


Monte Carlo study of a generalized icosahedral model on the simple cubic lattice

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We study the critical behavior of a generalized icosahedral model on the simple cubic lattice. The field variable of the icosahedral model might take 1 of 12 vectors of unit length which are given by the normalized vertices of the icosahedron as value. Similar to the Blume-Capel model, where in addition to -1 and 1 , as in the Ising model, the spin might take the value 0 , we add in the generalized model $(0,0,0)$ as allowed value. There is a parameter D that controls the density of these voids. For a certain range of D , the model undergoes a second-order phase transition. On the critical line, $O(3)$ symmetry emerges. Furthermore, we demonstrate that within this range, similar to the Blume-Capel model on the simple cubic lattice, there is a value of D , where leading corrections to scaling vanish. We perform Monte Carlo simulations for lattices of a linear size up to $L = 400$ by using a hybrid of local Metropolis and cluster updates. The motivation to study this particular model is mainly of technical nature. Less memory and CPU time are needed than for a model with $O(3)$ symmetry at the microscopic level. As the result of a finite-size scaling analysis we obtain $\nu = 0.711\,64(10)$, $\eta = 0.037\,84(5)$, and $\omega = 0.759(2)$ for the critical exponents of the three-dimensional Heisenberg universality class. The estimate of the irrelevant renormalization group eigenvalue that is related with the breaking the $O(3)$ symmetry is $y_{\text{ico}} = -2.19(2)$.

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I. INTRODUCTION

In the neighborhood of a second-order phase transition, thermodynamic quantities diverge, following power laws. For example, the correlation length behaves as

$$\xi = a_{\pm}|t|^{-\nu} (1 + b_{\pm}|t|^{\theta} + ct + \dots), \quad (1)$$

where $t = (T - T_c)/T_c$ is the reduced temperature. The subscript \pm of the amplitudes a_{\pm} and b_{\pm} indicates the high- (+) and the low- (−) temperature phase, respectively. There are nonanalytic or confluent and analytic corrections. The leading ones are explicitly given in Eq. (1). In the literature, the exponents associated with the specific heat, the magnetization, and the magnetic susceptibility are denoted by α , β , and γ , respectively. The exponent of the magnetization at the critical temperature for a nonvanishing external field is denoted by δ . The exponent η governs the behavior of the two-point function at the critical point. For the precise definition of these exponents and relations between them see, for example, Sec. 1.3 of [1]. Second-order phase transitions are grouped into universality classes. For all transitions within such a class, critical exponents assume identical values. Also, correction exponents such as $\theta = \omega\nu$ are universal. Universality classes are characterized by the symmetry properties of the order parameter at criticality, the range of the interaction, and the spatial dimension of the system. For reviews on critical phenomena see, for example, [1–4].

Note that in general the symmetry properties of the order parameter can not be naively inferred from the microscopic properties of the system. In particular, a symmetry might emerge that is not present in the classical Hamiltonian. For example, in a binary mixture, the two components are not related by a \mathbb{Z}_2 symmetry. However, in case the

mixing-demixing transition is of second order, it belongs to the Ising universality class, which is characterized by a \mathbb{Z}_2 symmetry of the order parameter.

In this work we are aiming at a precise determination of the critical exponents of the three-dimensional Heisenberg universality class. To this end, we study a generalized icosahedral model on the simple cubic lattice. The field variable takes the normalized vertices of the icosahedron as values. In addition, $(0,0,0)$ might be assumed. In the following we refer to field variables assuming this value as voids. The idea to use a discrete subset of the sphere as values of the field variable is rather old [5,6], however, received little attention. Note that the field variable is also referred to as spin. The model has a parameter D that controls the density of voids. For a certain range of this parameter, the model undergoes a second-order phase transition. Our numerical data show that at the phase transition, the model is in the domain of attraction of the $O(3)$ -invariant fixed point. Hence, the model shares the universality class of the three-dimensional $O(3)$ -invariant Heisenberg model. On the critical line, similar to the Blume-Capel model, the amplitude of the leading correction depends on D . We demonstrate that there is one value D^* of D , where leading corrections to scaling vanish. We refer to the model at $D \approx D^*$ as improved model. Simulating an improved model allows us to get more accurate estimates of universal quantities at a given budget of CPU time. For a brief discussion and references on improved models in general see [7].

As discussed below in more detail, a perturbation with the symmetry properties of the icosahedron is irrelevant at the $O(3)$ -invariant fixed point. In Appendix B we determine the corresponding renormalization group (RG) eigenvalue $y_{\text{ico}} = -2.19(2)$. Likely, analogous to the case of the clock

TABLE I. We give a selection of theoretical results for the critical exponents ν and η and the exponent ω of the leading correction to scaling for the three-dimensional Heisenberg universality class obtained by various methods. For a more comprehensive summary see, for example, Table 23 of Ref. [1]. For the definition of the acronyms and a discussion, see the text.

Ref.	Method	Year	ν	η	ω
[12]	3D exp.	1998	0.7073(35)	0.0355(25)	0.782(13)
[12]	ϵ exp. 5l	1998	0.7045(55)	0.0375(45)	0.794(18)
[13]	ϵ exp. 5l	2017	0.7056(16)	0.0382(10)	0.797(7)
[13]	ϵ exp. 6l	2017	0.7059(20)	0.0378(5)	0.795(7)
[17]	CB	2016	0.7121(28)	0.03856(124)	
[18]	NRG	2020	0.7114(9)	0.0376(13)	0.769(11)
[8]	MC	2001	0.710(2)	0.0380(10)	
[9]	MC+HT	2002	0.7112(5)	0.0375(5)	
[9,10]	MC+HT	2002	0.7117(5)	0.0378(5)	
[10]	MC	2011	0.7116(10)	0.0378(3)	
[9], this work	MC+HT, ϕ^4	2020	0.7116(2)	0.0378(3)	
This work	MC, ϕ^4	2020	0.71164(25)	0.03782(10)	
This work	MC, icosahedral	2020	0.71164(10)	0.03784(5)	0.759(2)

model discussed in Ref. [7], the perturbation is dangerously irrelevant. Meaning that in the low-temperature phase, in the thermodynamic limit, the spontaneous magnetization might only assume one of the 12 directions that are preferred by the Hamiltonian. Note, however, that this does not affect the finite-size scaling study at the critical point that we perform here.

Our motivation to study this model is that simulations take less CPU time than for an $O(3)$ -invariant model as, for example, the ϕ^4 model simulated in Refs. [8–10] and less memory is needed to store the field variables. The idea of this work is similar to that of Ref. [7], where we studied the $(q+1)$ -state clock model. In addition to the q values with unit length, the value $(0,0)$ might be assumed by the field variable. In the case of the $(q+1)$ -state clock model, an $O(2)$ -invariant model can be approached by taking the limit $q \rightarrow \infty$. In contrast, here we are restricted to the Platonic solids.

The Heisenberg universality class describes the critical behavior of isotropic magnets, for instance, the Curie transition in isotropic ferromagnets such as Ni and EuO, and of antiferromagnets such as RbMnF_3 at the Néel transition point. A summary of experimental results for critical exponents is given in the Tables 24 and 25 of [1]. An example for a more recent experimental study is Ref. [11]. In Table 2 of [11] estimates for the critical exponents β , γ , δ , and α are presented for four different materials. To get an idea of the accuracy that is achieved, let us pick out two results for GdScGe: $\alpha = -0.134 \pm 0.005$ for the exponent of the specific heat and $\delta = 4.799 \pm 0.006$ for the critical exponent of the magnetization on the critical isotherm. Using scaling relations, these exponents can be converted to $\nu = 0.7113(17)$ and $\eta = 0.0347(11)$, which are the exponents given in Table I below.

The three-dimensional Heisenberg universality class has been studied by using various theoretical approaches. Well-established field-theoretic methods are the ϵ expansion and the perturbation theory in three dimensions fixed. In order to extract numerical estimates for critical exponents, various resummation schemes are discussed in the literature. As examples we give in Table I the estimates obtained in Ref. [12].

Recently, there has been progress in the ϵ expansion and the six-loop coefficient has been computed for the $O(N)$ -invariant ϕ^4 theory [13]. In Table I, we give the results of the resummation used in Ref. [13] based on the five- and six-loop ϵ expansion. The five- and six-loop estimates are consistent. Note, however, that for the five-loop resummation, the estimate of the error differs a lot between Refs. [12] and [13]. For a discussion of the resummation schemes used, we refer the reader to Refs. [12,13]. The ϵ expansion has been extended to seven loop [14]. However, no numerical estimates for critical exponents have been computed so far.

Great progress has been achieved recently by using the so-called conformal bootstrap (CB) method. In particular, in the case of the three-dimensional Ising universality class, the accuracy that has been reached for critical exponents clearly surpasses that of other theoretical methods. See Ref. [15] and references therein. Very recently, also highly accurate estimates were obtained for the XY universality class [16], surpassing the accuracy of results obtained by lattice methods. Still for the Heisenberg universality class [17], the estimates are less precise than those obtained by other methods.

Considerable progress has also been achieved by using the functional renormalization group method. In Ref. [18] the authors have computed the critical exponents ν , η , and the correction exponent ω for various values of N . In the Tables IV, V, VI, and VII of [18] the authors summarize their results and compare them with estimates obtained by other methods for $N = 1, 2, 3$, and 4, respectively. A good agreement with the results of the conformal bootstrap is found. The same holds for the comparison with estimates obtained by studying lattice models. In Table I we report the estimates obtained for $N = 3$.

Finally, we report results obtained for the $O(3)$ -invariant ϕ^4 model on the simple cubic lattice. Note that there exists a value λ^* of the coupling constant λ of this model such that the leading correction to scaling vanishes. In Ref. [8] a finite-size scaling analysis of Monte Carlo (MC) data was performed. In Ref. [9] both Monte Carlo simulations and the high-temperature (HT) series expansion were used. In particular, the analysis of the HT series by using integral approximants [19] is biased by using the estimates of the inverse critical

temperature and $\lambda^* = 4.6(4)$ obtained by Monte Carlo simulations. In Ref. [10] we mainly focused on the RG eigenvalues of anisotropic perturbations at the $O(N)$ -invariant fixed point. As a by-product, we get the revised estimate $\lambda^* = 5.2(4)$. The values quoted for Refs. [9,10] are obtained by inserting this value into Eqs. (13), (14), and (19) of Ref. [9]. Next, we report the results of Monte Carlo simulations that we discuss in Appendix A. The estimate of the inverse critical temperature and λ^* are used to bias the HT analysis of Ref. [9]. Finally, we report the results obtained from the finite-size scaling study of the generalized icosahedral model. By using a hybrid of local and cluster algorithms we simulated lattices of a linear size up to $L = 400$. It is virtually impossible to give a comprehensive summary of the vast literature on the subject. For a more extensive summary see, for example, Table 23 of Ref. [1].

We notice that our results for ν and η obtained for the generalized icosahedral model are fully consistent with those that were obtained for the ϕ^4 model on the simple cubic lattice. Our results are also consistent with but more precise than those of Refs. [17,18] obtained by using the conformal bootstrap method and the functional renormalization group method, respectively.

Comparing with the results obtained from the resummation of the ϵ expansion we see clear differences. Our result for ν is larger than that obtained in Ref. [13] by about three times the error that is quoted. The result for the correction exponent ω obtained in Ref. [13] is roughly by five times the error that is quoted larger than ours.

The outline of the paper is the following: In Sec. II we define the model and the observables that we measured. Furthermore, we summarize theoretical results on subleading corrections to scaling. In Sec. III we discuss the Monte Carlo algorithm used in the simulations and outline our approach to the analysis of the data. In Sec. IV we analyze the data and present the results for the fixed-point values of phenomenological couplings, inverse critical temperatures, the correction exponent ω , and the critical exponents ν and η . In Sec. V we conclude and give an outlook. In Appendix A we discuss our results for the three-component ϕ^4 model on the simple cubic lattice. Finally, in Appendix B we determine the RG exponent γ_{ico} related with the breaking of the $O(3)$ symmetry.

II. MODEL

We consider a simple cubic lattice. A site is given by $x = (x_0, x_1, x_2)$, where $x_i \in 0, 1, 2, \dots, L_i - 1$. In our simulations $L_0 = L_1 = L_2 = L$ throughout and periodic boundary conditions are imposed. The model is analogous to the $(q + 1)$ -state clock model discussed in Ref. [7]. In the case of the $(q + 1)$ -state clock model the spins \vec{s}_x take either values on the unit circle or assume the value $(0, 0)$. Here, the circle is replaced by the two-sphere. In particular, the spin \vec{s}_x might take one of the 13 values \vec{v}_m tabulated below:

$$(0, 0, 0), \quad z(0, \pm 1, \pm \phi), \quad z(\pm 1, \pm \phi, 0), \quad z(\pm \phi, 0, \pm 1), \quad (2)$$

where $\phi = \frac{1}{2}(1 + \sqrt{5})$ is the golden ratio and $z = 1/\sqrt{1 + \phi^2} = 1/\sqrt{2 + \phi}$. The 12 vectors with unit length are the normalized vertices of the icosahedron. See, for example,

Eq. (A.20) of Ref. [20]. An alternative choice is given in Eq. (A.9). In our simulation program the field variables are stored by using the label $m \in \{0, 1, 2, \dots, 12\}$, where $\vec{v}_0 = (0, 0, 0)$ and $m \in \{1, 2, \dots, 12\}$ are assigned to the vectors of unit length.

In the following we shall refer to the model as generalized icosahedral model. The reduced Hamiltonian is given by

$$\mathcal{H} = -\beta \sum_{\langle xy \rangle} \vec{s}_x \cdot \vec{s}_y - D \sum_x \vec{s}_x^2 - \vec{H} \sum_x \vec{s}_x, \quad (3)$$

where $\langle xy \rangle$ denotes a pair of nearest-neighbor sites on the simple cubic lattice. We introduce the weight factor

$$w(\vec{s}_x) = \delta_{0, \vec{s}_x^2} + \frac{1}{12} \delta_{1, \vec{s}_x^2} \quad (4)$$

that gives equal weight to $(0,0,0)$ and the collection of the 12 values with $|\vec{s}_x| = 1$. Now, the partition function can be written as

$$Z = \sum_{\{\vec{s}\}} \prod_x w(\vec{s}_x) \exp(-\mathcal{H}), \quad (5)$$

where $\{\vec{s}\}$ denotes a configuration of the field.

The reduced Hamiltonian (3) and the weight (4) are the same as for the $(q + 1)$ -clock model defined in Sec. II B of Ref. [7]. The two models only differ in the set of allowed values of the field variables. Note that in the limit $D \rightarrow \infty$ the value $(0,0,0)$ is completely suppressed. In the following, we consider a vanishing external field $\vec{H} = (0, 0, 0)$ throughout.

A. Quantities studied

The most important quantities are dimensionless quantities R_i that are also called phenomenological couplings. In particular, we study the ratio of partition functions Z_a/Z_p , where a denotes a system with antiperiodic boundary conditions in one of the directions and periodic ones in the remaining two directions, while p denotes a system with periodic boundary conditions in all directions. Furthermore, we study the second moment correlation length over the linear lattice size $\xi_{2\text{nd}}/L$, the Binder cumulant U_4 , and its generalization U_6 . The exponent of the correlation length is determined by studying the finite-size scaling behavior of the slopes of dimensionless quantities. The critical exponent η is obtained from the finite-size scaling behavior of the magnetic susceptibility χ . These quantities are defined for example in Sec. II B of Ref. [7]. In our analysis, the observables are needed as a function of the inverse temperature β for a neighborhood of the inverse critical temperature β_c . To this end, we simulate at β_s , which is a preliminary estimate of β_c and compute the coefficients of the Taylor expansion in $(\beta - \beta_s)$ up to third order.

B. Subleading corrections to scaling

Analyzing our data, we use prior information on subleading corrections to scaling. These corrections are due to $O(N)$ -invariant perturbations of the fixed point and perturbations that break the $O(N)$ invariance. Let us first discuss the former. In Sec. III A of Ref. [7] we conclude, based on the literature, that there should be only a small dependence of the irrelevant RG eigenvalues on N . Therefore, the discussion of Sec. III A

of Ref. [7] should apply to the present case $N = 3$ at least on a qualitative level. In particular, we regard the subleading correction exponent $\omega_2 = 1.78(11)$ that we assumed in Refs. [9,10] as an artifact of the scaling field method [21]. Instead, the most important subleading correction should be due to the breaking of the rotational symmetry by the simple cubic lattice. Following Ref. [22], the associated correction exponent is $\omega_{NR} \approx 2.02$.

Now, let us turn to the corrections caused by the breaking of the $O(3)$ invariance. A good starting point of the discussion is provided by Ref. [20]. In Sec. II, polynomials are constructed that are invariant under the action of the discrete symmetry groups related with the Platonic solids and belong to an irreducible representation of the $O(3)$ group. Hence, they have a well-defined $O(3)$ spin n . In Eq. (3) of Ref. [20], polynomials associated with the tetrahedron, the cube, and the icosahedron are given. These are associated with the spin $n = 3, 4$, and 6 . Note that the tetrahedron is self-dual, the octahedron is dual to the cube, and the dodecahedron is dual to the icosahedron. There are no further Platonic solids in three dimensions. Note that dual Platonic solids share the symmetry properties. Hence, using the dodecahedron instead of the icosahedron as approximation of the sphere should result in the same irrelevant RG exponent.

In the case of a two-dimensional system, as discussed in Ref. [20], these perturbations of the $O(3)$ symmetry are relevant. In particular, the icosahedral model undergoes a phase transition at a finite temperature, while the $O(3)$ -symmetric model is asymptotically free and hence no phase transition occurs at a finite temperature.

In Ref. [10] we determined the RG exponents $y_n = 1.7906(3)$, $0.9616(10)$, and $0.013(4)$ for $N = 3$ and three spatial dimensions for spin $n = 2, 3$, and 4 , respectively. Hence, for example, a cubical model could not be used to study the properties of the $O(3)$ -invariant fixed point since the perturbation is relevant. We could not find a result for $N = 3$ and $n = 6$ in the literature. However, it is interesting to note that the estimates of y_n for $n = 2, 3$, and 4 for $N = 3$ are well approximated by the average of the corresponding values for $N = 2$ and 4 . In Refs. [23,24] the estimates $y_6 = -2.509(7)$ and $-2.069(7)$ are given for $N = 2$ and 4 , respectively. Therefore, we would expect $y_6 \approx -2.29$ for $N = 3$. In Appendix B we find $y_6 = -\omega_{ico} = -2.19(2)$. For a discussion of Platonic solids related with stable fixed points in three dimensions, see Ref. [25].

There are also corrections that are not related to irrelevant scaling field, such as the analytic background of the magnetic susceptibility. Effectively, it behaves as a correction with the exponent $2 - \eta$. For a more comprehensive discussion of subleading corrections see Sec. III of [7].

III. SIMULATION ALGORITHM

The algorithm used is very similar to the one discussed in Sec. IV of Ref. [7]. We simulated the model by using a hybrid of local updates and cluster updates [26]. In the case of the cluster algorithm, we have implemented the single cluster algorithm [27] and the wall cluster algorithm [28].

A. Local Metropolis updates

In order to speed up the local updates, in Ref. [7] we tabulate the contribution to the Boltzmann factor by pairs

$$B(m, n) = \exp[\beta \vec{s}(m) \cdot \vec{s}(n)] \quad (6)$$

and its inverse $B^{-1}(m, n)$, where m and n are the labels of the values of the spins. In order to adapt the implementation of the local updates of Ref. [7] to the present case, we just had to plug in the scalar products $\vec{s}(m) \cdot \vec{s}(n)$ for the vectors given in Eq. (2).

Similar to Ref. [7] we have used two versions of the Metropolis update that differ in the choice of the proposal. In the first version we always propose $\vec{s}_x' = (0, 0, 0)$ if $|\vec{s}_x| = 1$ and, with equal probability, one of the 12 values with unit length if $\vec{s}_x = (0, 0, 0)$.

In the second version, the proposal does not depend on \vec{s}_x . With probability $1/2$ we propose $\vec{s}_x' = (0, 0, 0)$ and with probability $1/24$ one of the 12 values with unit length. The second choice is used in addition to the first one since we were not able to prove ergodicity for the first one.

B. Cluster algorithms

Using the cluster algorithm, a spin is potentially changed by a reflection at one of the 15 symmetry planes of the icosahedron. The reflection can be written as

$$\vec{s}' = \vec{s} - 2(\vec{r} \cdot \vec{s})\vec{r}, \quad (7)$$

where \vec{r} is a unit vector perpendicular to the symmetry plane. Being too lazy to search the literature, we computed the possible values of \vec{r} by using a simple Python program. First, we define for all pairs of vertices \vec{v}_i of the icosahedron a candidate

$$\vec{c}_{ij} = \frac{\vec{v}_i + \vec{v}_j}{|\vec{v}_i + \vec{v}_j|}. \quad (8)$$

Then, we checked that the candidate is indeed a reflection. Finally, we search for multiple identifications of the same reflection. The remaining results for \vec{r} can be grouped in five triples of vectors that are mutually orthogonal:

$$\begin{array}{ccc} (1, 0, 0) & (0, 1, 0) & (0, 0, 1) \\ (-a, 1/2, b) & (1/2, b, a) & (b, a, -1/2) \\ (a, 1/2, b) & (-1/2, b, a) & (b, -a, 1/2) \\ (1/2, b, -a) & (b, a, 1/2) & (a, -1/2, b) \\ (-b, a, 1/2) & (1/2, -b, a) & (a, 1/2, -b) \end{array},$$

where $a = \phi/2$, $b = \phi/2 - 1/2$, and $\phi = \frac{1}{2}(1 + \sqrt{5})$ is the golden ratio.

As usual, the cluster algorithm is characterized by the delete probability of a pair of nearest-neighbor sites [27]

$$p_d(x, y) = \min[1, \exp(-2\beta[\vec{r} \cdot \vec{s}_x][\vec{r} \cdot \vec{s}_y])]. \quad (9)$$

Below we shall refer to a pair of nearest-neighbor sites as link. A link $\langle xy \rangle$ is deleted with probability $p_d(x, y)$. Otherwise, it is frozen.

In the program, we computed all possible values of $p_d(x, y)$ before the simulation is started, and store the results in a $15 \times 13 \times 13$ array of double-precision floating point values. Different cluster algorithms are characterized by the way clusters are selected. In the Swendsen-Wang algorithm [26],

the whole lattice is decomposed into clusters of sites that are connected by frozen links. In the Swendsen-Wang algorithm, a cluster is flipped with probability $1/2$. Flipping means that for all sites within a cluster, the reflection, Eq. (7), is performed. In the case of the single cluster algorithm [27], one site of the lattice is randomly selected. Then, only the cluster that contains this site is constructed. This cluster is flipped with probability 1. In the wall cluster algorithm [28], instead of a single site, a plane perpendicular to one of the lattice axis is chosen. The position on this axis is randomly chosen. Then, all clusters that contain sites within this plane are constructed and flipped with probability one. The measurement of Z_a/Z_p is discussed in Appendix A 2 of Ref. [29].

C. Update cycle

The update steps discussed above are compounded into a complete update cycle. Below we give a piece of pseudo C-code that represents the cycle that is used in our simulations:

```
Metropolis_2();
for(k=0;k<3;k++)
{
  Metropolis_1();
  ir=5*rand();
  wall_cluster((k+1)%3,triples[ir][0]);
  wall_cluster((k+1)%3,triples[ir][1]);
  wall_cluster((k+1)%3,triples[ir][2]);
  Metropolis_1();
  for(j=0;j<L;j++) single_cluster();
  Metropolis_1();
  ir=5*rand();
  wall_cluster_measure(k%3,triples[ir][0]);
  wall_cluster_measure(k%3,triples[ir][1]);
  wall_cluster_measure(k%3,triples[ir][2]);
  measurements();
}
```

Here, `Metropolis_1()` and `Metropolis_2()` are sweeps, using the first and second types of the Metropolis update discussed in Sec. III A. The single cluster update is given by `single_cluster()`. For each call, the reflection \vec{r} and the site where the cluster is started are randomly selected with a uniform distribution. `wall_cluster(k%3,triples[ir][i])` is a wall cluster update. The first argument selects the spatial direction. The array `triples[ir][i]` determines which reflection \vec{r} is chosen for the cluster update. The first index `ir` selects the set of mutually orthogonal \vec{r} that is taken. Then, within such a set we run through all three \vec{r} . The wall cluster update is either called just for updating the configuration or, in the case of `wall_cluster_measure`, to perform a measurement of the ratio of partition functions Z_a/Z_p in addition.

Most of the simulations were performed by using the update cycle discussed above. Below we shall refer to this cycle as cycle A. At a certain stage of the simulation, we realized that for some of the quantities it is more efficient to measure more frequently. Therefore, we skipped the wall cluster updates without measurement and reduced the number of single cluster updates from L to $L/2$. Furthermore, one of

the `Metropolis_1()` sweeps is skipped. Below we shall refer to this cycle as cycle B.

We implemented the code in standard C and used the SIMD-oriented Fast Mersenne Twister algorithm [30] as random number generator. Since the program is essentially the same as the one used to simulate the $(q+1)$ -state clock model, the CPU times needed for the update of a single site, are identical to those quoted in Sec. IV C of Ref. [7]: our Metropolis update type one requires 1.2×10^{-8} s per site. In the case of the single cluster update about 3.8×10^{-8} s per site are needed. These timings refer to running the program on a single core of an Intel(R) Xeon(R) CPU E3-1225 v3. Compared with the simulation of the $O(3)$ -symmetric ϕ^4 model on the simple cubic lattice discussed below in Appendix A, we roughly gain a factor of 3.

D. General remarks on the analysis of the data

The quantities that we study follow a power law that is subject to corrections

$$A(L) = aL^u \left(1 + \sum_i c_i L^{-\epsilon_i} \right), \quad (10)$$

where L is the linear size of the lattice. By using Monte Carlo simulations, we obtain estimates of $A(L)$ that have statistical errors. Mostly we intend to determine the exponent u , which is either the RG exponent of the thermal scaling field $y_t = 1/\nu$ or $2 - \eta$ here. The amplitudes a and c_i are in general unknown. In the case of the correction exponents we have some prior knowledge. This is gained by theoretical considerations or the analysis of other numerical data, as discussed in Sec. II B above. We denote the correction exponents in Eq. (10) by ϵ_i since not all are related to a single irrelevant scaling field. The correction exponents of irrelevant scaling fields are given by irrelevant RG exponents $\omega_i = -y_i$. Performing least-square fits, we need *Ansätze* that contain only a few free parameters. Hence, the series of corrections in Eq. (10) have to be truncated. In our case, there is the leading correction with the exponent $\omega = 0.759(2)$ [see Eq. (21) below]. Extracting the critical exponents ν and η , we consider $D \approx D^*$ and on top of that improved observables that are constructed such that the leading correction is suppressed. Therefore, it is safe to ignore the leading correction. As discussed in Sec. II B there are a number of different corrections with $\epsilon_i \approx 2$. These are the analytic background of the magnetic susceptibility that effectively corresponds to $\epsilon_1 = 2 - \eta$, the violation of the rotational symmetry by the simple cubic lattice $\omega_{NR} \approx 2.02$, and $\omega_{ico} = 2.19(2)$ related to the breaking of the $O(3)$ symmetry. In the case of the slopes that are used to determine ν , there is also $\omega + 1/\nu \approx 2.164$. In principle, there is an infinite series of corrections with increasing correction exponents. Since we can deal only with a few free parameters in fits, the sequence has to be truncated at some stage. Even the different corrections with an exponent $\epsilon_i \approx 2$ have to be represented by a single or by two effective correction terms. Hence, in general the *Ansatz* will never perfectly represent the data. Therefore, in addition to the statistical error there is a systematic one that is caused by this imperfection. With increasing linear lattice size L , the magnitude of corrections decreases. If one would

consider the linear lattice sizes $L_{\min} \leq L \leq cL_{\min}$, where $c > 1$, then the estimate of the exponent u would converge with increasing L_{\min} , up to the statistical error, to the true answer. Of course, the CPU time that is available sets an upper limit to cL_{\min} . Since we would like to squeeze out most from the data we proceed in a different way, similar to most analyses in the literature, all data with $L \geq L_{\min}$ are taken into account. The quality of the fit is measured as usual by

$$\chi^2 = \sum_j [(f(x_j, \{p\}) - y_j)/\sigma_j]^2, \quad (11)$$

where f is the *Ansatz* and $\{p\}$ the parameters of the *Ansatz*. In our case, x_j are the linear lattice sizes, y_j the numerical estimates of the observable, and σ_j its statistical error. In some of the fits below we consider several observables jointly. In this case,

$$\chi^2 = rC^{-1}r^T, \quad (12)$$

where C is the covariance matrix and $r_j = y_j - f(x_j, \{p\})$. Note that now x_j refers to the linear lattice size and the type of the observable. We also perform joint fits for several values of D . Then, x_j also refers to D . A fit usually is regarded as acceptable if $\chi^2/\text{d.o.f.} \approx 1$, where d.o.f. is the number of degrees of freedom. Furthermore, we consider the goodness-of-fit $Q = \Gamma_{\text{inc}}^{\text{up}}(\text{d.o.f.}/2, \chi^2/2)$, where $\Gamma_{\text{inc}}^{\text{up}}$ is the regularized upper incomplete gamma function. For a Gaussian distribution of the numerical estimates y_j , Q gives the probability that, assuming that the *Ansatz* is correct, χ^2 is equal to or larger than the value that we find for our data.

Here, we are dealing with *Ansätze* that are only correct up to corrections that decay with a power of the linear lattice size L . As a result, taking into account the smallest L that we have simulated, $\chi^2/\text{d.o.f.}$ is large and Q very small. Increasing L_{\min} , typically $\chi^2/\text{d.o.f.}$ decreases and Q increases. In all cases discussed below, eventually acceptable values of $\chi^2/\text{d.o.f.}$ and Q are reached. In our plots below we give only estimates that correspond to $Q > 0.01$. Typically, Q rapidly increases going to slightly larger L_{\min} . For most of the estimates shown $Q > 0.1$. A large value of $\chi^2/\text{d.o.f.}$ or a small value of Q certainly indicates that the *Ansatz* that is used is not sufficient to describe the data. Unfortunately, however, an acceptable value of $\chi^2/\text{d.o.f.}$ or Q says little about the systematic error on the parameters such as the exponent u . In particular, the systematic error can be considerably larger than the statistical one that is provided by the fit. This can be seen explicitly, for example, in our data for the slopes of different phenomenological couplings. While the correction exponents are the same for different quantities, very likely the corresponding amplitudes are not. Hence, the systematic effect on, for example, the result for the exponent y_t is likely different for different quantities. And in fact we see differences in y_t obtained from different quantities that are clearly larger than the statistical error, despite the fact that Q is acceptable. This effect can also be easily seen by generating synthetic data according to a function g with given values of the parameters and then fitting by using the *Ansatz* f , where f is obtained from g by skipping correction terms.

In order to get some handle on the systematic error, we compare results obtained by the same *Ansatz* but different quantities or by different *Ansätze*, containing a different number of correction terms for the same quantity. The final analysis is performed graphically. We plot the estimate of, for example, y_t as a function of L_{\min} . The final result and its error is then chosen such that for all quantities or all *Ansätze* considered the estimate obtained by fitting is, including the respective statistical error, within the interval given by the final estimate plus or minus its error. This procedure is not fully automatized and subject to some judgment.

The least-square fits were performed by using the function `curve_fit` contained in the SCIPY library [31]. The function `curve_fit` acts as a wrapper to functions contained in the MINPACK library [32]. We checked the outcome of the fit by varying the initial values of the parameters. Furthermore, we performed fits both by using the Levenberg-Marquardt algorithm and the trust region reflective algorithm. In particular in the case of fits with many free parameters, the trust region reflective algorithm turns out to be more reliable than the Levenberg-Marquardt algorithm. Plots were generated by using the MATPLOTLIB library [33].

IV. SIMULATIONS

Our simulations were performed on various PCs and servers. The CPU times quoted below refer to a single core of an Intel(R) Xeon(R) CPU E3-1225 v3 running at 3.20 GHz, which is the CPU of our PC at home. For example, for an AMD EPYCTM 7351P CPU we find very similar times, running the program on a single core.

First, we performed a number of preliminary simulations to map out the phase diagram of the model. There is a line of second-order phase transitions that starts at $D = \infty$ extending to $D_{\text{tri}} \approx -0.5$. For smaller values of D , the transition is of first order. Our preliminary estimate for the improved model is $D^* \approx 2$. We also obtained preliminary estimates of the inverse critical temperature $\beta_c(D)$ for various values of D . Based on these preliminary results, we arranged our main simulations.

For $D = 2.05$ and 2.1 we simulated the linear lattice sizes $L = 4, 5, \dots, 14, 16, \dots, 24, 28, \dots, 48, 56, \dots, 80, 90, 100, 140, 200$, and 400 . In the case of $D = 2.0$ we simulated the same lattice sizes up to $L = 64$. Larger lattice sizes are only $L = 80$ and 200 . For example, for $D = 2.1$ we performed about 3×10^9 measurements up to $L = 32$. Then, the statistics is slowly decreasing to 6.6×10^8 measurements for $L = 100$. We performed 3.5×10^8 , 1.45×10^8 , and 1.8×10^7 measurements for $L = 140, 200$, and 400 . Most of the simulations were performed by using cycle A. For $L = 90$ and 400 , cycle B was used.

The simulations for $D = 2.0, 2.05$, and 2.1 took in total 60 years of CPU time. These simulations were performed to accurately determine the fixed-point values R_i^* of dimensionless quantities and D^* . The critical exponents ν and η are determined by using data generated for $D = 2.05$ and 2.1 .

In addition, we simulated at $D = \infty, 1.4, 1.0, 0.5, 0.0$, and -0.3 using lattice sizes up to $L = 90$. This set of simulations mainly serves to determine the exponent of leading corrections to scaling ω . Furthermore, improved observables are

TABLE II. In the first column the phenomenological coupling is specified. In the second column we give the corresponding estimates of the fixed-point values R^* obtained by separate fits for each phenomenological coupling. In the third column we give the estimates of the fixed-point values R^* obtained by joint fits of all four phenomenological couplings. In the fourth column we give the estimates of D^* , where leading corrections to scaling vanish. In the following columns, the estimates of the inverse critical temperature β_c for $D = 2.0, 2.05,$ and 2.1 are given. In rows 1–4 we give the estimates obtained by fitting the phenomenological coupling separately, while in the last row we give estimates obtained from joint fits.

R	R_{sep}^*	R_{joint}^*	D^*	$\beta_c(2.0)$	$\beta_c(2.05)$	$\beta_c(2.1)$
Z_a/Z_p	0.19479(6)	0.19477(2)	2.1(1)	0.74542805(10)	0.74296024(7)	0.74060257(7)
ξ_{2nd}/L	0.564005(30)	0.56404(2)	2.14(5)	0.74542795(8)	0.74296021(6)	0.74060251(6)
U_4	1.13933(4)	1.13929(2)	2.06(3)	0.74542800(9)	0.74296018(8)	0.74060255(8)
U_6	1.41985(15)	1.41974(5)	2.06(3)	0.74542800(9)	0.74296018(8)	0.74060255(8)
Joint			2.08(2)	0.74542801(5)	0.74296024(5)	0.74060256(5)

constructed based on these data. Also for these simulations, we spent in total 60 years of CPU time.

A. Fixed-point values of the RG-invariant quantities and critical temperatures

In this section, we determine the critical temperature for $D = 2.0, 2.05,$ and 2.1 , D^* , and the fixed-point values of phenomenological couplings. First, we analyze the phenomenological couplings one by one, similar to the analysis performed in Sec. V A of Ref. [7]. We use the *Ansätze*

$$R_i(L, D, \beta_c(D)) = R_i^*, \quad (13)$$

$$R_i(L, D, \beta_c(D)) = R_i^* + b_i(D)L^{-\epsilon_1}, \quad (14)$$

$$R_i(L, D, \beta_c(D)) = R_i^* + b_i(D)L^{-\epsilon_1} + c_i(D)L^{-\epsilon_2}, \quad (15)$$

$$R_i(L, D, \beta_c(D)) = R_i^* + b_i(D)L^{-\epsilon_1} + c_i(D)L^{-\epsilon_2} + d_i(D)L^{-\epsilon_3}, \quad (16)$$

where we approximate

$$b_i(D) = b_{s,i} (D - D^*) \quad (17)$$

linearly and $c_i(D)$ and $d_i(D)$ being constant for $D = 2.0, 2.05,$ and 2.1 that we consider here. Note that writing the leading correction amplitude $b_i(D)$ this way, D^* is an immediate parameter of the fit. Also, the slopes $b_{s,i}$ contain important information as we shall see below [Eq. (18)]. We take $\epsilon_1 = 0.76$, $\epsilon_2 = 2$, and either $\epsilon_3 = 2.2$ or $\epsilon_3 = 4$. Note that ϵ_1 is close to the estimate $\omega = 0.759(2)$ obtained below. The analytic background of the magnetic susceptibility and the violation of the rotational invariance are effectively taken into account by the term $cL^{-\epsilon_2}$. The choice $\epsilon_3 = 2.2$ is motivated by a preliminary estimate of ω_{ico} . We checked that taking, for example, $\epsilon_3 = 2.17$ instead, changes the estimates of the critical temperature and R_i^* by little. Adding a term cL^{-4} is mainly driven by the observation that this way $\chi^2/\text{d.o.f.} \approx 1$ are obtained down to $L_{\text{min}} = 5$. This observation suggests that there is a correction with an RG exponent $y \approx -4$ that has a quite large amplitude. Also, analyzing different quantities we find that adding a term cL^{-4} results in acceptable fits down to $L_{\text{min}} = 5$.

Our final results are mainly based on fits with two correction terms [Eq. (15)]. Other fits serve to estimate systematic errors. Our results are summarized in Table II.

In contrast to previous work [7], we made an attempt to jointly fit all four phenomenological couplings R that we consider. To this end, we computed the covariances of the different R . Since only quantities with the same D and L are correlated, the covariance matrix is sparse. Only 4×4 blocks are nonvanishing. For example, for $L_{\text{min}} = 8$, there are $22 + 27 + 27$ different (D, L) pairs. Hence, the covariance matrix is a $[4(22 + 27 + 27)] \times [4(22 + 27 + 27)]$ matrix. We passed the full $[4(22 + 27 + 27)] \times [4(22 + 27 + 27)]$ covariance matrix to `optimize.curve_fit` since we found no simple way to indicate that the matrix is sparse. Since the optimization typically took a few seconds, we made no effort to improve on this.

It turns out that in these joint fits, we can include more correction terms. As above, we fixed $\epsilon_1 = 0.76$ corresponding to the exponent of the leading correction ω . We consider the sequence $\epsilon_i = 2 - \eta$, 2.02 , and 2.19 of subleading correction exponents. We find that $Q > 0.1$ for $L_{\text{min}} \geq 22$, 11 , and 9 , taking into account 1, 2, or 3 subleading correction terms, respectively. Here, adding a correction $\propto L^{-4}$ does not improve the fits much.

In Fig. 1 we give the results obtained for D^* by using these fits as a function of the minimal lattice size L_{min} . In the plot, only results that correspond to $Q > 0.01$ are given. The final estimate of D^* and its error bar are chosen such that the estimates of D^* obtained by the individual fits, including their respective error bars, are contained in the interval that is given by the final estimate plus or minus its error for some range of L_{min} . For example, the results for the *Ansatz* only containing the subleading correction term $\propto L^{-2+\eta}$ are within this interval up to $L_{\text{min}} = 40$. In a similar fashion, we determine the final estimates of the fixed-point values of the phenomenological couplings and the inverse critical temperatures. These are summarized in Table II.

Eliminating leading corrections to scaling in dimensionless quantities

We construct linear combinations of two phenomenological couplings

$$R_{\text{imp},i,j} = R_i + p_{i,j}R_j, \quad (18)$$

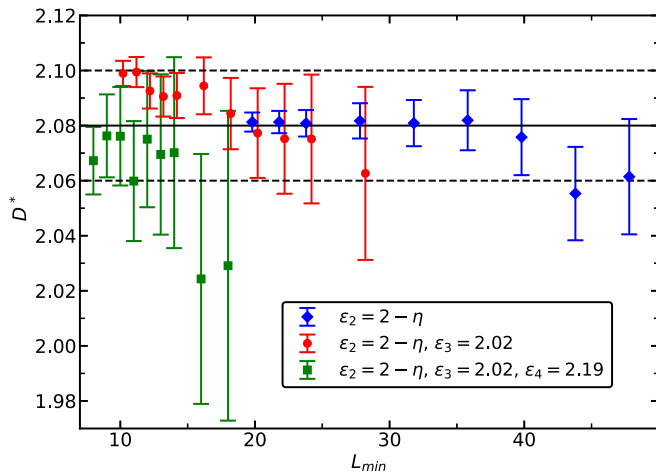


FIG. 1. We plot the estimate of D^* obtained by using joint fits of phenomenological couplings as function of the minimal lattice size L_{\min} that is taken into account. Data for $D = 2.0, 2.05,$ and 2.1 are used in the fits. Only results for $Q > 0.01$ are given. To make the figure more readable, we have shifted the values of L_{\min} slightly. The solid line indicates the preliminary estimate based on this set of fits. The dashed lines give the error estimate. The exponents ϵ_i given in the legend refer to the correction terms that are included in the *Ansätze*.

such that leading corrections to scaling are eliminated. To this end, we make use of the parameter $b_{s,i}$ [Eq. (17)], which we determined in the analysis of the phenomenological couplings discussed above. One gets

$$p_{i,j} = -\frac{b_{s,i}}{b_{s,j}}. \quad (19)$$

Note that these results hold for any model in the three-dimensional Heisenberg universality class. Here, we consider the combination of either Z_a/Z_p or ξ_{2nd}/L with U_4 . Our numerical estimates are summarized in Table III.

Jointly fitting the data for $D = 2.0, 2.05,$ and 2.1 , assuming that leading corrections to scaling vanish, we find $(Z_a/Z_p + 0.575 U_4)^* = 0.84987(3)$ and $(\xi_{2nd}/L - 0.75 U_4)^* = -0.290437(10)$. Note that these results are consistent with those obtained by naively combining the estimates of R^* given in Table II. Since the leading correction to scaling is eliminated up to the numerical uncertainty of the coefficients p_{ij} , these linear combinations are well suited to determine the inverse critical temperature β_c of models that are not improved. Furthermore, in the slope of these combinations the effective correction $\propto L^{-y_t - \omega}$ is eliminated. In the analysis of the generalized

TABLE III. Estimates of the coefficient $p_{i,j}$ [Eq. (18)] needed to construct improved phenomenological couplings.

R_i	R_j	$p_{i,j}$
Z_a/Z_p	U_4	0.575(25)
ξ_{2nd}/L	U_4	-0.750(25)

TABLE IV. Estimates of β_c for the values of D different from $D = 2.0, 2.05, 2.1$.

D	β_c
∞	0.6925051(2)
1.4	0.7854535(2)
1.0	0.8260052(2)
0.5	0.8979286(2)
0.0	0.9986988(2)
-0.3	1.0742253(4)

icosahedral model we shall not make use of this fact since ω_{ico} assumes a value that is similar to $y_t + \omega$.

In Table IV we give estimates of β_c obtained by analyzing $Z_a/Z_p + 0.575 U_4$. We use the estimate of $(Z_a/Z_p + 0.575 U_4)^*$ given above as input. The error quoted also takes into account the uncertainty of $(Z_a/Z_p + 0.575 U_4)^*$. Note that the value of β_c for $D = \infty$ is slightly smaller than $\beta_c = 0.693002(2)$ for the $O(3)$ -invariant Heisenberg model on the simple cubic lattice [34].

B. Leading corrections to scaling

In this section we focus on leading corrections to scaling. To this end, we consider the cumulants U_4 and U_6 at $Z_a/Z_p = 0.19477$ or $\xi_{2nd}/L = 0.56404$, which are our estimates of the fixed-point values of these quantities. This means that U_4 and U_6 are taken at β_f , where β_f is chosen such that either $Z_a/Z_p = 0.19477$ or $\xi_{2nd}/L = 0.56404$. In the following, we denote a cumulant at a fixed value of Z_a/Z_p or ξ_{2nd}/L by \bar{U} .

To get a first impression, we plot in Fig. 2 the Binder cumulant U_4 at $Z_a/Z_p = 0.19477$ for $D = 0.5, 1.0, 1.4, 2.05,$ and ∞ . To keep the figure readable, we do not plot the data for $D = 2.0$ and 2.1 , which are similar to those of $D = 2.05$. For $D = 1.4$ the correction amplitude has roughly the same modulus as for $D = \infty$, but opposite sign. The amplitude of leading corrections increases with decreasing D . The analysis performed below shows that the amplitudes of leading corrections for $D = 0.0$ and -0.3 are about 2.7 and 7 times as

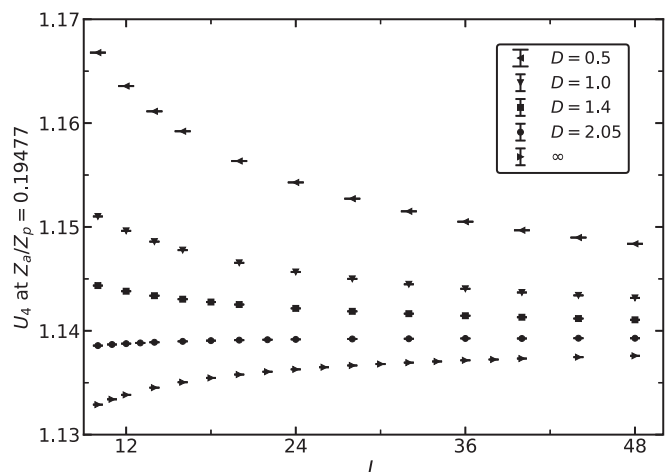


FIG. 2. We plot U_4 at $Z_a/Z_p = 0.19477$ for $D = 0.5, 1.0, 1.4, 2.05,$ and ∞ as a function of the linear lattice size L .

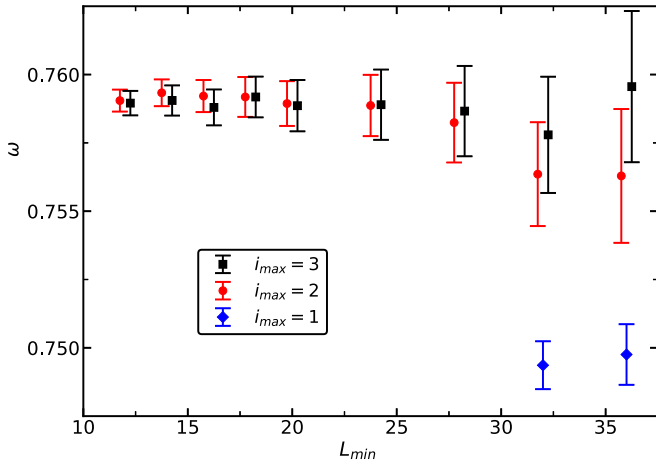


FIG. 3. We plot the estimates of ω obtained by fitting U_4 at $\xi_{2\text{nd}}/L = 0.56404$ for $D = 0.5, 1.0, 1.4, 2.0, 2.05, 2.1,$ and ∞ by using the *Ansatz* (20) as a function of the minimal lattice size L_{min} that is taken into account. Only results for $Q > 0.01$ are given. To make the figure more readable, we have shifted the values of L_{min} slightly.

large as for $D = 0.5$, respectively. The results obtained for $\xi_{2\text{nd}}/L = 0.56404$ are similar. The results obtained for U_6 are qualitatively the same as for U_4 .

We performed joint fits for several sets of D . Either all values of D are taken into account, or subsets of them. These subsets are obtained by skipping values of D starting from the smallest one. The minimal set that we consider consists of $D = 1.4, 2.0, 2.05, 2.1,$ and ∞ . Similar to Ref. [7] we analyzed our data by using the *Ansatz*

$$\bar{U} = \bar{U}^* + \sum_{i=1}^{i_{\text{max}}} c_i [b(D)L^{-\omega}]^i + dL^{-\epsilon} \quad (20)$$

for various values of i_{max} . In order to avoid ambiguity, we set $c_1 = 1$. The free parameters of the fit are $\bar{U}^*, c_2, c_3, \dots, b(D)$ for each value of D . In our fits, the parameter d is the same for all values of D . In our fits we set $\epsilon = 2$.

A preliminary study shows that the results obtained for $\xi_{2\text{nd}}/L = 0.56404$ are more stable than those for $Z_a/Z_p = 0.19477$. In particular, the values of c_2, c_3, \dots have a smaller modulus for $\xi_{2\text{nd}}/L = 0.56404$ than for $Z_a/Z_p = 0.19477$. Therefore, in the following we shall focus on U_4 and U_6 at $\xi_{2\text{nd}}/L = 0.56404$.

In Fig. 3 we plot our results for the correction exponent ω obtained from joint fits of U_4 at $\xi_{2\text{nd}}/L = 0.56404$ for $D = 0.5, \dots, \infty$. We give our results for $i_{\text{max}} = 1, 2,$ and 3 as a function of the minimal lattice size L_{min} that is taken into account. We find that fits with $i_{\text{max}} = 2$ and 3 are consistent and acceptable fits are obtained starting from $L_{\text{min}} = 12$. In contrast, for $i_{\text{max}} = 1$ we get $Q > 0.1$ only for $L_{\text{min}} \geq 40$. The estimate of ω obtained with $i_{\text{max}} = 1$ is considerably smaller than for $i_{\text{max}} = 2$ and 3 . As our preliminary estimate we take $\omega = 0.7589(10)$ from $L_{\text{min}} = 20$.

Performing a similar analysis, taking into account all values of D , we get consistent results for ω starting from $i_{\text{max}} = 3$ and 4 . Here, we take $\omega = 0.7595(10)$ from $L_{\text{min}} = 24$ as preliminary estimate. Taking the set $D = 1.4, \dots, \infty$ we get

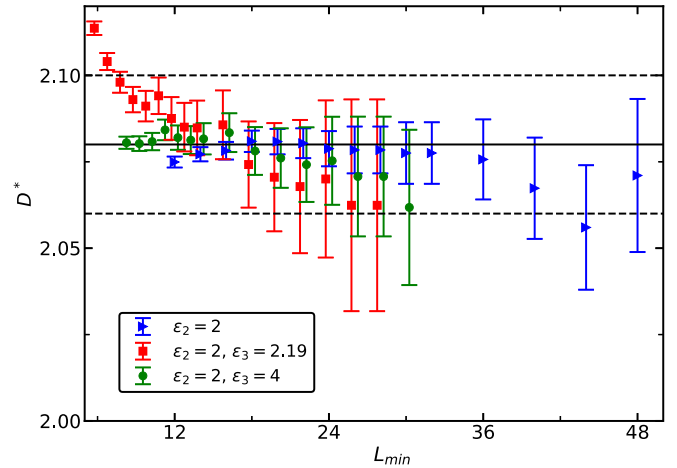


FIG. 4. We plot estimates of D^* obtained from fitting U_4 at $\xi_{2\text{nd}}/L = 0.56404$ by using the *Ansätze* (22) and (23) as a function of L_{min} . Only results for $Q > 0.01$ are given. To make the figure more readable, we have shifted the values of L_{min} slightly. The solid line indicates the preliminary results based on this set of fits. The dashed lines give the error estimate.

already consistent results for ω with $i_{\text{max}} = 1$ and 2 . Here, we take $\omega = 0.7584(10)$ from $L_{\text{min}} = 20$ as preliminary estimate. As a check we performed fits without the term $dL^{-\epsilon}$. We find that $\chi^2/\text{d.o.f.}$ considerably increases. However, the estimates for ω are consistent with those obtained from fits with such a term.

Analyzing U_6 at $\xi_{2\text{nd}}/L = 0.56404$ we get similar results as for U_4 . As our final estimate we quote

$$\omega = 0.759(2), \quad (21)$$

which covers the preliminary estimates discussed above. From the analysis of U_4 and U_6 at $Z_a/Z_p = 0.19477$ we would arrive at $\omega = 0.758(4)$. Since the fits for $\xi_{2\text{nd}}/L = 0.56404$ are clearly better behaved than those for $Z_a/Z_p = 0.19477$, we stick with the result obtained from $\xi_{2\text{nd}}/L = 0.56404$ as our final estimate.

C. D^*

Here, we analyze \bar{U}_4 and \bar{U}_6 at $D = 2.0, 2.05,$ and 2.1 . To this end we consider the *Ansätze*

$$\bar{U} = \bar{U}^* + b(D)L^{-\epsilon_1} + cL^{-\epsilon_2}, \quad (22)$$

$$\bar{U} = \bar{U}^* + b(D)L^{-\epsilon_1} + cL^{-\epsilon_2} + dL^{-\epsilon_3}, \quad (23)$$

where we fix $\epsilon_1 = 0.759$, which is our estimate of ω obtained above, $\epsilon_2 = 2$, which effectively takes into account $2 - \eta$ and ω_{NR} . We take either $\epsilon_3 = 2.19$, which corresponds to ω_{ico} or $\epsilon_3 = 4$. We parametrize the leading correction as $b(D) = b_s (D - D^*)$, where b_s and D^* are free parameters of the fit. Furthermore, $\bar{U}^*, c,$ and d are free parameters. Here, we assume that c and d are the same for all three values of D , which should be a reasonable approximation. Below we focus on \bar{U}_4 since the results for \bar{U}_6 are similar. In Fig. 4 we plot results obtained for D^* by fitting U_4 at $\xi_{2\text{nd}}/L = 0.56404$ using the *Ansätze* (22) and (23). As our preliminary estimate

we take $D^* = 2.08(2)$. In a similar fashion we arrive at $\bar{U}_4^* = 1.139\,295(20)$ for $\xi_{2\text{nd}}/L = 0.564\,04$.

Analyzing U_4 at $Z_a/Z_p = 0.194\,77$ we arrive at $D^* = 2.07(3)$ and $\bar{U}_4^* = 1.139\,30(2)$. As our final estimate for the improved model we take

$$D^* = 2.08(2), \quad (24)$$

which is the result of the joint analysis performed in Sec. IV A and the analysis of U_4 at $\xi_{2\text{nd}}/L = 0.564\,04$.

D. Critical exponent ν of the correlation length

We compute the exponent $\nu = 1/y_t$ from the slope of a phenomenological coupling R_j at a given value $R_{i,f}$ of a second quantity R_i , where R_j and R_i might be the same. Following the discussion of Sec. III C of Ref. [7] these slopes behave as

$$\begin{aligned} \bar{S}_{i,j} &= \left. \frac{\partial R_j}{\partial \beta} \right|_{R_i=R_{i,f}} = aL^{y_t} [1 + bL^{-\omega} + \dots \\ &+ c_{\text{back}}L^{-2-\eta} + c_{NR}L^{-\omega_{NR}} + c_{\text{ico}}L^{-\omega_{\text{ico}}} + \dots] \\ &+ dL^{-\omega} + \dots \end{aligned} \quad (25)$$

Note that the coefficients a , b , c_{back} , c_{NR} , c_{ico} , and d depend on the quantity that is considered and on the model, which means in the present case on the parameter D . As discussed in Ref. [7] and references therein, it is advantageous to take $R_{i,f} \approx R_i^*$ since otherwise an effective correction $\propto (R_{i,f} - R_i^*)L^{-y_t}$ has to be taken into account.

Below we consider $D = 2.05$ and 2.1 which are close to D^* . Therefore, the coefficient b of the leading correction is small for all quantities. In order to ensure that leading corrections to scaling can be safely ignored at the level of our accuracy, we construct improved slopes by multiplying \bar{S} by a certain power p of the Binder cumulant \bar{U}_4 :

$$\bar{S}_{\text{imp}} = \bar{S}\bar{U}_4^p, \quad (26)$$

where both \bar{S} and \bar{U}_4 are taken at $R_{i,f}$. The exponent p is chosen such that, at the level of our numerical accuracy, leading corrections to scaling are eliminated. This idea is discussed systematically in Ref. [35]. To determine p , we consider the pair $(D_1, D_2) = (1.4, \infty)$. Note that as discussed in Sec. IV B, the amplitude of the leading correction to scaling for these two values of D has approximately the same modulus but opposite sign. We fit ratios of $\bar{S}_{i,j}$ and \bar{U}_4 with the *Ansätze*

$$\begin{aligned} \frac{\bar{S}_{i,j}(D_1)}{\bar{S}_{i,j}(D_2)} &= a_S(1 + b_S L^{-\epsilon_1}), \\ \frac{\bar{S}_{i,j}(D_1)}{\bar{S}_{i,j}(D_2)} &= a_S(1 + b_S L^{-\epsilon_1} + c_S L^{-\epsilon_2}) \end{aligned} \quad (27)$$

and

$$\frac{\bar{U}_4(D_1)}{\bar{U}_4(D_2)} = 1 + b_U L^{-\epsilon_1}, \quad \frac{\bar{U}_4(D_1)}{\bar{U}_4(D_2)} = 1 + b_U L^{-\epsilon_1} + c_U L^{-\epsilon_2}, \quad (28)$$

where we fixed $\epsilon_1 = 0.76$ and $\epsilon_2 = 2$. The exponent p is given by

$$p = -\frac{b_S}{b_U}. \quad (29)$$

TABLE V. Numerical result for the exponent p that eliminates leading corrections to scaling in $\bar{S}_{i,j}$ [Eq. (26)].

Fixing\Slope of	Z_a/Z_p	$\xi_{2\text{nd}}/L$	U_4	U_6
$Z_a/Z_p = 0.19477$:	1.65(10)	0.24(10)	-3.6(2)	-5.0(3)
$\xi_{2\text{nd}}/L = 0.56404$:	0.07(7)	0.24(10)	-4.22(10)	-5.64(10)

In Table V we give our final results for p . The error bar takes into account statistical errors as well as systematical ones, which are estimated by comparing the results of the two different *Ansätze*.

Let us briefly comment on the statistical error of the different quantities. We find that both fixing $Z_a/Z_p = 0.194\,77$ and $\xi_{2\text{nd}}/L = 0.564\,04$ changes, compared with fixed β , the relative statistical error of the slopes only little. The same holds for the comparison of the improved and the unimproved slopes. We see big differences between the relative statistical errors of the slopes of the different phenomenological couplings. The relative error is the smallest in the case of $\xi_{2\text{nd}}/L$. The ratios of the statistical errors vary only little with the linear lattice size. For example, for $D = 2.05$ and $L = 40$ at β_c we find that the relative statistical error of the slope of Z_a/Z_p , U_4 , and U_6 is by a factor of 1.24, 1.99, and 2.00 larger than that of $\xi_{2\text{nd}}/L$. As a measure of the effort to reach a certain accuracy, beyond the increase due to the increasing lattice size, we studied $w = n_{\text{stat}}\epsilon_r^2$, where n_{stat} is the number of measurements and ϵ_r the relative statistical error. We fitted the data for the slope of Z_a/Z_p at $Z_a/Z_p = 0.194\,77$ for $D = 2.05$. We find a behavior $w \propto L^x$ with $x \approx 0.36$. Since we already averaged over bins during the simulation, we can not determine to what extent this degradation of the efficiency is due to an increase of autocorrelation times or and increase of the variance of the slope.

Below we perform throughout joint fits of the data for $D = 2.05$ and 2.1 . In these fits, the overall amplitude for each value of D is a free parameter of the fit. In contrast, we assume that the correction amplitudes are similar, and are taken to be the same in the *Ansatz*.

First, we have analyzed the improved slopes of the different phenomenological couplings separately. We have fitted these quantities by using the *Ansätze*

$$S = aL^{y_t}, \quad (30)$$

$$S = aL^{y_t} (1 + bL^{-\epsilon_1}), \quad (31)$$

where we take $\epsilon_1 = 2$, which should effectively take into account corrections due to the analytic background of the magnetic susceptibility and the violation of the rotational invariance by the simple cubic lattice. First, we analyzed our data by using *Ansatz* (30) without correction term. In Fig. 5 we give our results for improved slopes at $\xi_{2\text{nd}}/L = 0.564\,04$. We do not give results for U_6 since they are very similar to those for U_4 . Note that, for example, for $L_{\text{min}} = 40$ we get $\chi^2/\text{d.o.f.} = 0.949$ and 1.150 for the slopes of $\xi_{2\text{nd}}/L$ and Z_a/Z_p , respectively. This corresponds to $Q = 0.526$ and 0.286 , respectively. Despite this fact, the estimates of y_t obtained for $L_{\text{min}} = 40$ clearly differ for $\xi_{2\text{nd}}/L$ and Z_a/Z_p . As our preliminary estimate we take $y_t = 1.405\,20(32)$. It is

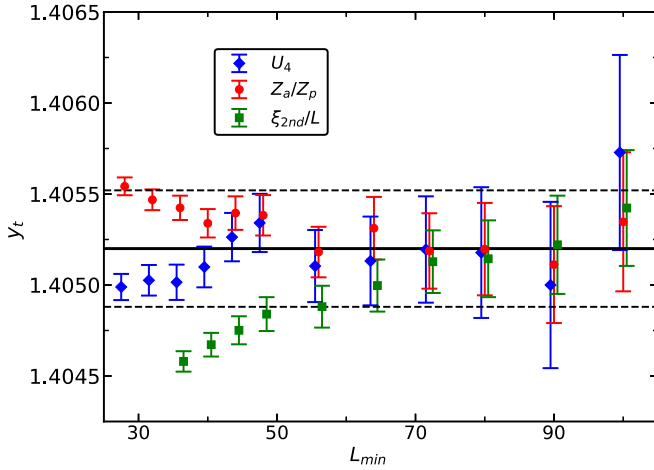


FIG. 5. We plot the estimates of y_t obtained from fitting the improved slopes of Z_a/Z_p , ξ_{2nd}/L , and U_4 at $\xi_{2nd}/L = 0.56404$ for $D = 2.05$ and 2.1 by using the *Ansatz* (30) as a function of the minimal lattice size L_{\min} that is taken into account. Only results for $Q > 0.01$ are given. To make the figure more readable, we have shifted the values of L_{\min} slightly. The solid line indicates the preliminary result based on this set of fits. The dashed lines give the error estimate.

chosen such that all three results for $L_{\min} = 72$ are covered. The estimates obtained for $Z_a/Z_p = 0.19477$ are similar.

In Fig. 6 we give results obtained from fitting the improved slopes of Z_a/Z_p , ξ_{2nd}/L , and U_4 at $\xi_{2nd}/L = 0.56404$ by using the *Ansatz* (31). As our preliminary estimate of this set of fits we take $y_t = 1.40520(20)$. It covers all three estimates obtained for $L_{\min} = 24$. Analyzing the slopes at $Z_a/Z_p = 0.19477$ in a similar way, we find consistent results.

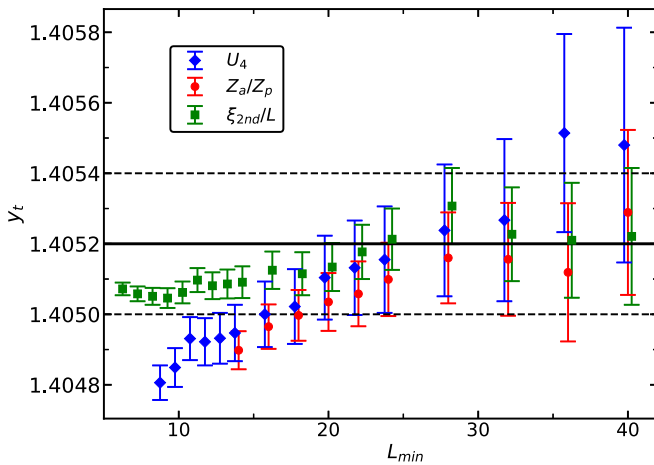


FIG. 6. We plot the estimates of y_t obtained from fitting the improved slopes of Z_a/Z_p , ξ_{2nd}/L , and U_4 at $\xi_{2nd}/L = 0.56404$ for $D = 2.05$ and 2.1 by using the *Ansatz* (31) as a function of the minimal lattice size L_{\min} that is taken into account. Only results for $Q > 0.01$ are shown. To make the figure more readable, we have shifted the values of L_{\min} slightly. The solid line indicates our preliminary estimate of this set of fits. The dashed lines give the error estimate.

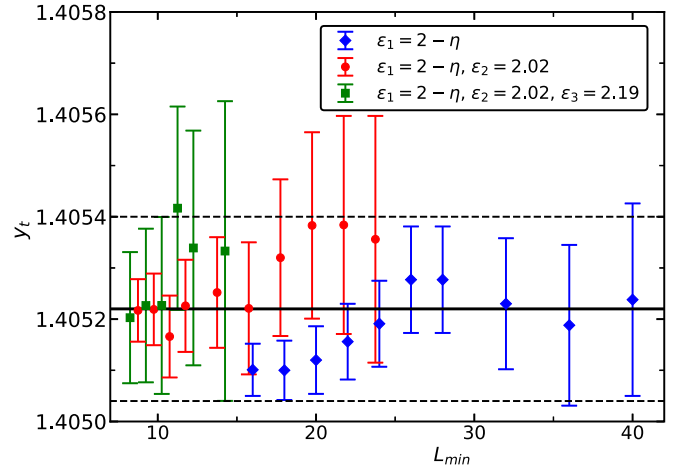


FIG. 7. We plot the estimates of y_t obtained by fitting the improved slopes of Z_a/Z_p , ξ_{2nd}/L , U_4 , and U_6 at $\xi_{2nd}/L = 0.56404$ jointly by using up to three different correction terms as a function of the minimal lattice size L_{\min} that is taken into account. Data for $D = 2.05$ and 2.1 are taken. Only results for $Q > 0.01$ are given. To make the figure more readable, we have shifted the values of L_{\min} slightly. The solid line indicates our preliminary estimate based on this set of fits. The dashed lines give the error estimate. The exponents ϵ_i given in the legend refer to the correction terms that are included in the *Ansätze*.

Finally, we performed a joint analysis of the improved slopes of all four phenomenological couplings at either $\xi_{2nd}/L = 0.56404$ or $Z_a/Z_p = 0.19477$. Similar to Sec. IV A, we took the covariances of the different quantities into account. In these fits, we used *Ansätze* with up to three different correction terms with the effective correction exponents $\epsilon_1 = 2 - \eta$, $\epsilon_2 = 2.02$, and $\epsilon_3 = 2.19$. The first is motivated by the analytic background of the magnetic susceptibility, the second by ω_{NR} , and the third by ω_{ico} . Note that in the slope we also expect corrections with the exponent $y_t + \omega$, which is effectively taken into account by ϵ_3 . In Fig. 7 we give our results for the improved slopes at $\xi_{2nd}/L = 0.56404$. In Fig. 8 we give the corresponding results for $Z_a/Z_p = 0.19477$.

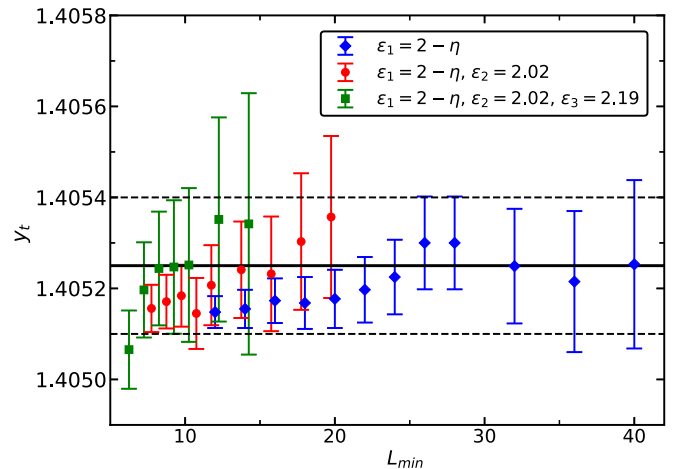


FIG. 8. Same as Fig. 7 but for $Z_a/Z_p = 0.19477$ instead of $\xi_{2nd}/L = 0.56404$.

As our preliminary estimates we take $y_t = 1.405\,22(18)$ and $1.405\,25(15)$ obtained for $\xi_{2\text{nd}}/L = 0.564\,04$ and $Z_a/Z_p = 0.194\,77$, respectively.

Based on these results and the preliminary estimate obtained by fitting the slopes of the different phenomenological couplings separately by using the *Ansatz* (31) we conclude

$$y_t = 1.4052(2), \quad (32)$$

which corresponds to $\nu = 0.711\,64(10)$.

E. Critical exponent η

We analyzed the improved quantities

$$\bar{\chi}_{\text{imp}} = \bar{\chi} \bar{U}_4^p, \quad (33)$$

where both χ and U_4 are taken either at $Z_a/Z_p = 0.194\,77$ or $\xi_{2\text{nd}}/L = 0.564\,04$. We computed the exponent p in a similar way as in the previous section for the slopes S . Therefore, we skip a detailed discussion and only report our results $p = -1.31(3)$ and $-0.23(4)$ for $Z_a/Z_p = 0.194\,77$ and $\xi_{2\text{nd}}/L = 0.564\,04$, respectively.

Let us briefly discuss the effect of taking χ at $Z_a/Z_p = 0.194\,77$ or $\xi_{2\text{nd}}/L = 0.564\,04$ on the statistical error. In previous work (see Ref. [7] and references therein), we observed that the statistical error is reduced compared with χ at a fixed value of $\beta \approx \beta_c$. Here we see for $Z_a/Z_p = 0.194\,77$ only a small effect, while for $\xi_{2\text{nd}}/L = 0.564\,04$ we see, for example, for $D = 2.05$ a reduction of the statistical error by a factor of about 2. The relative statistical error of the improved susceptibility is by a few percent larger than that of the unimproved counterpart.

We fitted our data with the *Ansätze*

$$\chi = aL^{2-\eta}, \quad (34)$$

$$\chi = aL^{2-\eta} + b, \quad (35)$$

$$\chi = aL^{2-\eta} (1 + cL^{-\epsilon_2}) + b, \quad (36)$$

where the analytic background b can be viewed as an effective correction with the exponent $\epsilon_1 = 2 - \eta$. Similar to the analysis of the slopes, we performed joint fits of the data for $D = 2.05$ and 2.1 .

In Fig. 9 we give the estimates obtained from fitting the improved magnetic susceptibility at $Z_a/Z_p = 0.194\,77$ by using the *Ansätze* (35) and (36). In the case of *Ansatz* (36) we plot results for $\epsilon_2 = 2.02$ and 4 . We also performed fits using $\epsilon_2 = 2.19$, which give consistent results for η . Our preliminary estimate $\eta = 0.037\,84(5)$ for this set of fits is consistent with the estimate obtained by using *Ansatz* (36) with $\epsilon_2 = 2.02$ for $L_{\text{min}} = 16$. Furthermore, it covers the results obtained by using the *Ansatz* (35) for $L_{\text{min}} = 14$ up to 32 and *Ansatz* (36) with $\epsilon_2 = 4$ for $L_{\text{min}} \leq 24$. Fitting the improved magnetic susceptibility at $\xi_{2\text{nd}}/L = 0.564\,04$ by using the *Ansätze* (35) and (36) we find results that are consistent with the estimate $\eta = 0.037\,84(5)$.

Finally, in Fig. 10 we plot the estimates obtained from fits of the data for the improved magnetic susceptibility at $\xi_{2\text{nd}}/L = 0.564\,04$ without correction term (34) and with a correction corresponding to the analytic background of the magnetic susceptibility (35). Based on these fits we arrive

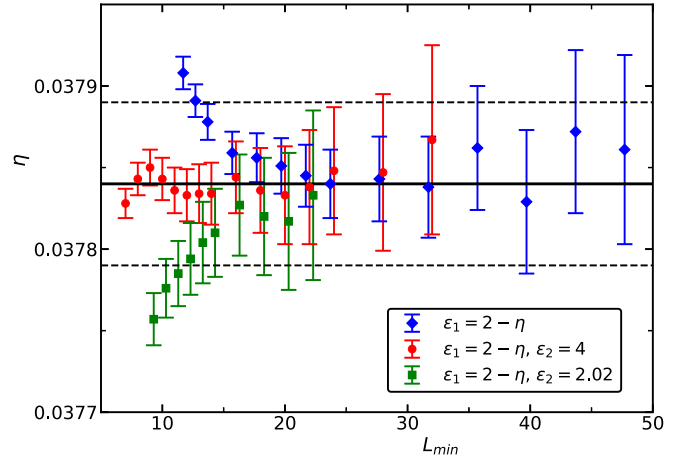


FIG. 9. We plot the estimates of η obtained from fitting the improved magnetic susceptibility at $Z_a/Z_p = 0.194\,77$ by using the *Ansätze* (35) and (36) as a function of the minimal lattice size L_{min} that is taken into account. Data for $D = 2.05$ and 2.1 are used in the fits. Only results for $Q > 0.01$ are given. The numbers given in the legend refer to the corrections that are taken into account in the *Ansatz*. To make the figure more readable, we have shifted the values of L_{min} slightly. The solid line gives our preliminary estimate and the dashed lines indicate the error bar.

at the preliminary estimate $\eta = 0.037\,84(7)$ which takes into account the result obtained by using the *Ansatz* (34) with $L_{\text{min}} = 140$ and the results obtained by using the *Ansatz* (35) up to $L_{\text{min}} = 80$. As our final estimate of η we quote

$$\eta = 0.037\,84(5) \quad (37)$$

obtained by using *Ansätze* that include correction terms.

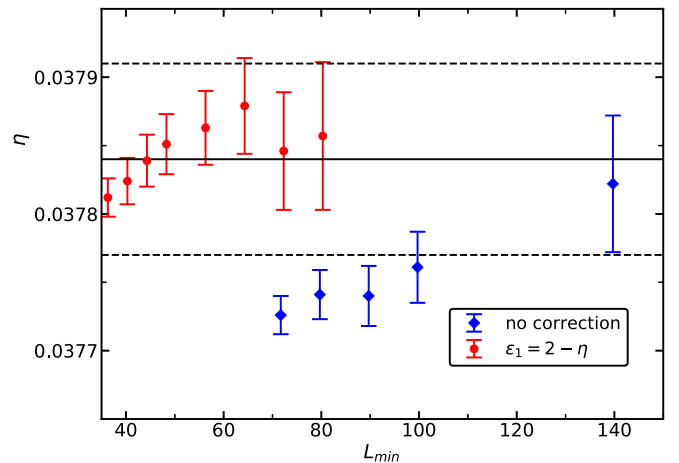


FIG. 10. We plot the estimates of η obtained from fitting the improved magnetic susceptibility at $\xi_{2\text{nd}}/L = 0.564\,04$ as a function of the minimal lattice size L_{min} that is taken into account. Data for $D = 2.05$ and 2.1 are used. To make the figure more readable, we have shifted the values of L_{min} slightly. Only results for $Q > 0.01$ are given. Either no correction or a term corresponding to the analytic background is used in the *Ansatz*. The solid line gives our preliminary estimate of η and the dashed lines indicate the error bar.

V. SUMMARY AND DISCUSSION

We have studied the generalized icosahedral model on the simple cubic lattice. In this model, the field variable might take a normalized vertex of the icosahedron as value. Analogous to the Blume-Capel model, in addition (0,0,0) is allowed. The density of the (0,0,0) sites is governed by the parameter D of the reduced Hamiltonian. For a certain range of D , the model undergoes a second-order phase transition. At the critical line, the symmetry is enhanced to $O(3)$. Hence, the transition belongs to the universality class of the three-dimensional Heisenberg model. In Appendix B we find that a perturbation of the $O(3)$ -invariant fixed point with the symmetry of the icosahedron is related with the irrelevant RG eigenvalue $y_{\text{ico}} = -2.19(2)$. On the critical line, the amplitude of leading corrections to scaling depends of the parameter D . Numerically, we find that for $D^* = 2.08(2)$ this amplitude vanishes. Based on a finite-size scaling analysis of phenomenological couplings, such as the Binder cumulant, their slopes and the magnetic susceptibility we arrive at accurate estimates of the critical exponents ν and η and the correction exponent ω . In Appendix A we analyze data obtained for the three-component ϕ^4 model on the simple cubic lattice, leading to consistent results for the exponents ν and η , confirming that both models share the same universality class. The precision of our results clearly surpasses that of experiments. However, one should note that there had been theoretical advances in recent years made by different methods. Here, our results serve as a benchmark. In the Introduction, in Table I we confront our results with ones given in the literature. Comparing with the ϵ expansion, we find significant deviations. The estimates obtained by using the conformal bootstrap method [17] and the recent implementation of the functional renormalization group method [18] are consistent with, but less precise than, ours.

Our precise estimates of the inverse critical temperature for various values of D and λ for the generalized icosahedral model and the ϕ^4 model, respectively, might serve as input for studies focusing on other properties of these models. In particular, we intend to compute the structure constants using a similar approach as in Ref. [36] for the Ising universality class. Furthermore, it would be interesting to investigate the symmetry properties of the icosahedral model in the low-temperature phase.

Our motivation to study the icosahedral model is of technical nature. In order to save the field variable at one site, only 4 bits are needed. For practical reasons, in our program an 8-bit char variable is used. Furthermore, probabilities needed for the Metropolis and the cluster update can be computed and tabulated at the beginning of the simulation. For our implementation we find a speedup by roughly a factor of 3 compared with the ϕ^4 model studied, for example, in Refs. [8–10]. This advantage is partially abrogated by the correction $\propto L^{y_{\text{ico}}}$ that is not present in a model with $O(3)$ symmetry at the microscopic level. Note that the situation is different for the $(q+1)$ -state clock model studied in Ref. [7]. In this case, the irrelevant exponent y_q is rapidly decreasing with q . In Ref. [7] we focused on $q = 8$, where $y_{q=8} = -5.278(9)$ (see Ref. [23]). Hence, the correction can be ignored in the analysis of the data, meaning that we have the technical advantage without a downside.

ACKNOWLEDGMENT

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APPENDIX A: THE ϕ^4 MODEL ON THE LATTICE

The ϕ^4 model on the simple cubic lattice is defined by the reduced Hamiltonian

$$\mathcal{H}_{\phi^4} = -\beta \sum_{\langle xy \rangle} \vec{\phi}_x \cdot \vec{\phi}_y + \sum_x [\vec{\phi}_x^2 + \lambda(\vec{\phi}_x^2 - 1)^2], \quad (\text{A1})$$

where $\vec{\phi}_x \in \mathbb{R}^N$ with $N = 3$ in our case. We performed simulations for $\lambda = 5$ and 5.2. Note that $\lambda^* = 5.2(4)$ [Eq. (B13) of Ref. [10]]. We simulated at $\beta = 0.687\,563\,8$ and $0.687\,985$ in the case of $\lambda = 5$ and 5.2, respectively. These are the estimates of β_c obtained in Ref. [9] and in preliminary simulations, respectively. The simulations are organized in a similar fashion as for the generalized icosahedral model. For $\lambda = 5.2$ we have simulated the linear lattice sizes $L = 8, 9, \dots, 20, 22, \dots, 30, 34, 40, 50, 60, 80, 100, 140, 200$, and 300. The number of measurements decreases with increasing lattice size. Up to $L = 19$ we performed about 3×10^9 measurements. For $L = 300$ we performed 3.75×10^7 measurements. In total, we spent about 13.5 years of CPU time on these simulations. In the case of $\lambda = 5.0$ we performed simulations for fewer lattice sizes. We simulated at $L = 8, 10, \dots, 30, 34, 40, 50, 60, 80, 100$ and 140. In total, we spent about 5.5 years of CPU time on these simulations.

1. Inverse critical temperature

First, we determine the inverse critical temperature by analyzing the improved phenomenological coupling $Z_a/Z_p + 0.575 U_4$. We fit our data with the *Ansatz*

$$R(L, \beta_c) = R^* + cL^{-2}, \quad (\text{A2})$$

where we use the estimate $(Z_a/Z_p + 0.575 U_4)^* = 0.849\,87(3)$ that we obtained from the analysis of the data for the icosahedral model in Sec. IV A. As final result we get

$$\beta_c(\lambda = 5.0) = 0.687\,561\,27(13)[6], \quad (\text{A3})$$

$$\beta_c(\lambda = 5.2) = 0.687\,985\,21(8)[3], \quad (\text{A4})$$

where the error in [...] is due to the uncertainty of $(Z_a/Z_p + 0.575 U_4)^*$.

2. Improved model

Here, we study the behavior of U_4 at $\xi_{2\text{nd}}/L = 0.564\,04$ or $Z_a/Z_p = 0.194\,77$. We perform fits similar to those performed in Sec. IV C. Here, we only include the data obtained for $\lambda = 5.0$ and 5.2. Furthermore, we fix the value of \bar{U}_4^* to that obtained in Sec. IV C. For $\xi_{2\text{nd}}/L = 0.564\,04$ we get $\lambda^* = 5.19(2)[6]$, while for $Z_a/Z_p = 0.194\,77$ we get $\lambda^* = 5.14(2)[6]$, where the error in [...] is due to the uncertainty of \bar{U}_4^* . Our final estimate

$$\lambda^* = 5.17(11) \quad (\text{A5})$$

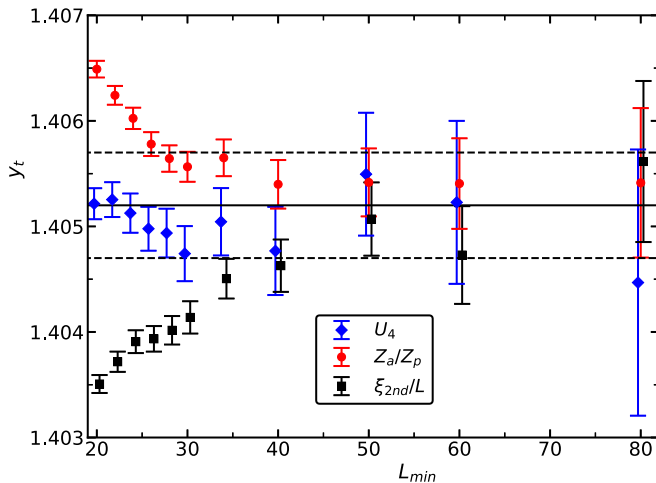


FIG. 11. We plot the estimates of y_t obtained by fitting the improved slopes of phenomenological couplings at $Z_a/Z_p = 0.19477$ by using the *Ansatz* (30) without corrections as a function of L_{\min} . Data for the ϕ^4 model at $\lambda = 5$ and 5.2 are analyzed jointly. Only results for $Q > 0.01$ are given. To make the figure more readable, we have shifted the values of L_{\min} slightly. The solid line indicates the final estimate that we obtain from this set of fits. The dashed lines give the error bar.

is chosen such that both estimates, including their errors are covered.

3. Finite-size scaling estimate of ν

Here, we performed an analysis similar to that for the icosahedral model in Sec. IV D. The data for $D = 2.05$ and 2.1 for the icosahedral model are replaced by those for $\lambda = 5.0$ and 5.2. Below we discuss results obtained by analyzing improved slopes at $Z_a/Z_p = 0.19477$. The corresponding results for $\xi_{2nd}/L = 0.56404$ differ only by little.

In Fig. 11 we give estimates of y_t obtained by using an *Ansatz* without correction term, Eq. (30). Similar to Fig. 5 we find that for small L_{\min} the estimates obtained from different phenomenological couplings do not agree within their respective error bars. As our final estimate we take

$$y_t = 1.4052(5), \quad (\text{A6})$$

corresponding to $\nu = 0.71164(25)$. This estimate is consistent with results obtained for some range of L_{\min} for each of the three phenomenological couplings.

4. Finite-size scaling estimate of the exponent η

We performed joint fits of the data for the magnetic susceptibility at $\lambda = 5.0$ and 5.2 by using the *Ansatz* (35) or the *Ansatz* (36) using either $\epsilon_2 = 2.02$ or 4. We analyzed both the improved magnetic susceptibility at $Z_a/Z_p = 0.19477$ and $\xi_{2nd}/L = 0.56404$. The results of such fits for $Z_a/Z_p = 0.19477$ are plotted in Fig. 12. Our preliminary estimate $\eta = 0.03784(8)$ is chosen such that the estimates of η obtained by using the three different *Ansätze* are contained in the range 0.03784 ± 0.00008 for some range of the minimal lattice size L_{\min} . Performing a similar analysis for $\xi_{2nd}/L = 0.56404$ we arrive at the slightly smaller estimate $\eta = 0.03780(8)$. As the

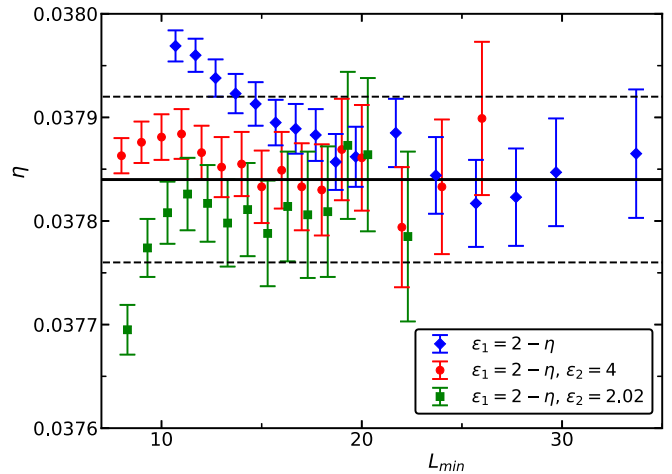


FIG. 12. We plot the estimates of η obtained by fitting the improved magnetic susceptibility χ at $Z_a/Z_p = 0.19477$ for $\lambda = 5.0$ and 5.2 jointly. The effective correction exponents given in the legend refer to the *Ansätze* (35) and (36). In the fits all lattice sizes $L \geq L_{\min}$ are taken into account. To make the figure more readable, we have shifted the values of L_{\min} slightly. The solid line gives the preliminary estimate that we obtain from this set of fits. The dashed lines give the error bar.

final estimate we quote

$$\eta = 0.03782(10), \quad (\text{A7})$$

which covers both the estimates obtained from the data for fixing $Z_a/Z_p = 0.19477$ and $\xi_{2nd}/L = 0.56404$.

5. Reanalysis of the high-temperature series expansion

Our more precise estimates of λ^* and the more accurate estimate of β_c at $\lambda = 5.0$ are used to bias the analysis of the high-temperature series performed in Ref. [9]. We start from Tables XXV and XXVI in Appendix B of Ref. [9]. For $\lambda = 4, 4.5, \text{ and } 5$ the high-temperature series of the magnetic susceptibility and the second moment correlation length is analyzed by using biased integral approximants [19]. To this end, the estimates of β_c given in Eqs. (B5)–(B7) of [9] are used. Let us discuss the details of our reanalysis at the example of the exponent ν , given in Table XXVI of Appendix B. The estimates of ν carry two types of error estimates: the number given (...) is obtained from the spread of different approximants, while the number given in [...] is due to the uncertainty of the estimate of β_c . It was obtained by reanalyzing the series for $\beta_c \pm \Delta\beta_c$, where $\Delta\beta_c$ is the estimate of the error of β_c . Note that the value obtained for the exponent is increasing with an increasing estimate of β_c . Hence, the new estimate of the exponent is given by

$$\nu_{\text{new}} = \nu_{\text{old}} + (\beta_{c,\text{new}} - \beta_{c,\text{old}}) \frac{\Delta\nu}{\Delta\beta_c}. \quad (\text{A8})$$

Here, “new” refers to this work, while “old” refers to Ref. [9]. $\Delta\beta_c$ is the error estimate of β_c in Ref. [9] and $\Delta\nu$ refers to the number given in [...] in Table XXVI of Ref. [9].

Shifting the estimates of ν for $\lambda = 5$ for the approximants bIA1 and bIA2 we arrive at $\nu = 0.71141(5)[1]$ and $0.71144(6)[1]$, respectively. For $\lambda = 4.5$, using the estimate

of β_c obtained in Ref. [10], we arrive at $\nu = 0.71103(3)$ [4] and $0.71102(6)$ [4]. Finally, extrapolating to $\lambda^* = 5.17(11)$ we arrive at

$$\nu = 0.7116(2). \quad (\text{A9})$$

Performing a similar analysis for the exponent of the magnetic susceptibility we arrive at $\gamma = 1.3965(3)$. Note that in Ref. [9] $\gamma = 1.3960(9)$ is quoted. Plugging in $\lambda^* = 5.17(11)$ into Eq. (19) of Ref. [9],

$$\eta\nu = 0.02665(18) + 0.00035(\lambda - 4.5), \quad (\text{A10})$$

we arrive at

$$\eta = 0.0378(3), \quad (\text{A11})$$

where we took into account the uncertainty of λ^* and ν .

APPENDIX B: CORRECTION EXPONENT ω_{ico}

We consider the quantity

$$q = \frac{\langle \max_j \vec{v}_j \cdot \vec{m} \rangle}{\langle |\vec{m}| \rangle}, \quad (\text{B1})$$

where $\vec{m} = \sum_x \vec{s}_x$ is the magnetization of a given configuration, and \vec{v}_j , with $j = 1, 2, \dots, 12$, are the 12 possible values of the spin with unit length. Alternatively, one might define a quantity based on the polynomial given in Ref. [20]. For an O(3)-invariant distribution of \vec{m} the value of q can be easily computed by using numerical integration. We get

$$q^* = 0.915874306174. \quad (\text{B2})$$

The deviation from an O(3)-invariant distribution is now quantified by

$$\bar{q} = q - q^*. \quad (\text{B3})$$

We computed q at either $Z_a/Z_p = 0.19477$ or $\xi_{2\text{nd}}/L = 0.56404$. It turns out that the numbers for $Z_a/Z_p = 0.19477$ and $\xi_{2\text{nd}}/L = 0.56404$ are very similar and the estimates of ω_{ico} are essentially the same for these two cases. Therefore, we restrict the discussion below on $Z_a/Z_p = 0.19477$. We fitted \bar{q} by using the *Ansätze*

$$\bar{q} = aL^{-\omega_{\text{ico}}} \quad (\text{B4})$$

and

$$\bar{q} = aL^{-\omega_{\text{ico}}} (1 + cL^{-2}). \quad (\text{B5})$$

First, we checked the effect of leading corrections to scaling. To this end we fitted our data for $D = \infty, 1.4$, and 1.0 using

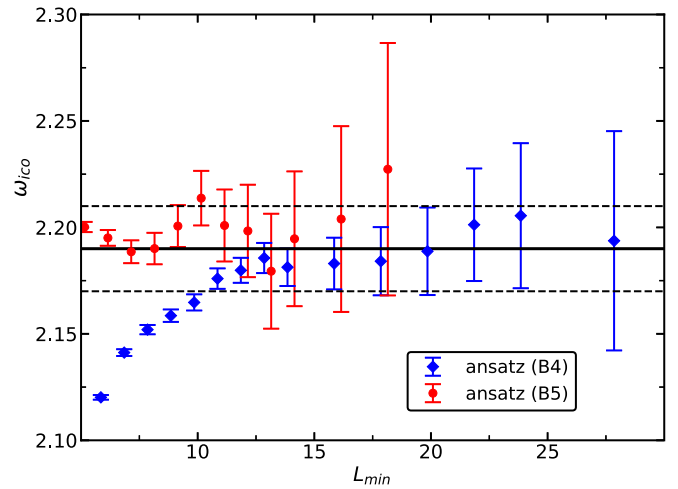


FIG. 13. We plot the estimate of ω_{ico} obtained from fitting \bar{q} at $Z_a/Z_p = 0.19477$ as a function of the minimal lattice L_{min} size that is taken into account. Data for $D = 2.1$ and 2.05 are jointly analyzed by using the *Ansätze* (B4) and (B5). For readability, the values of L_{min} are slightly shifted. The solid line gives our final estimate $\omega_{\text{ico}} = 2.19$, while the dashed lines indicate the error.

the *Ansatz* (B4) and $L_{\text{min}} = 12$ in all three cases. We get $\omega_{\text{ico}} = 2.237(13), 2.125(7)$, and $2.069(5)$ and $\chi^2/\text{d.o.f.} = 0.91, 0.78$, and 2.22 for $D = \infty, 1.4$, and 1.0 , respectively. We see a clear dependence of the result for ω_{ico} on D . Note that for both $D = \infty$ and 1.4 we get an acceptable $\chi^2/\text{d.o.f.}$, while the estimates of ω_{ico} are inconsistent.

Based on fits with the *Ansatz* (20) discussed in Sec. IV B we know that the modulus of the amplitude of leading corrections to scaling at $D = 2.05$ and 2.1 is by about a factor of 30 smaller than for $D = \infty$ or 1.4 . Therefore, the effect on the estimate of ω_{ico} should roughly be given by $[2.237(13) - 2.125(7)]/60$, which we might ignore in the following.

In Fig. 13 we plot the result of joint fits for $D = 2.05$ and 2.1 of \bar{q} at $Z_a/Z_p = 0.19477$. The free parameters are $a(D = 2.05)$, $a(D = 2.1)$, and ω_{ico} for *Ansatz* (B4). In the case of *Ansatz* (B5), c is an additional free parameter, where we assume c to be the same for $D = 2.05$ and 2.1 .

As our final estimate we take

$$\omega_{\text{ico}} = 2.19(2). \quad (\text{B6})$$

This estimate is chosen such that the estimates obtained by using the *Ansatz* (B4) for $5 \leq L_{\text{min}} \leq 9$ and the *Ansatz* (B5) for $11 \leq L_{\text{min}} \leq 20$ including the respective error bars are covered.

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