# Antiferromagnetic Ising model in small-world networks

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The antiferromagnetic Ising model in small-world networks generated from two-dimensional regular lattices has been studied. The disorder introduced by long-range connections causes frustration, which gives rise to a spin-glass phase at low temperature. Monte Carlo simulations have been carried out to study the paramagnetic to spin-glass transition, as a function of the rewiring probability p, which measures the disorder strength. The transition temperature  $T_c$  goes down for increasing disorder, and saturates to a value  $T_c \approx 1.7J$  for p > 0.4, J being the antiferromagnetic coupling. For small p and at low temperature, the energy increases linearly with p. In the strong-disorder limit  $p \rightarrow 1$ , this model is equivalent to a short-range  $\pm J$  spin glass in random networks.

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# I. INTRODUCTION

In the last few years, there has been a surge of interest in modeling complex systems as networks or graphs, with nodes representing typical system units and edges playing the role of interactions between connected pairs of units. Thus, complex networks have been used to model several types of real-life systems (social, economic, biological, technological), and to study various processes taking place on them [1-5]. In this context, some models of networks have been designed to explain empirical data in several fields, as is the case of the so-called small-world networks, introduced by Watts and Strogatz in 1998 [6].

These small-world networks are well suited to study systems with underlying topological structure ranging from regular lattices to random graphs [7,8], by changing a single parameter [9]. They are based on a regular lattice, in which a fraction p of the links between nearest-neighbor sites are replaced by new random connections, creating long-range "shortcuts" [6,9]. In the networks so generated one has at the same time a local neighborhood (as in regular lattices) and some global properties of random graphs, such as a small average topological distance between pairs of nodes. These networks are suitable to study different kinds of physical systems, as neural networks [10] and manmade communication and transportation systems [6,11,12]. The importance of a short global length scale has been emphasized for several statistical physical problems on small-world networks. Among these problems, one finds the spread of infections [13,14], signal propagation [6,15,16], random spreading of information [17-21], as well as site and bond percolation [14,22,23].

Cooperative phenomena in this kind of networks are expected to display unusual characteristics, associated to their peculiar topology [24–27]. Thus, a paramagnetic-ferromagnetic phase transition of mean-field type was found for the Ising model on small-world networks derived from one-dimensional (1D) lattices [24,28,29]. This phase transition occurs for any value of the rewiring probability p > 0, and the transition temperature  $T_c$  increases as p is raised. A similar mean-field-type phase transition was found in small-world networks generated from 2D and 3D regular lattices

[30,31], as well as for the XY model in networks generated from one-dimensional chains [32]. In recent years, the Ising model has been thoroughly studied in complex networks, such as the so-called scale-free networks, where several unusual features were observed [33-36].

Here we study the antiferromagnetic (AFM) Ising model in small-world networks generated by rewiring a 2D square lattice. One expects that the AFM ordering present in the regular lattice at low temperature will be lost when random connections are introduced, for an increasing number of bonds will be frustrated as p rises. In particular, this model includes the two basic ingredients necessary to have a spin glass (SG), namely, disorder and frustration. The former appears due to the random long-range connections introduced in the rewiring process, and the latter because half of these rewired links connect sites located in the same sublattice of the starting regular lattice.

In some spin-glass models, such as the Sherrington-Kirkpatrick model, all spins are mutually connected [37,38]. An intermediate step between these globally connected networks and finite-dimensional models consists in studying spin glasses on random graphs with finite (low) connectivity [39–42]. A further step between random graphs with finite mean connectivity and regular lattices is provided by smallworld networks, where one can modify the degree of disorder by changing the rewiring probability p. Then, for the AFM Ising model on small-world networks, we expect to find features close to those of short-range spin-glass systems. In this line, a spin-glass phase has been recently found and characterized for the AFM Ising model in scale-free networks [43]. In this paper, we employ Monte Carlo (MC) simulations to study the paramagnetic to spin-glass phase transition occurring in small-world networks. Apart from temperature and system size, another variable is the rewiring probability, which controls the degree of disorder, and allows us to interpolate from a paramagnetic-AFM transition at p =0 to a paramagnetic-SG transition in a random graph at p=1.

The paper is organized as follows. In Sec. II we describe the networks and the computational method employed here. In Sec. III we give results for the heat capacity, energy, and spin correlation, as derived from MC simulations. In Sec. IV we present and discuss the overlap parameter, transition temperature, and absence of long-range ordering. The paper closes with the conclusions in Sec. V.

#### **II. MODEL AND METHOD**

We consider the Hamiltonian:

$$H = \sum_{i < j} J_{ij} S_i S_j, \tag{1}$$

where  $S_i = \pm 1$  (i=1,...,N), and the coupling matrix  $J_{ij}$  is given by

$$J_{ij} = \begin{cases} J(>0) & \text{if } i \text{ and } j \text{ are connected,} \\ 0 & \text{otherwise.} \end{cases}$$
(2)

This means that each edge in the network is an AFM interaction between spins on the two linked nodes. Note that, contrary to the usually studied models for spin glasses, in this model all couplings are antiferromagnetic. This model with AFM couplings can be mapped onto one in which all unrewired bonds are ferromagnetic (FM), and the rewired links are 50% FM and 50% AFM. In the limit  $p \rightarrow 0$ , this mapping is the well-known correspondence between AFM and FM Ising models on bipartite lattices [44]. In the limit  $p \rightarrow 1$ , our AFM Ising model is equivalent to a spin-glass model on a random graph of mean connectivity  $\langle k \rangle = 4$ , with 50% AFM and 50% FM bonds.

Small-world networks have been built up according to the model of Watts and Strogatz [6,9], i.e., we considered in turn each of the bonds in the starting 2D lattice and replaced it with a given probability p by a new connection. In this rewiring process, one end of the selected bond is changed to a new node chosen at random in the whole network. We impose the conditions: (i) no two nodes can have more than one bond connecting them and (ii) no node can be connected by a link to itself. With this procedure we obtained networks where more than 99.9% of the sites were connected in a single component. Moreover, this rewiring method keeps constant the total number of links in the rewired networks, and the average connectivity  $\langle k \rangle$  coincides with z=4. This allows us to study the effect of disorder upon the physical properties of the model, without changing the mean connectivity.

We note that other ways of generating small-world networks from regular lattices have been proposed [22,45]. In particular, instead of rewiring each bond with probability p, one can add shortcuts between pairs of sites taken at random, without removing bonds from the regular lattice. This procedure turns out to be more convenient for analytical calculations, but does not keep constant the mean connectivity  $\langle k \rangle$ , which in this case increases with p. Spin glasses on such small-world networks, generated from a one-dimensional ring, have been studied earlier by replica symmetry breaking [46] and transfer matrix analysis [47].

From the 2D square lattice, we generated small-world networks of different sizes. The largest networks employed here included  $200 \times 200$  nodes. Periodic boundary conditions were assumed. For a given network, we carried out Monte Carlo simulations at several temperatures, sampling the spin



FIG. 1. Heat capacity per site  $c_v$  vs temperature for small-world networks generated from a 2D lattice of size  $80 \times 80$ . The plotted curves correspond to different values of the rewiring probability *p*. From top to bottom: p=0.1, 0.3, 0.5, and 1.

configuration space by the Metropolis update algorithm [48], and using a simulated annealing procedure. Several variables characterizing the considered model have been calculated and averaged for different values of p, T, and system size N. In general, we have considered 300 networks for each rewiring probability p, but we used 1000 networks to determine accurately the transition temperature from paramagnetic to SG phase. In the following, we will use the notation  $\langle \cdots \rangle$  to indicate a thermal average for a network, and  $[\cdots]$  for an average over networks with a given degree of disorder p.

## **III. THERMODYNAMIC OBSERVABLES**

The heat capacity per site,  $c_v$ , was obtained from the energy fluctuations  $\Delta E$  at a given temperature, by using the expression

$$c_v = \frac{\left[(\Delta E)^2\right]}{NT^2},\tag{3}$$

where  $(\Delta E)^2 = \langle E^2 \rangle - \langle E \rangle^2$ . We have checked that the results coincide within numerical noise with those derived by calculating  $c_v$  as  $[d\langle E \rangle/dT]/N$ . Note that we take the Boltzmann constant  $k_B = 1$ .

The temperature dependence of  $c_v$  is displayed in Fig. 1 for several values of the rewiring probability p and for networks built up from a 80×80 2D lattice. For increasing p, one observes two main features: the maximum of  $c_v$  shifts to lower T and the peak broadens appreciably. This broadening agrees with the behavior expected for systems with increasing disorder, similarly to that found for the FM Ising model in these networks [30]. However, in the AFM model, the shift of the peak to lower temperature suggests a phase transition with a temperature  $T_c$  that decreases as p is raised, contrary to the FM case, where an increase in  $T_c$  with p was observed. This difference between both Ising models on small-world networks occurs in addition to the nature of the transition itself, which in the FM case is a paramagneticferromagnetic transition vs a paramagnetic-SG transition in



FIG. 2. Minimum energy per site obtained in our simulations for the AFM Ising model on small-world networks with  $N=10^4$  nodes. The dashed line corresponds to  $e_m=(-2+2p)J$ , as explained in the text. The dotted line indicates the ground-state energy obtained in Ref. [41] for a spin glass on random networks with  $\langle k \rangle = 4$ . The solid line is a guide to the eye.

the AFM model (see below). A decrease in  $T_c$  for the AFM Ising model in this kind of networks was also suggested in Ref. [49] from the behavior of the heat capacity for several values of p.

The increase in disorder as p is raised is accompanied by an increase in frustration of the links at low temperatures. This can be quantified by the low-temperature energy of the system, which will rise as the rewiring probability is raised. To obtain insight into this energy change for p near zero (low disorder), let us remember that the square lattice is bipartite, in the sense that one can define two alternating sublattices, say A and B, so that neighbors of each node in sublattice Abelong to sublattice B, and vice versa. In the rewiring process, one introduces links between nodes in the same sublattice, and the resulting networks are no longer bipartite. However, for small rewiring probability p, we can still speak about two sublattices, with some "wrong" connections. Since each link is rewired with probability p, each connection in the starting regular lattice will be transformed into a wrong connection (of types A-A or B-B) with probability p/2. The remaining links are of A-B type, and the number of wrong connections is on average zNp/4. Then, for small p, the lowest energy can be approximated by  $E_m = -zNJ(1-p)/2$ , under the assumption that the AFM long-range ordering of the square lattice is still preserved. For z=4, we have an energy per node:  $e_m = -2(1-p)J$ . Note that for finite p the lowtemperature long-range ordering in fact decays due to the appearance of domains driven by the rewired connections (see below), but the AFM ordering is a good reference to obtain insight into the energy change as a function of rewiring probability p.

We now turn to the results for the minimum energy reached in our simulations for different p values, which are shown in Fig. 2. For rising p,  $E_m$  increases from the value corresponding to AFM ordering in the regular lattice,  $e_m$ =-zJ/2. The dashed line in Fig. 2 displays the behavior expected for small p, in the case of a strict AFM ordering on the underlaying lattice. This estimation is close to the minimum energy obtained in our simulations for p < 0.15. For larger p values, it departs appreciably from the results of the simulations, and  $e_m$  lies below the dashed line. In the limit p=1 we find a value  $e_m = -1.444(2)J$ . In this limit, our smallworld networks are very close to random networks with a Poisson distribution of connectivities, but are not identical to the latter because of the restriction that no nodes have zero links, imposed in the rewiring process [15,24]. For a  $\pm J$ Ising spin glass on random graphs with a Poisson distribution of connectivities and  $\langle k \rangle = 4$ , Boettcher [41] found a groundstate energy  $e_m = -1.431(1)J$  by using extremal optimization. This value is plotted in Fig. 2 as a dotted line close to p=1, and is near the minimum energy we found for the smallworld networks in this limit. For a more direct comparison with random networks, we have carried out some simulations for small-world networks with p=1, where we allowed the presence of isolated sites (with connectivity k=0). For the AFM Ising model in these networks, we found a minimum energy  $e_m = -1.438(2)J$ , between those of our standard networks (with minimum connectivity k=1) and random networks in Ref. [41]. Note that our error bar in  $e_m$  corresponds to a standard deviation in the distribution of minimum energy obtained for different networks. We emphasize that the energy  $e_m$  found here for each value of p is an upper limit for the lowest energy of the system.

Even though the random connections present in smallworld networks introduce disorder in the starting regular lattice, these networks still keep memory of the original bipartite lattice, but the actual meaning of the partition in two sublattices is gradually reduced as p rises. This can be measured by the number of wrong links, which amounts to a fraction p/2 of the total number of links, as indicated above. In the limit p=1, half of the links connect sites in the same original sublattice, and the memory of the partition in sublattices has completely disappeared. One can visualize the loss of AFM ordering on the 2D lattice, by plotting the spin correlation vs distance for several values of p. We define  $\xi$  as

$$\xi(r) = [\langle S_i S_j \rangle_r], \tag{4}$$

where the subscript r indicates that the average is taken for the ensemble of pairs of sites at distance r. Note that r $=d/d_0$  refers here to the dimensionless distance between sites in the starting regular lattice, not to the actual topological distance or minimum number of links between nodes in the rewired networks ( $d_0$  is the distance between nearest neighbors). The correlation  $\xi(r)$  is shown in Fig. 3 for several values of the rewiring probability p, at temperature T=1.5J. This temperature is below the critical temperature  $T_c$  of the paramagnetic-SG transition for all values of p (see below). As expected,  $\xi(r)$  decreases faster for larger p, and vanishes at p=1 for any distance  $r \ge 1$ . In general, after a short transient for small r, we find an exponential decrease of the spin correlation with the distance. This indicates that, in spite of the disorder present in the networks for p > 0, there remains some degree of short-range AFM ordering on the starting regular lattice, which is totally lost in the limit  $p \rightarrow 1$ .



FIG. 3. Absolute value of the spin correlation function vs distance in small-world networks with  $N=10^4$  nodes, at temperature T=1.5J. The dimensionless distance between nodes  $r=d/d_0$  is measured on the starting regular lattice. Data are shown for several values of p, as indicated by the labels.

### **IV. SPIN-GLASS BEHAVIOR**

#### A. Overlap parameter

As is usual in the study of spin glasses, we now consider two copies of the same network, with a given realization of the disorder, and study the evolution of both spin systems with different initial values of the spins and different random numbers for generating the spin flips [50,51]. It is particularly relevant the overlap q between the two copies, defined as

$$q = \frac{1}{N} \sum_{i} S_{i}^{(1)} S_{i}^{(2)}, \tag{5}$$

where the superscripts (1) and (2) denote the copies. Obviously, q is defined in the interval [-1, 1].

We have calculated the overlap parameter q for smallworld networks with various rewiring probabilities p, and obtained its probability distribution P(q) from MC simulations. This distribution is shown in Fig. 4 for p=0.1 and 0.5 at several temperatures. At high temperature, P(q) shows a single peak centered at q=0, characteristic of a paramagnetic state. The width of this peak is a typical finite-size effect, which should collapse to a Dirac  $\delta$  function at q=0 in the thermodynamic limit  $N \rightarrow \infty$ . When the temperature is lowered, the distribution P(q) broadens, as a consequence of the appearance of an increasing number of edges displaying frustration. At still lower temperatures, two peaks develop in P(q), symmetric respect to p=0, and characteristic of a spinglass phase [43,51,52].

Information on the "freezing" of the spins as temperature is lowered can be obtained from the evolution of the average value of |q|, for a given degree of disorder p. This average value is shown in Fig. 5 as a function of temperature for several rewiring probabilities p. It is close to zero at the high-temperature paramagnetic phase, and increases as temperature is reduced, indicating a break of ergodicity associated to the spin-glass phase [43,51]. For p=0, |q| converges



FIG. 4. Distribution of the overlap parameter q for two rewiring probabilities and various temperatures, as derived from our simulations for networks with  $N=10^4$  nodes. (a) p=0.1 at T/J=2.4, 2.1, and 1.7 and (b) p=0.5 at T/J=1.7, 1.3, and 1.

to unity at low temperatures, reflecting the AFM ordering present in the regular lattice. For increasing p, we find a decrease in the low temperature |q| values, due to an increasing degree of frustration.

#### **B.** Transition temperature

The overlap parameter q can be used to obtain accurate values of the paramagnetic-SG transition temperature, by using the fourth-order Binder cumulant [48,51]

$$g_N(T) = \frac{1}{2} \left( 3 - \frac{\left[\langle q^4 \rangle\right]_N}{\left[\langle q^2 \rangle\right]_N^2} \right).$$
(6)

This parameter can change in the interval [0,1]. One has  $g_N=0$  for a Gaussian distribution P(q) (high temperatures), and  $g_N=1$  when |q|=1 (in the particular case of a single ground state). In general  $g_N$  rises for decreasing temperature, and  $T_c$  can be obtained from the crossing point for different network sizes N. As an example, we present in Fig. 6  $g_N(T)$  as a function of temperature for several system sizes and a rewiring probability p=0.05. From the crossing point we find  $T_c/J=2.175\pm 0.005$  for this value of p.



FIG. 5. Average of the absolute value of the overlap parameter  $[\langle q \rangle]$  for various rewiring probabilities *p* and several temperatures. Symbols are data points derived from MC simulations on networks of size  $N=10^4$ .

By using this procedure, we have calculated the transition temperature  $T_c$  for several values of p, and the results so obtained are shown in Fig. 7. For small p,  $T_c$  decreases linearly from the transition temperature corresponding to the AFM model on the 2D square lattice, and for p > 0.4 it saturates to a value of about 1.7J. Close to p=0, we find a change in  $T_c$  induced by the long-range links:  $T_c=T_c^0-apJ$ , where  $T_c^0$  is the paramagnetic-AFM transition temperature in the square lattice and  $a \approx 2$ .

It is interesting the approximately linear decrease in  $T_c$  for increasing p up to  $p \approx 0.3$ . This decrease could be expected from the larger number of frustrated links appearing as p is raised. For increasing frustration, the paramagnetic phase is favored, and the spin glass appears at lower temperature ( $T_c$ is reduced). This change of  $T_c$  as a function of rewiring probability could be also expected from the behavior of the heat



FIG. 6. Fourth-order Binder's cumulant  $g_N$  as a function of temperature for small-world networks generated from 2D square lattices with rewiring probability p=0.05. Symbols represent different system sizes  $N=L^2$ : squares, L=60; circles, L=80; triangles, L=100; diamonds, L=150. Error bars are on the order of the symbol size.



FIG. 7. Transition temperature  $T_c$  as a function of the rewiring probability p for small-world networks. Error bars are on the order of the symbol size. The dashed line is a guide to the eye.

capacity shown in Fig. 1. For p > 0.4, small-world networks behave in this respect similarly to Poissonian random networks, in the sense that the transition temperature is roughly independent of p, and is close to that found for the strong-disorder limit p=1.

#### C. Absence of long-range AFM ordering

Even though all data indicate that the AFM Ising model in small-world networks yields a spin-glass phase at low temperature, one can ask if such disordered phase appears for any finite value of the rewiring probability. One could argue that some residual long-range AFM ordering could be present for finite but small p values. From our considerations in the preceding sections, one could think that, for small p, the low-temperature phase still keeps the long-range ordering characteristic of the 2D regular lattice, with some defects caused by the long-range connections. To analyze this question, we consider a staggered magnetization defined for the square lattice as usual:

$$M_s = M_A - M_B, \tag{7}$$

with

$$M_A = \sum_{i \in A} S_i, \tag{8}$$

and similarly for  $M_B$ . Our question then refers to the possibility of a finite value for  $M_s$  for small-world networks with p > 0.

To check this point, we have carried out simulations starting from an AFM ordered configuration and followed the evolution of  $M_s$ . We prefer this procedure to directly calculating the low-temperature staggered magnetization from simulated annealing, since in this case a long-range ordering can be difficult to find for large networks, due to the appearance of different spin domains. Thus, we analyzed the decay of  $M_s$  at temperatures lower than the transition temperature  $T_c$  for different system sizes. In particular, in Fig. 8 we show the relaxation of  $|M_s|$  on networks with p=0.1 at a tempera-



FIG. 8. Relaxation of the staggered magnetization on the underlying regular lattice for small-world networks with p=0.1, from simulations starting from an ordered AFM configuration. The plot shows the evolution of  $|M_s|$  at T=1.7J and different system sizes  $L \times L$ . From top to bottom: L=40, 60, 80, 100, 150, and 200.

ture T=1.7J, well below the transition temperature for this rewiring probability  $(T_c/J=2.10\pm0.01)$ . For each system size,  $|M_s|$  decreases from unity to reach a plateau at a finite value, which is clearly a finite-size effect, as seen in the figure. As the system size increases, such a plateau appears after longer simulation times, and the corresponding value of  $|M_s|$  decreases. These results are consistent with the decay of the spin correlation  $\xi(r)$  at temperatures below  $T_c$ , as presented in Fig. 3 for several values of p.

For p < 0.1, a relaxation of  $M_s$  is expected to appear for larger system sizes and longer simulation runs. Everything indicates that at low T the long-range ordering disappears in the thermodynamic limit  $N \rightarrow \infty$  for any p > 0. This is in line with earlier results for the FM Ising model on this kind of networks, in the sense that the paramagnetic-FM transition occurring in those systems changes from an Ising-type transition at p=0 to a mean-field-type one (typical of random networks) for any finite value of the rewiring probability p > 0 [24]. The observation of this mean-field character for the paramagnetic-FM transition requires system sizes that increase as the rewiring probability is lowered [30], similarly to the decay of the staggered magnetization in the AFM case shown here.

### V. CONCLUSIONS

The combination of disorder and frustration in the AFM Ising model on small-world networks gives rise to a spinglass phase at low temperatures. The transition temperature from a high-temperature paramagnet to a low-temperature spin-glass phase goes down for increasing disorder, and saturates to a value  $T_c \approx 1.7J$  for p > 0.4.

The overlap parameter provides us with clear evidence of the frustration associated to the spin-glass phase at low temperatures. The degree of frustration increases as the disorder (or rewiring probability) rises. For small rewiring probability p, the energy of the ground state increases linearly with p up to  $p \approx 0.15$ , and for larger p, it converges to  $e_m = -1.44J$ . In the limit  $p \rightarrow 1$  one recovers the behavior of a  $\pm J$  Ising spin glass in random networks. An interesting feature of the physics here is that a small fraction of random connections is able to break the long-range AFM ordering present in the 2D square lattice at low temperature.

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