Recent Developments in Ising Spin Glasses

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We have studied the three-dimensional Ising spin glass with a $\pm J$ distribution by Monte Carlo simulations. Using larger sizes and much better statistics than in earlier work, a finite size scaling analysis shows quite strong evidence for a finite transition temperature, T_c, with ordering below T_c. Our estimate of the transition temperature is rather lower than in earlier work, and the value of the correlation length exponent, ν, is somewhat higher. Because there may be (unknown) corrections to finite size scaling, we do not completely rule out the possibility that $T_c = 0$ or that T_c is finite but with no order below T_c. However, from our data, these possibilities seem less likely.

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1 Introduction

Spin glasses are random magnetic systems, where the disorder is sufficiently severe that one has "frustration", *i.e.* no spin configuration simultaneously minimizes each term in the Hamiltonian. Frustration leads to a complicated energy landscape with many local minima separated by barriers. As the temperature is lowered, relaxation times become very long, because the probability of overcoming a barrier by thermal fluctuation is low. Even in the presence of disorder and frustration, it is possible to have a sharp second order transition, with a divergent correlation length, and all the usual critical phenomena, such as exponents. The question of whether such a transition actually does occur at a finite temperature, T_c, in an Ising spin glass in three dimensions has aroused a lot of interest for the last two decades [1], and the consensus of opinion has changed several times. About one decade ago, several pieces of work [2, 3, 4, 5] seemed to show that there is a finite T_c, and this conclusion has generally been restated since then[6]. However, on closer inspection, the case is not completely closed. For example, the work of one of us [2], henceforth referred to as BY, is unable to rule out the possibility that $T_c = 0$ and the correlation length, ξ, diverges *exponentially* as $T \to 0$, as happens in the two-dimensional Heisenberg ferromagnet. The data is also consistent with a line of critical points terminating at $T_c \simeq 1.2$, as occurs in the Kosterlitz-Thouless-Berezinskii theory of the

two-dimensional XY ferromagnet. In this scenario there would be no long range spin glass order below T_c. Furthermore, recent results of Marinari et al. [7] were found to be consistent *both* with a finite T_c and with a zero temperature transition where the correlation length diverges exponentially, $\xi \sim \exp(A/T^4)$. We therefore feel there are three possible behaviors: T_c is finite and there is spin glass order at lower temperatures, T_c is finite but there is a line of critical points (i.e. no spin glass order) at lower temperatures, and $T_c = 0$ and the correlation length diverges exponentially as $T \to 0$. During the last decade available computer power has increased enormously so, given these uncertainties, it is useful to look at the problem again. The calculations presented here are similar to those of BY, but we are able to study larger system sizes in the temperature range of interest and obtain *much* better statistics by averaging over many more samples. As a result, unlike BY, we are able to see clear evidence for *ordering below a finite* T_c.

We study the standard Edwards-Anderson[8] hamiltonian,

$$\mathcal{H} = -\sum_{\langle i,j \rangle} J_{ij} S_i S_j \ , \tag{1}$$

where the spins S_i take values ± 1, and the nearest neighbor interactions, J_{ij} take values ± 1 with equal probability. The simple cubic lattice contains $N = L^3$ spins and has periodic boundary conditions. In some previous work, $\{J_{ij}\}$ was generated so that the the number of ferromagnetic couplings is exactly the same as that of antiferromagnetic couplings. We do not impose such a condition in the present work.

2 Numerical Techniques

The Monte Carlo simulation uses a multispin coding technique [9] in which each spin and bond is represented by a single bit of a computer word. On a 32 bit machine we then flip in parallel 32 spins (on the same lattice site but in different samples with different realizations of the disorder). For this method to be efficient the *same* random number is used for each bit. We use a shift register random number generator [10, 11], commonly known as R250. The code runs at 27 million spin updates per second on one node (IBM 390 RISC workstation) of the SP2 computer at the Maui High Performance Computing Center. Since we need many more than 32 samples, we ran the same code independently on many nodes at the same time. Each node produces its own output file from which the final averaging is easily done using a unix shell script. Monte Carlo simulations of random systems thus provide an example where parallel computing can be done in a trivial (and almost perfectly efficient!) way. The total CPU time used for the data presented here is about 9 node–years.

To get good statistics we average over a large number of samples, N_s, where for each size, N_s is at least the value in the third column in Table 1. After t_0

L	largest t_0	minimum N_s
6	4×10^5	8192
8	1×10^6	8192
12	8×10^6	6880
16	15×10^6	3392
24	5×10^6	2080

Table 1: For each size, L, we show the largest value of t_0, (where, as explained in the text, the simulation ran for $3t_0$ sweeps) and the minimum number of samples, N_s.

sweeps for equilibration, an additional $2t_0$ sweeps are carried out for measurements. For each size, the largest value of t_0 used is also shown in Table 1 (this is for the lowest temperature: at higher temperatures many fewer sweeps are generally needed).

3 Finite Size Scaling

As usual [2], for each realization of the bonds, two copies of the system are studied with different initial values of the spins and different random numbers for generating the spin flips. Of particular importance is the overlap between the two copies,

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

where the labels "1" and "2" denote the copies. From measurements of q we compute the Binder ratio [12, 2]

$$g = \frac{1}{2} \left[3 - \frac{\langle q^4 \rangle}{\langle q^2 \rangle^2} \right] \ , \tag{3}$$

where the average $\langle \ldots \rangle$ denotes *both* a thermal average for a given set of bonds and an average over the disorder [13]. At high temperature, $g \to 0$, whereas $g \to 1$ in the spin glass phase, at least if there is a unique thermodynamic state.

Because g is dimensionless it has the finite size scaling form [2]

$$g = \tilde{g}(L^{1/\nu}(T - T_c)) \tag{4}$$

and so is *independent of* L at T_c. If there is ordering below of finite T_c, the curves for g will intersect at T_c and splay out again at lower T (with the larger sizes having the larger values, the opposite of the situation above T_c). This is the signature that we are looking for.

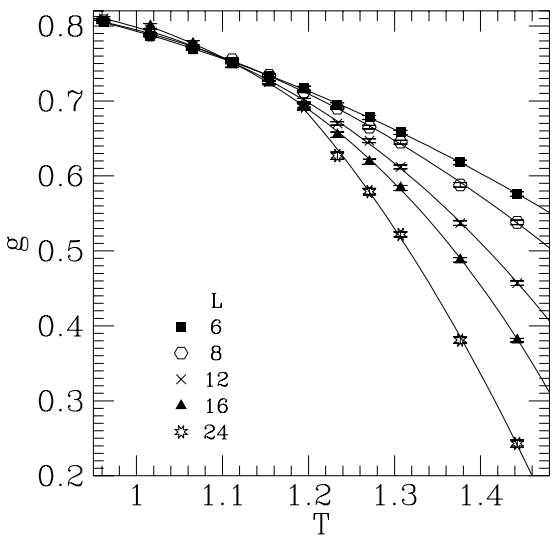

Figure 1: Results for the Binder ratio g, defined in Eq. (3), for different sizes and temperatures. The lines are smooth curves through the data and are only intended as guides to the eye.

In addition to g, we also computed the spin glass susceptibility,

$$\chi_{SG} = N\langle q^2 \rangle \ , \tag{5}$$

and $P(q)$, the distribution of q. These have the finite size scaling forms,[2]

$$\chi_{SG}(L^{1/\nu}(T - T_c)) \ , \tag{6}$$

and

$$P(q) = L^{\beta/\nu}\tilde{P}(L^{\beta/\nu}q,\ L^{1/\nu}(T - T_c)) \ , \tag{7}$$

where β is the order parameter exponent and is related to η, which gives the power law decay of the correlations at the critical point, by

$$\frac{\beta}{\nu} = \frac{1}{2}(d - 2 + \eta) \ . \tag{8}$$

It is very important to ensure that enough Monte Carlo sweeps have been carried out to equilibrate the sample. Following BY we compare the results for g obtained, as described above, from the overlap between two replicas with the results obtained from one replica at two different times (see BY for details). BY found that these two estimates approach the equilibrium value from opposite directions as the length of the simulation increased. Once the two values agree,

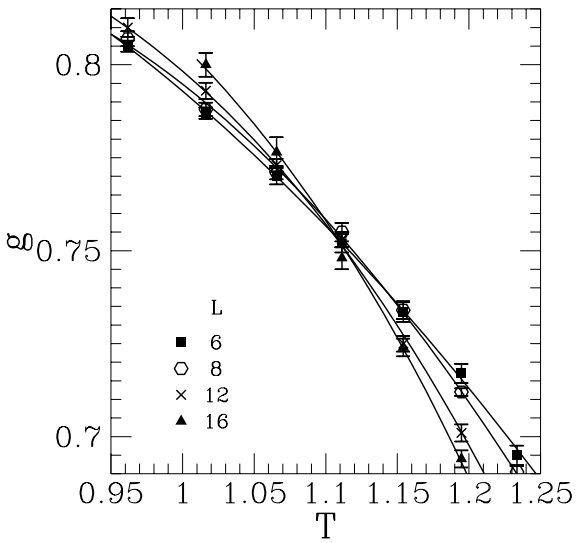

Figure 2: An enlarged view of the data in Fig. 1 in the crucial region where the curves come together.

they do not change further if more sweeps are carried out. We have also tested this by doing the run for $L = 8, T = 0.9618$ for an order of magnitude longer time than needed for the two estimates to agree. Again we find that there is no subsequent change within our (much smaller) errors.

4 Results

Our data for g is shown in Fig. 1 and an enlarged view of the region where the curves for different sizes intersect is shown in Fig. 2. From Fig. 2 one sees clear evidence for splaying out of the data below a temperature of about 1.10. Estimating T_c to be approximately 1.10 from the intersection point we can scale most of the data according to Eq. (4) with $\nu = 2.0$, see Fig. 3. The only point which does not lie on a common curve is the result for $L = 24, T = 1.1948$, which is significantly higher. One can see from Fig. 1 that this point has almost the same value of g as the data for $L = 16$ at the same temperature. This data point being rather higher than expected *may* reflect corrections to finite size scaling, and indicate that the true critical temperature is higher than the straightforward estimate based on data for g with $L \leq 16$.

Once T_c has been estimated, one can obtain β/ν, or equivalently η, from the expected scaling form of $P(q)$ at criticality given by Eq. (7) with $T = T_c$. The corresponding plot is shown in Fig. 4 for $T = 1.1113$ (well within the bounds

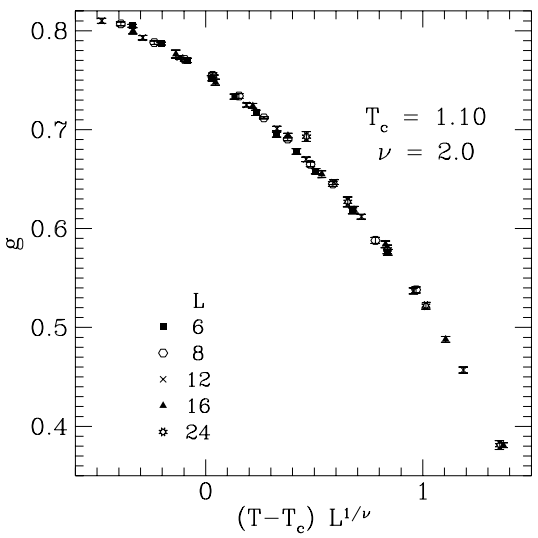

Figure 3: A scaling plot for g according to the form in Eq. (4).

of our estimate of T_c), and has $\beta/\nu = 0.3$ which corresponds to $\eta = -0.4$ from Eq. (8) with $d = 3$.

We have also performed finite size scaling plots for χ_{SG} according to Eq. (6). This data does not locate T_c precisely, so we have used the same T_c as obtained from the scaling plot for g in Fig. 3, i.e. $T_c = 1.10$. Furthermore the value of η is constrained by requiring that the data scales at T_c and from Fig. 4 this gives $\eta = -0.4$ The only remaining parameter is ν and the best fit, shown in Fig. 5, is for $\nu = 1.6$.

The values for ν obtained from g and χ_{SG} are somewhat different. If we try to use $\nu = 2.0$ in the data for χ_{SG} or $\nu = 1.6$ in the data for g, the fit is visibly worse. Presumably this difference indicates that corrections to finite size scaling are not negligible for the range of sizes that we can study. Taking into account all the data we estimate

$$\begin{aligned} T_c &= 1.11 \pm 0.04 \\ \nu &= 1.7 \pm 0.3 \\ \eta &= -0.35 \pm 0.05 \ . \end{aligned} \tag{9}$$

As discussed above, the $L = 24$ data indicates that T_c may be higher than that estimated from the intersections of g for $L \leq 16$. This is reflected in the estimated error for T_c in Eq. (9). The estimated errors in ν and η then come largely from the uncertainty in T_c. Our value of T_c is rather lower than earlier estimates which were close to 1.2, and the value of ν is higher, previous estimates

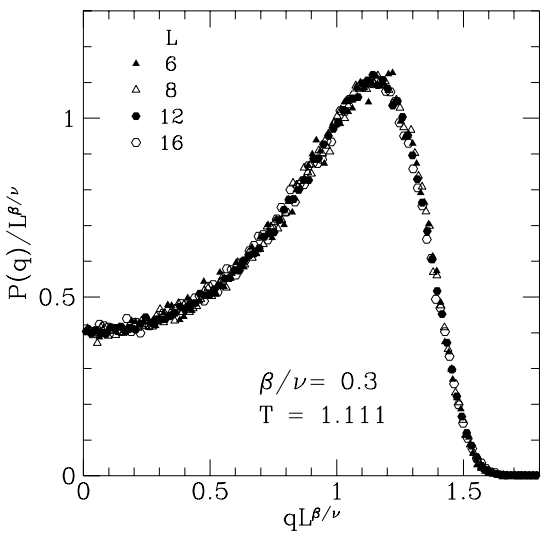

Figure 4: A scaling plot for $P(q)$ at $T = 1.1113$ (which is close to the the critical point) according to the form in Eq. (7). According to Eq. (8), the value $\beta/\nu = 0.3$ corresponds to $\eta = -0.4$.

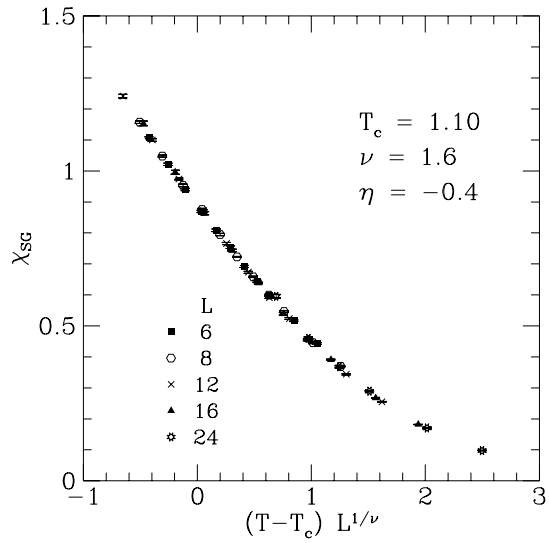

Figure 5: A scaling plot for χ_{SG} according to the form in Eq. (6).

generally being in the vicinity of 1.3. Our value of η is not very different from

earlier estimates.

5 Conclusions

Powerful computing facilities are necessary for spin glass simulations because relaxation times become very long for larger sizes as one approaches a critical point or the temperature becomes low. We have seen that a computer environment consisting of, essentially, a large number of powerful workstations, provides an excellent environment for Monte Carlo simulations of such systems. Parallelism can be achieved with 100% efficiency and it is not even necessary to use message passing libraries. Since the present study used a substantial computer effort, an investigation of larger sizes, which is necessary to confirm our results beyond reasonable doubt, may need a better algorithm than single spin flip Monte Carlo. There are already some promising results from the "replica exchange" method [14] (where, in addition to local moves, global moves are made which cause the temperature of the system to cycle up and down).

We have found evidence for a finite transition temperature with spin glass order below T_c. However, it is difficult to estimate the size of systematic errors, such as possible correlations in the random numbers (though we believe that these are very small[11]), and corrections to finite size scaling. Because of this, and because the crossing of the data for g that we observe in Fig. 2 is rather small, we cannot rule out, *for sure*, the other two possibilities, T_c is finite but there is no spin glass order at lower temperature, and $T_c = 0$. However, from our data, these possibilities now seem less likely.

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[10] The R250 random number generator keeps a table of random integers and generates a new number, X_n, from $X_n = X_{n-103}$ .XOR .X_{n-250}. For most of the runs we used 16 bit random numbers, and so could gain a bit of extra speed by extracting 2 random numbers from a single word. Each bit is separately initialized by a linear congruential generator. In the multispin coding method[9] one generates tables of length 2^{16} (the length is 2^n for random numbers of n bits) from which the flipping probabilities, $\exp(-\Delta E/T)$, are represented as "integer" divided by 2^{16} where $0 \leq$ "integer" $\leq 2^{16}$. Thus, only for certain temperatures can the flipping probabilities be represented exactly. We chose temperatures where the probability is exact for $\Delta E = 4J$ and $8J$ but only for $T = 1.15416$ is the probability for $\Delta E = 12J$ represented exactly with 16 bit random numbers. For the two lowest temperatures, $T = 1.01633$ and 0.96180, the flipping probability for $\Delta E = 12J$ would have been rounded down to zero. To avoid this we use 18 bit random numbers (and hence tables of length 2^{18}) for these temperatures.

[11] As a partial check of the random number generator, we also did high precision runs for $L = 6, 8$ and 12 at the lowest temperature, $T = 0.96180$, using a stride of 1279, rather than 250, and with multiply, rather than XOR, since this generator is better for the Metropolis algorithm (P. Coddington, private communication). We find that the results are unchanged within statistical errors; in particular we still find a statistically significant increase in g with L.

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[13] In a random systems one can define the Binder ratio in different ways. Another possibility would be to determine the ratio of moments for a single sample and then average this over samples. However, there are systematic errors in evaluating the ratio. Since we obtain only a few statistically independent measurements for each sample, these systematic errors would be large and uncontrolled if the Binder ratio were determined in this way.

[14] K. Hukushima, K. Nemoto and H. Takayama, (these proceedings).