kagome, hyper-kagome or pyrochlore lattice), we focus the rest of our discussion on the triangular lattice case for concreteness. In this case, each triangular plaquette has one \( A \) sublattice site, one \( B \) sublattice site, and one \( C \) sublattice site (here, the labels \( A, B, \) and \( C \) correspond to the natural tripartite decomposition of the triangular lattice). Thus, the nearest-neighbour Ising-exchange terms are now written as

\[
H_{1,i} = \frac{3}{2} J_1 - \frac{J_1}{2} \left( \sigma_A(i) \sigma^z_B(i) + \sigma^z_A(i) \sigma_B(i) + \sigma^z_A(i) \sigma_C(i) + \sigma^z_A(i) \sigma^z_C(i) \right)
\]  

(7)

where \( J_1 > 0 \) is the nearest neighbour antiferromagnetic exchange coupling, \( i \) runs over the elementary triangular plaquettes, and the composite labels \( A(i), B(i), \) and \( C(i) \) denote the sites belonging to the respective sublattice in the triangular plaquette \( i \). The overall factor of half in front of the second term compensates for the fact that each link is shared by two plaquettes, and the first term represents a constant shift of energy designed to ensure that all matrix elements of \( H_{1,i} \) in the \( \sigma^z \) basis are positive. The other operators \( H_{2,i} \) and \( H_{3,i} \) (where \( i \) now labels sites of the lattice) in the decomposition of the Hamiltonian remain unchanged. The vertices corresponding to this decomposition are shown in Fig. 3.

The diagonal update proceeds exactly in the same way, with the new update probabilities given by

\[
P(H_{0,0} \rightarrow \text{diag. operator}) = \min \left( 1, \frac{\beta (N \Gamma + 2 J_1 N_\Delta)}{L - n} \right),
\]

\[
P(\text{diag. operator} \rightarrow H_{0,0}) = \min \left( 1, \frac{\beta (N \Gamma + 2 N_\Delta J_1)}{L - n + 1} \right),
\]

(8)

where \( N_\Delta \) denotes the number of elementary triangular plaquettes in the lattice. As before, once the decision to replace an identity operator with a diagonal operator is made, each diagonal operator \( H_{2,i} \) (where \( i \) is a site) is chosen with a probability \( \Gamma / (N \Gamma + 2 J_1 N_\Delta) \), while each Ising exchange operator \( H_{1,i} \) (where \( i \) is a triangular plaquette) is chosen with a probability \( 2 J_1 / (N \Gamma + 2 J_1 N_\Delta) \). If a particular \( H_{2,i} \) is chosen, but not allowed due to the matrix element corresponding to \( H_{2,i} \) in Eq. (5) being 0, then no change is made at that step of the sweep through the operator string.

To design a useful quantum cluster algorithm for such frustrated Ising antiferromagnets, we note that every Ising exchange vertex corresponds to a triangular plaquette on which two sites host a majority spin and one site hosts a minority spin. Thus each Ising-exchange vertex has two pairs of majority-spin legs and one pair of minority-spin legs. Flipping the spin state of one of the sites that hosts a majority spin corresponds to flipping the spin state of one pair of majority-spin legs. This gives another valid Ising-exchange vertex with the same weight. The new Ising-exchange vertex obtained in this way satisfies three key properties. First, this new Ising-exchange vertex does not correspond to a global spin-flip of the entire triangular plaquette. Second, the pair of majority-spin legs that were flipped are again majority-spin legs of the new Ising-exchange vertex. Third, the pair of majority-spin legs that were not flipped become minority-spin legs of the new vertex. The first property suggests that physically important large-scale fluctuations within the minimally-frustrated subspace are likely to be captured by a cluster algorithm which assigns a

![FIG. 3. Allowed vertices in the new quantum cluster algorithm: (a) Triangular plaquette vertices. (b) Single site vertices (constant operators and transverse field operators). Solid and open circles denote eigenstates of \( \sigma^z \) with eigenvalues +1 and −1 respectively.](image-url)
single pair of majority-spin legs of a vertex to one cluster, while assigning the other two pairs of legs of the same vertex to a different cluster. The second property suggests that it should be possible to ensure that such a cluster-construction protocol satisfies detailed balance so long as the subtlety introduced by the third property is properly accounted for.

We now build on this intuition to design our new quantum cluster algorithm, valid for any frustrated transverse-field Ising antiferromagnet on a lattice with triangular plaquettes. In each of the triangular plaquettes \( i \) of the lattice, we single out one site as being privileged and label it \( \mathcal{P}_i \). For a given lattice, there may be considerable freedom for the protocol to be followed in choosing these privileged sites. If there are \( M \) different natural choices for this protocol, one can either randomly pick one choice \( p \) out of these \( M \) at the start of each quantum cluster update step, and use it to obtain the privileged sites \( \{ \mathcal{P}_i^p \} \), or cycle through the various choices in some order. Using this choice for the privileged site of any triangular plaquette, we can now mark one pair of privileged legs and two pairs of ordinary legs in each Ising-exchange vertex that lives on this triangular plaquette. Thus all Ising-exchange vertices that live on a given triangular plaquette are decomposed in the same way into two privileged legs and four ordinary legs.

With these preliminaries out of the way, we are now in a position to specify the cluster construction procedure used in our new quantum cluster algorithm: If an entrance leg belongs to a two-leg vertex corresponding to a single-site operator, only that entrance leg is added to the cluster. Thus, for two-leg vertices, there is no change from the procedure followed in the conventional quantum cluster algorithm. However, if the entrance leg belongs to a six-leg Ising-exchange vertex, the cluster construction rule now depends on whether the privileged legs of this Ising-exchange vertex are majority-spin legs of this vertex. If the answer is yes, the rule is as follows: If the entrance leg is a privileged leg, only the two privileged legs of the vertex are added to the cluster, whereas, if the entrance leg is an ordinary leg, the four ordinary legs are added to the cluster. On the other hand, if the privileged legs of this vertex are not majority-spin legs, then all six legs of this vertex are added to the cluster regardless of the entrance leg.

Specializing this to the case of the triangular lattice Ising antiferromagnet, we note that there are three natural choices of protocol for choosing the privileged sites, corresponding to the natural tripartite decomposition of the triangular lattice into \( A \), \( B \) and \( C \) sublattices. Thus, the privileged site of each triangular plaquette can be consistently chosen to be the \( A \)-sublattice site, or the \( B \)-sublattice site, or the \( C \)-sublattice site, and the procedure is to randomly pick one out of these three choices at the start of each cluster update step, giving us either a \( A \)-update or a \( B \)-update or a \( C \)-update. The rules for the \( A \)-update are depicted pictorially in Fig. 4. In each update, we construct all possible clusters and then flip each cluster with a probability \( 1/2 \). As in the conventional quantum cluster algorithm, one Monte Carlo step can be defined as a set of diagonal updates and quantum cluster updates. In practice, we choose to define a Monte Carlo step as one cycle through a set of \( A \), \( B \), and \( C \) quantum cluster updates interspersed with diagonal updates.

Finally, we sketch the incorporation of additional exchange couplings, using the triangular lattice transverse-field Ising antiferromagnet as a paradigmatic example. Consider first the case with an additional ferromagnetic next-nearest neighbour coupling \( J_3 < 0 \). In this case, one has additional diagonal operators in the decomposition of the Hamiltonian:

\[
\begin{align*}
H_{\text{TFIM}} &= - \sum_{a,i} H_{a,i} \\
H_{1,i} &= \frac{3}{2} J_1 - \frac{J_1}{2} \left( \tilde{\sigma}_A^z(i) \tilde{\sigma}_B^z(i) + \tilde{\sigma}_B^z(i) \tilde{\sigma}_C^z(i) + \tilde{\sigma}_A^z(i) \tilde{\sigma}_C^z(i) \right) \\
H_{2,i} &= \Gamma \mathbf{1} \\
H_{3,i} &= \Gamma \tilde{\sigma}_i^z \\
H_{4,i} &= |J_2| - J_2 \sigma_1^z(i) \sigma_2^z(i)
\end{align*}
\]

In the above, the sum over \( a \) ranges from \( a = 1 \) to \( a = 4 \). When \( a = 1 \), the sum over \( i \) corresponds to a sum over triangular plaquettes of the lattice as discussed in the foregoing. When \( a = 2 \) or \( 3 \), the sum over \( i \) corresponds to a sum over sites of the lattice as before. Finally, when \( a = 4 \), the sum over \( i \) is over links of the lattice, since we use the original link-based decomposition for ferromagnetic interactions. The composite labels \( 1(i) \) and \( 2(i) \) in this case refer to the two sites connected by link \( i \). Thus, the operator string now has six-leg vertices corresponding to antiferromagnetic nearest-neighbour Ising-exchange couplings, four-leg vertices corresponding to ferromagnetic next-nearest-neighbour Ising-exchange couplings, and two-leg vertices that represent the single-site operators. The quantum cluster update now uses our new procedure for handling

![FIG. 4. Cluster construction rules for the ‘A’ update, shown for representative vertices. The ‘B’ and ‘C’ updates proceed in similar ways. Incoming and outgoing arrows denote entrance and exit legs respectively.](image)
six-leg vertices while continuing to use the original rules of the conventional quantum cluster update when dealing with four-leg vertices. The diagonal update also involves obvious changes, since there are now three kinds of diagonal operators.

On the other hand, if the next-nearest neighbour coupling is antiferromagnetic \((J_2 > 0)\), we use a triangular plaquette based decomposition of the new terms in the Hamiltonian, with the new triangular plaquettes being the elementary triangular plaquettes of the three triangular lattices formed from the \(A\), \(B\) and \(C\) sublattice sites of the original triangular lattice. Thus, the operator string now has two different kinds of six-leg vertices, apart from the two-leg vertices that represent single-site operators.

The modifications needed to account for the different kinds of diagonal vertices in the diagonal update are again straightforward, and not spelled out here. The only subtlety is that our cluster construction rules must now use two different notions of privileged sites when dealing with the two different kinds of six-leg vertices. For the six-leg vertices corresponding to the nearest neighbour antiferromagnetic interactions, we choose the privileged sites to be either the \(A\), \(B\) or \(C\) sublattice sites of the triangular lattice. For the six-leg vertices corresponding to next-nearest neighbour antiferromagnetic interactions, one can think of several different notions of privileged sites. The one we have tested most thoroughly in our work uses a three-sublattice decomposition of the three new triangular lattices to define the privileged sites for the next-nearest neighbour interaction vertices. This leads to \(3^3 \equiv 81\) different possibilities for the full cluster construction rules, each of which obeys detailed balance.

In our simulations, we define a Monte Carlo step as one cycle through a set of three randomly chosen quantum cluster updates interspersed with diagonal updates.

### IV. RESULTS

Our simulations use both the conventional and the new quantum cluster algorithm to study the transverse field Ising antiferromagnet on \(L \times L\) triangular lattices with periodic boundary conditions and \(L\) a multiple of six ranging from \(L = 36\) to \(L = 108\). To facilitate direct comparison of the two algorithms, we define one Monte Carlo step of the conventional quantum cluster algorithm in the same way as the new quantum cluster algorithm, i.e. as a set of three cluster updates interspersed with diagonal updates.

A transverse field \(\Gamma\) is known to induce long-range three-sublattice order in the Ising antiferromagnet on a triangular lattice [5]. This order melts in a two-step manner through an intermediate phase with power-law three-sublattice order [5]. Long-range three-sublattice order is known to persist up to the highest temperatures in the vicinity of an optimal value of the transverse field of around \(\Gamma \approx 0.8\) [5] in units of \(J_1\). Therefore, we choose \(J_1 = 1\) and a transverse field of \(\Gamma = 0.8\) in many of the simulations performed to compare the performance of the new algorithm with that of the conventional quantum cluster algorithm. Most of our results are deep in the ordered state, which extends to roughly a temperature of \(T \approx 0.2\) in units of \(J_1\) when \(\Gamma = 0.8\) [5].

We study the physics of three-sublattice ordering in such magnets by computing the the complex three-sublattice order parameter \(\psi\) and the kinetic energy density at the three-sublattice ordering wave vector \(\epsilon_Q\). Additionally, we also measure the uniform easy axis magnetization \(m\). These quantities are defined as

\[
m = \frac{1}{L^2} \sum \sigma^2 \tag{10}
\]

\[
\psi = \frac{1}{L^2} \sum \sigma^2 \exp(iQ \cdot \vec{r}) \tag{11}
\]

\[
\epsilon_Q = \frac{1}{L^2} \sum \sigma^2 \exp(iQ \cdot \vec{r}) \tag{12}
\]

where \(Q\) is the three-sublattice ordering wave vector and \(\vec{r}\) represents the coordinates of triangular lattice sites. The corresponding static susceptibilities are defined in the standard way

\[
\chi_0 = \frac{L^2}{\beta} \langle |\sum_0^\beta d\tau m(\tau)|^2 \rangle \tag{13}
\]

\[
\chi_Q = \frac{L^2}{\beta} \langle |\sum_0^\beta d\tau \psi(\tau)|^2 \rangle \tag{14}
\]

\[
\chi_{Qx} = \frac{L^2}{\beta} \langle |\sum_0^\beta d\tau \epsilon_Q(\tau)|^2 \rangle \tag{15}
\]

where \(O(\tau) \equiv e^{\tau H_{TFIM}} \mathcal{O} e^{-\tau H_{TFIM}}\) denotes the conventional imaginary-time version of the Heisenberg operator corresponding to any Schrodinger operator \(\mathcal{O}\). Monte Carlo estimators for these quantities, presented in Ref. [20], can be used without any changes with the new quantum cluster algorithm.

To test our implementation of the new quantum cluster algorithm, we have first obtained high precision results for a small system of linear size \(L = 3\) and compared these against the results of exact diagonalization. Results of such tests, for a simulation with \(2 \times 10^8\) Monte Carlo steps preceded by a warm-up of \(2 \times 10^8\) Monte Carlo steps, is shown in Fig. 5 for the total energy \(E\), and in Fig. 6 and Fig. 7 for \(\chi_Q\) and \(\chi_{Qx}\) respectively.

Since the physics of the three-sublattice ordered low temperature state is dominated by the effects of quantum and thermal fluctuations within the subspace of minimally frustrated classical configurations, this represents the most challenging regime for the Monte Carlo work. With this in mind, we focus on a comparison of the two algorithms deep in this ordered state. In Fig. 8, we first confirm that \(H_{TFIM}\) does indeed have long-range three-sublattice order at \(T = 0.1\) when \(J_1 = 1\) [5] and there is no next-nearest-neighbour coupling \(J_2\). In this ordered state, we characterize the performance of the two algorithms by measuring the autocorrelation function of the
Monte Carlo estimators of the physical observables defined in the foregoing.

For an observable $O$, with Monte Carlo estimator given by $\langle O(\tau) \rangle$, the autocorrelation function $A_O(\tau)$, for a run of $\tau_m$ Monte Carlo steps, is defined as

$$A_O(\tau) = \langle O(n + \tau)O(n) \rangle - \langle O(n) \rangle^2,$$

where the angular brackets denote averaging of $O(n)$ over the Monte Carlo configurations generated in the simulation. After obtaining this function, we normalize it so that $A_O(0) = 1$. The function $A_{\chi_Q}$, normalized in this way, is displayed in Fig. 9 and Fig. 10 for systems of linear size $L = 48$ and $L = 72$ respectively. The corresponding function $A_{\chi_Q}$, normalized in the same way, is displayed in Fig. 11 and Fig. 12 for the same values of $L$. All these autocorrelation measurements are performed during a Monte Carlo simulation of $2 \times 10^6$ Monte Carlo steps, after a warm up of $2 \times 10^5$ Monte Carlo steps.

We note that the auto-correlation function for both algorithms are not simple exponentials in the ordered state at $J_2 = 0$. Rather, they appear to have multiple time-scales. Further, it is clear that the new quantum cluster algorithm is significantly more efficient at decorrelating the Monte Carlo estimators of quantities studied, and this advantage appears to be roughly independent of size.
To investigate the mechanics behind this fairly dramatic improvement at low temperature, we take the number of vertex-legs $n_c$ in a cluster as a measure of cluster size, and look at the statistics of sizes of clusters constructed in the ordered state, as well as at higher temperatures. Our first observation is that both algorithms make two kinds of clusters, small and large. Small clusters have a size that is roughly independent of $n_{\text{tot}}$, the total number of legs in the operator string, while large clusters have a size that scales with $n_{\text{tot}}$. We find that the distribution of small clusters does not depend much on system size and temperature for either algorithm, being presumably set by the short-distance physics of flippable spins and clusters of spins, which have no net exchange field on them due to interactions with the rest of the system. When compared to the distribution of small clusters made by the conventional quantum cluster algorithm, the corresponding distribution for the new quantum cluster algorithm is nevertheless considerably broader, as is clear from Fig. 13. We ascribe part of the improved performance of the new quantum cluster algorithm to this difference in the distributions of small clusters.

The other reason for the dramatically better performance of the new quantum cluster algorithm is clear from a comparison of the distribution of large clusters made by the two algorithms in the ordered state. This comparison is shown in Fig. 14. Clearly, the conventional quantum cluster algorithm mostly makes clusters that, when flipped, amount to nearly a global spin flip of the entire operator string, while the new quantum cluster algorithm produces a fairly broad distribution of $n_c/n_{\text{tot}}$ which presumably plays a major role in improving the
FIG. 13. The distribution of small clusters built by the conventional quantum cluster algorithm and the new quantum cluster algorithm during simulations of the triangular lattice transverse field Ising antiferromagnet, displayed as a function of \( n_c \), the cluster size (number of legs that belong to a cluster).

FIG. 14. The distribution of large clusters built by the conventional quantum cluster algorithm and the new quantum cluster algorithm during simulations of the triangular lattice transverse field Ising antiferromagnet, displayed as a function of \( n_c \), the cluster size (number of legs that belong to a cluster) normalized by the total number of legs in the operator string. Inset zooms in on the extremely narrow peak that dominates this distribution for the conventional quantum cluster algorithm.

FIG. 15. System size dependence of the distribution of large clusters built by the new quantum cluster algorithm during simulations of the triangular lattice transverse field Ising antiferromagnet, displayed as a function of \( n_c/n_{tot} \), the cluster size (number of legs that belong to a cluster) normalized by the total number of legs in the operator string.

FIG. 16. Temperature dependence of the distribution of large clusters built by the new quantum cluster algorithm during simulations of the triangular lattice transverse field Ising antiferromagnet, displayed as a function of \( n_c/n_{tot} \), the cluster size (number of legs that belong to a cluster) normalized by the total number of legs in the operator string.

The performance of the algorithm. For our new algorithm, we also monitor the change in this distribution of large clusters as a function of system size (Fig. 15) and temperature (Fig. 16). As is clear from these figures, the distribution remains usefully broad even at larger sizes at low temperature, and narrows somewhat only on exiting the ordered state upon heating.

Histograms of the phase \( \theta \) of the Monte Carlo estimator for the complex order parameter \( \psi \) provide another fairly striking signature of the improved performance of the new quantum cluster algorithm. For instance, deep in the ordered state with \( J_2 = 0 \), the histogram of \( \theta \) produced by the new quantum cluster algorithm exhibits six well-defined and clearly separated peaks at the values \( (2m + 1)\pi/6 \) \( (m = 1, 2 \ldots 6) \), characteristic of antiferromagnetic three sublattice order [21]. Upon heating, the peaks get less pronounced, and finally disappear in the power-law ordered phase, providing a nice visual representation of the fact that the long-wavelength physics of the power-law ordered phase is expected to be controlled by a \( U(1) \) symmetric fixed line, with six-fold anisotropy in \( \theta \) expected to be an irrelevant perturbation of this fixed
antiferromagnetic three-sublattice order to ferrimagnetic three-sublattice order beyond some critical value of $J_2$ beyond which the energetic effects of this coupling dominate over the quantum fluctuations induced by $\Gamma$. This critical value will depend on the value of $\Gamma$ and $T$ (with $J_1$ fixed at $J_1 = 1$). In Fig. 19 we see that the system has three sublattice order when $J_2 = -0.1$, $J_1 = 1.0$ and $T = 0.2$. This order is of the ferrimagnetic kind. This is clear from the histogram of $\theta$ obtained using the new quantum cluster algorithm. This histogram, shown in Fig. 19, exhibits six well defined peaks at values $2m\pi/6$ ($m = 1.2 \ldots 6$) in the low temperature state, characteristic of ferrimagnetic three-sublattice order.

As in the antiferromagnetic phase, these peaks become less pronounced on heating and are finally replaced by a flat histogram, which presumably signals the onset of the the power-law ordered phase studied in Ref. 11 (we have not studied this transition in any detail here). As a further demonstration of the capabilities of the new quantum cluster algorithm, we use it to track changes in the histogram of order parameter phase $\theta$ as a function of $J_2$ at a fixed low temperature $T = 0.1$, and pinpoint the location of this transition from antiferromagnetic three-sublattice order to ferrimagnetic three-sublattice order. As is clear from the results shown in Fig. 19, we obtain in this manner an accurate estimate of $J_2c \approx -0.028(4)$ for the location of this transition when $\Gamma = 0.8$, $T = 0.1$, and $J_1 = 1$. 

V. DISCUSSION AND OUTLOOK

In the conventional quantum cluster algorithm, the cluster decomposition of a given operator string is completely deterministic, although each cluster is randomly
FIG. 20. Histograms of $\theta$, the phase of the estimator of the complex three-sublattice order parameter $\psi$, obtained using the new quantum cluster algorithm, allow us to locate the transition between antiferromagnetic and ferrimagnetic three-sublattice ordered states of the triangular lattice transverse field Ising antiferromagnet with $J_1 = 1$, $T = 0.1$, and $\Gamma = 0.8$.

Cluster algorithm can be viewed as a quantum generalization of the Swendsen-Wang cluster algorithm.

Thought of in this way, it is not surprising that it fails when the Ising exchange interactions are frustrated, since the Swendsen-Wang algorithm is known to perform poorly at low temperature in classical Ising models with frustrated interactions. Since our new quantum cluster algorithm solves this problem, a natural question that might be worth exploring in more detail is the classical limit of our new quantum cluster algorithm. Preliminary analysis suggests that this classical limit is closely related to the classical cluster algorithm proposed by Zhang and Yang [14] for the specific case of triangular lattice Ising model with nearest neighbour antiferromagnetic interactions, as well as the general plaquette algorithms explored by Coddington and Han [15]. It would be useful to explore this further in future work.

In Sec. III and Sec. IV, we have already studied extensions of our quantum cluster algorithm to systems in which some of the exchange couplings are frustrated, while others are unfrustrated, for instance, transverse-field Ising antiferromagnets on the triangular lattice with nearest-neighbour antiferromagnetic interactions and next-nearest neighbour ferromagnetic interactions. Additionally, we have also tested the extension to the case in which both nearest and next-nearest neighbour interactions are antiferromagnetic. Another, far more nontrivial, extension would involve generalizations of our quantum cluster approach to the simulation of models such as the quantum dimer model [24, 25], in which the Hilbert space consists of states that obey a set of local constraints. Preliminary analysis suggests that this may also be possible, and is therefore worth exploring.

VI. ACKNOWLEDGEMENTS

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