

# Histogram Monte Carlo Position-Space Renormalization Group: Applications to the Site Percolation

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We study site percolation on the square lattice and show that, when augmented with histogram Monte Carlo simulations for large lattices, the cell-to-cell renormalization group approach can be used to determine the critical probability accurately. Unlike the cell-to-site method and an alternate renormalization group approach proposed recently by Sahimi and Rassamdana, both of which rely on *ab initio* numerical inputs, the cell-to-cell scheme is free of prior knowledge and thus can be applied more widely.

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**KEY WORDS:** Percolation; renormalization group; Monte Carlo; critical point.

## 1. INTRODUCTION

The problem of percolation has been a subject of much contention in recent years.<sup>(1,2)</sup> Percolation has been investigated using a variety of approaches, including series expansions,<sup>(3)</sup> Monte Carlo simulations,<sup>(4-7)</sup> position-space renormalization group analyses,<sup>(8,9)</sup> histogram Monte Carlo renormalization group studies,<sup>(10-15)</sup> and conformal invariance analyses.<sup>(16)</sup> In most of these investigations one invariably considers the evaluation of  $E(L, p)$ , the existence probability<sup>(11-15)</sup> that the system percolates, as a function of  $L$ , the linear dimension of the lattice, and  $p$ , the site (or bond) occupation probability. The existence probability has also been termed the crossing probability by Kesten<sup>(1)</sup> and the spanning probability by Ziff.<sup>(6)</sup> In the limit of  $L \rightarrow \infty$ ,  $E(L, p)$  approaches a step function  $\theta(p - p_c)$ ,<sup>(2)</sup> where  $p_c$  is the critical probability. For the square lattice with free boundary

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conditions,  $E(\infty, p_c)$  is given by a universal constant  $\alpha_c = 1/2$  for both bond and site percolation.<sup>(6)</sup>

Consider the determination of  $p_c$  for site percolation on the square lattice. Although early attempts using series expansion<sup>(3)</sup> yielded the value  $p_c = 0.593 \pm 0.02$ , the best value to date is that of Ziff<sup>(6)</sup>, who used extensive Monte Carlo simulations to arrive at  $p_c = 0.592\,7460 \pm 0.000\,0005$ . In addition, the scaling behavior of  $p_c$  has been found to assume the form

$$p_c(L) - p_c \sim L^{-1/\nu} \quad (1)$$

under a cell-to-site renormalization group scheme,<sup>(8,9)</sup> where  $\nu = 4/3$  is the correlation exponent, and  $p_c(L)$  is the fixed point solved from the transformation equation

$$p_c(L) = E[L, p_c(L)] \quad (\text{cell-to-site}) \quad (2)$$

Similarly, under a cell-to-cell renormalization group scheme, the scaling behavior is found to be<sup>(6)</sup>

$$p_c(L) - p_c \sim L^{-1-1/\nu} \quad (3)$$

with  $p_c(L)$  solved from

$$E[L - 1, p_c(L)] = E[L, p_c(L)] \quad (4)$$

In addition, Ziff<sup>(6)</sup> has proposed an alternate approach based on the scaling relation (3) and the solution of the equation

$$E[L, p_c(L)] = \alpha \quad (5)$$

where the value of  $\alpha = \alpha_c = 0.5$  was used. Very recently, Sahimi and Rassamdana (SR)<sup>(17)</sup> showed that the usefulness of (5) can be extended to any value of  $0 < \alpha < 1$ . For this reason we shall refer to (5) in the context of general  $\alpha$  as the SR equation. In discussions in refs. 6 and 17, however, the primary purpose was the determination of the scaling behavior using data for relatively small values of  $L \leq 7$ . It is also necessary in their analyses that specific values of the exponents are to be used.

The purpose of this paper is twofold. First, we carry out a cell-to-cell renormalization group scheme proposed recently by one of us<sup>(13)</sup> for large lattices with  $L \leq 512$ , a process made possible by using histogram Monte Carlo simulations. Second, we show that the cell-to-cell scheme is fundamentally more useful. Using the same Monte Carlo data, we show that the cell-to-cell approach determines  $p_c$  accurately, and that the determination is independent of the value of the scaling exponent used in the

extrapolation. Thus, the cell-to-cell approach is more useful and capable of a wider range of applications.

## 2. HISTOGRAM MONTE CARLO SIMULATIONS

The histogram Monte Carlo renormalization approach<sup>(11)</sup> to site percolations has been described in recent papers.<sup>(13-15)</sup> Here, for completeness and to make our presentation self-contained, we briefly sketch the main ideas. Consider site percolation on a lattice  $G$  of  $N$  sites, with each site either occupied with a probability  $p$  or empty with a probability  $1-p$ . Two neighboring sites belong to a cluster if both are occupied. The occupied sites form subgraphs  $G' \subset G$  consisting of clusters of sites. Let  $v(G')$  be the number of occupied sites in  $G'$ . The probability of the appearance of a particular subgraph  $G'$  is

$$\pi(G', p) = p^{v(G')} (1-p)^{N-v(G')}, \quad 0 \leq v(G') \leq N \quad (6)$$

A cluster is percolating if it spans across opposite borders of  $G$ . A percolating subgraph, denoted by  $G'_{\text{per}}$ , is one whose largest cluster is percolating. The existence probability  $E(L, p)$  for site percolation on a square lattice  $G$  of size  $L \times L$  with free boundaries is then the summation of the probability (6) over all percolating clusters, namely,

$$E(L, p) = \sum_{G'_{\text{per}} \subseteq G} \pi(G'_{\text{per}}, p) = \sum_{v=0}^N p^v (1-p)^{N-v} M(v) \quad (7)$$

where  $M(v)$  is the number of percolating subgraphs of  $G$  for a fixed  $v$ . This latter expression permits one to evaluate the existence probability  $E(L, p)$  numerically.

We evaluate (7) by computing  $M(v)$  from histogram Monte Carlo simulations. One first chooses  $w$  distinct  $p$  values, and for each of the  $p$  values generates  $N_R$  subgraphs at random. For each of the  $wN_R$  subgraphs thus generated, one counts  $v$ , the number of occupied sites, and checks whether the subgraph is percolating. In this way, one obtains  $N_{\text{per}}(v)$ , the number of percolating subgraphs, and  $N_{\text{np}}(v)$ , the number of nonpercolating subgraphs, for each given  $v$ . The central idea of ref. 13 is to approximate  $M(v)/\binom{N}{v}$ , the actual fraction of subgraphs that are percolating, by  $N_{\text{per}}(v)/[N_{\text{per}}(v) + N_{\text{np}}(v)]$ , the fraction of subgraphs that are percolating in the  $wN_R$  subgraphs generated in the simulations. Then (7) becomes

$$E(L, p) \approx \sum_{v=0}^N p^v (1-p)^{N-v} \binom{N}{v} \left[ \frac{N_{\text{per}}(v)}{N_{\text{per}}(v) + N_{\text{np}}(v)} \right] \quad (8)$$

This equation forms the basis of our numerical analysis.

### 3. THREE DIFFERENT CALCULATION SCHEMES AND NUMERICAL ANALYSES

We use the cell-to-cell renormalization group transformation<sup>(11, 12)</sup>

$$E(L/2, p') = E(L, p) \tag{9}$$

connecting two cells of linear sizes  $L$  and  $L/2$ . This transformation gives the renormalized occupation probability  $p'$  as a function of  $p$ . The fixed point  $p_c(L)$  of (9), obtained by solving the equation

$$E[L/2, p_c(L)] = E[L, p_c(L)] \quad (\text{cell-to-cell}) \tag{10}$$

gives an estimate of the critical probability  $p_c(L)$  for each value of  $L$ .

We have carried out histogram Monte Carlo simulations as described in Section 2 with  $w \sim 400$  and  $N_R \sim 10^5 - 10^6$  for systems of sizes  $L = 32, 64, 128, 256,$  and  $512$ . The data are then applied to calculate  $p_c(L)$  using the three calculations schemes described by (10), (2), and (5), namely the cell-to-cell, cell-to-site and SR approaches. Results obtained from (10) and (2) are listed in Table I, and results from (5) are listed in Table II for three different values of  $\alpha = 0.1, 0.5,$  and  $0.9$ . These results are further plotted in Figs. 1–3 by assuming the scaling behavior

$$p_c - p_c(L) \sim L^{-c} \tag{11}$$

for  $c = 0.75, 1.00,$  and  $1.75$ . The value of  $p_c = p_c(\infty)$  is then extrapolated by least-square fits in each of the cases. Results are shown in Table III.

Table III show clearly that, among the three schemes, the cell-to-cell scheme leads to the most accurate determination of  $p_c$ . More importantly, numbers in the first line of Table III show that the determination of  $p_c$  using the cell-to-cell scheme is *insensitive* to the value of the scaling power  $c$  in the scaling relation (11). This is due to the relatively large values of  $L$

**Table I. Values of  $p_c(L)$  Under the Cell-to-Cell and Cell-to-Site Schemes<sup>a</sup>**

$L$	Cell-to-cell	Cell-to-site
32	$0.59287 \pm 0.00054$	$0.60257 \pm 0.00019$
64	$0.59288 \pm 0.00030$	$0.59845 \pm 0.00007$
128	$0.59273 \pm 0.00018$	$0.59607 \pm 0.00006$
256	$0.59291 \pm 0.00024$	$0.59476 \pm 0.00008$
512	$0.59283 \pm 0.00014$	$0.59397 \pm 0.00005$

<sup>a</sup> Numbers for cell-to-cell are solved from (10) and cell-to-site from (2).

**Table II. Values of  $p_c(L)$  Under the SR Scheme Solved from (5)**

$L$	$\alpha = 0.1$	$\alpha = 0.5$	$\alpha = 0.9$
32	$0.54217 \pm 0.00019$	$0.59264 \pm 0.00017$	$0.64138 \pm 0.00017$
64	$0.56263 \pm 0.00010$	$0.59273 \pm 0.00007$	$0.62215 \pm 0.00008$
128	$0.57484 \pm 0.00011$	$0.59273 \pm 0.00006$	$0.61040 \pm 0.00007$
256	$0.58219 \pm 0.00009$	$0.59280 \pm 0.00009$	$0.60336 \pm 0.00005$
512	$0.58648 \pm 0.00008$	$0.59280 \pm 0.00007$	$0.59910 \pm 0.00006$

used in the extrapolation. In contrast, numbers in the second line from the cell-to-site scheme indicate that it works well only when one takes  $c = 1/\nu = 0.75$ , and the last three lines indicate that the SR scheme works well only when  $\alpha$  is taken to assume the critical value  $\alpha_c = 0.5$ . Thus, both of the latter approaches rely on some form of *ab initio* input.

These situations are also illustrated in Figs. 1–3. In Fig. 1, where one takes  $c = 1/\nu = 0.75$ , it is seen that all three schemes work almost equally well, and in Figs. 2 and 3, where  $c = 1.0$  and  $1.75$ , respectively, it is seen that only the cell-to-cell scheme and the SR scheme with the special input value of  $\alpha = 0.5$  yield good results.

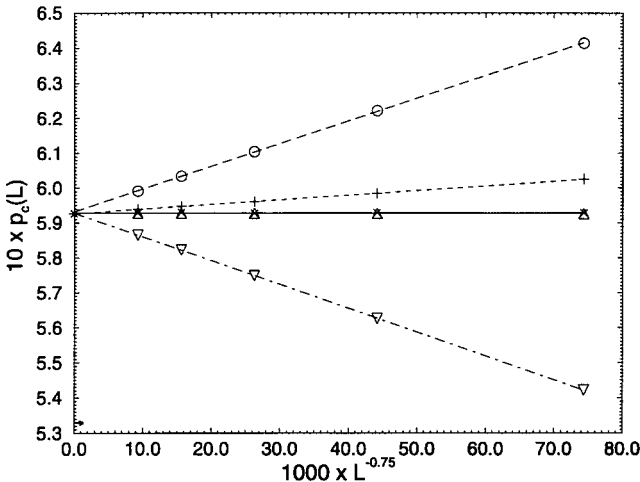


Fig. 1. Plot of  $p_c(L)$  as a function of  $L^{-0.75}$ . Data points are those obtained using the cell-to-cell scheme ( $\times$ ), the cell-to-site scheme ( $+$ ), and the SR scheme ( $\Delta$ ,  $\alpha = 0.5$ ;  $\nabla$ ,  $\alpha = 0.1$ ;  $\circ$ ,  $\alpha = 0.9$ ). The star ( $*$ ) indicates the value  $p_c = 0.592746$  determined by Ziff.<sup>(6)</sup> Straight lines represent least square fits, and intersections of the straight lines with the y axis give the values of  $p_c$  listed in Table III.

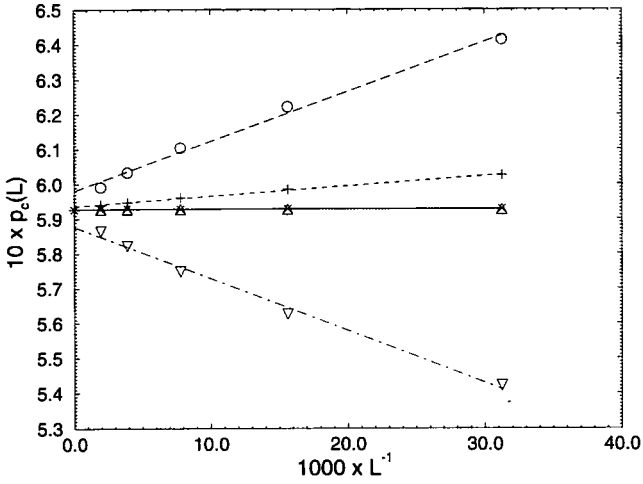


Fig. 2. Plot of  $p_c(L)$  as a function of  $L^{-1.00}$ . Data points and notations are the same as those in Fig. 1.

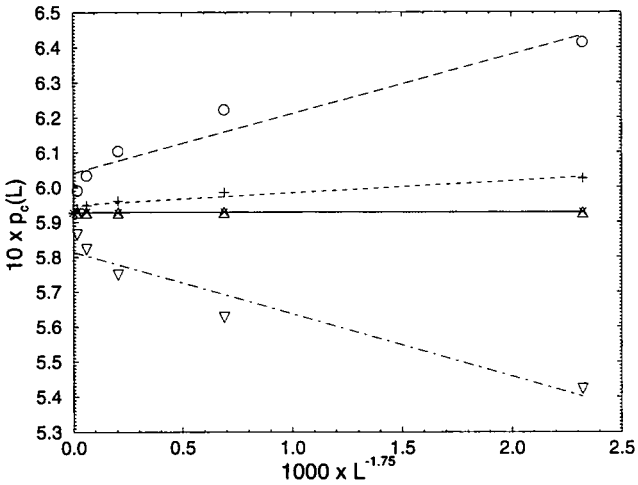


Fig. 3. Plot of  $p_c(L)$  as a function of  $L^{-1.75}$ . Data points and notations are the same as those in Fig. 1.

**Table III. The Critical Probability  $p_c$  Extrapolated from Data of Tables I and II by Assuming the Scaling Relation  $L^{-c}$** 

$c$	0.75	1.00	1.75
Cell-to-cell	$0.59282 \pm 0.00006$	$0.59283 \pm 0.00005$	$0.59283 \pm 0.00004$
Cell-to-site	$0.59267 \pm 0.00006$	$0.59364 \pm 0.00016$	$0.5949 \pm 0.0005$
SR, $\alpha = 0.1$	$0.59279 \pm 0.00004$	$0.58772 \pm 0.00115$	$0.5814 \pm 0.0030$
SR, $\alpha = 0.5$	$0.59283 \pm 0.00002$	$0.59281 \pm 0.00002$	$0.59278 \pm 0.00002$
SR, $\alpha = 0.9$	$0.59322 \pm 0.00014$	$0.59807 \pm 0.00120$	$0.6041 \pm 0.0030$

#### 4. CONCLUSION AND DISCUSSION

We have considered the site percolation problem on the square lattice using the cell-to-cell renormalization group schemes, as well as an alternate scheme proposed by Ziff and extended recently by Sahimi and Rassamdana (SR). The transformation relation determining  $p_c(L)$  is solved in each case using data from histogram Monte Carlo simulations for cells of size  $L \times L$ , with  $L = 2^l$ ,  $l = 5, 6, \dots, 9$ . The results are then used to extrapolate the critical probability  $p_c$ . Our results show that all three schemes can be used to determine  $p_c$  with almost the same degree of accuracy. However, the cell-to-site scheme makes use of the prior knowledge of the exponent  $-1/\nu = -3/4$  of the scaling relation, and the SR scheme, while of a lesser dependence on the exponent used, requires the use of the universal constant  $\alpha_c = 1/2$  in the determining equation. The cell-to-cell scheme, in contrast, does not require any *ab initio* input.

Finally, we remark that, being insensitive to the scaling power used in the extrapolation of  $p_c$ , the present cell-to-cell scheme, which works well for systems with relatively large  $L$ , does not settle a recent controversy on the precise value of the scaling power.<sup>(18, 19)</sup> Instead, the present scheme's usefulness rests on precisely this insensitivity so that it can be applied to other systems when the scaling power is not known.

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## REFERENCES

1. H. Kesten, *Percolation Theory for Mathematicians* (Birkhauser, Boston, 1982).
2. D. Stauffer and A. Aharony, *Introduction to Percolation Theory*, 2nd ed. (Taylor and Francis, London, 1992).
3. M. F. Syskes, D. S. Gaunt, and M. Glen, *J. Phys. A: Math. Gen.* **9**:L97 (1976).
4. T. Gebele, *J. Phys. A: Math. Gen.* **17**:L51 (1984).
5. F. Yonezawa, S. Sakamoto, and M. Hori, *Phys. Rev.* **B40**:636 (1989).
6. R. M. Ziff, *Phys. Rev. Lett.* **69**:2670 (1992).
7. R. P. Langlands, C. Pichet, Ph. Pouliot, and Y. Saint-Aubin, *J. Stat. Phys.* **67**:553 (1992).
8. P. J. Reynolds, H. E. Stanley, and W. Klein, *J. Phys. A: Math. Gen.* **11**:L199 (1978).
9. P. J. Reynolds, H. E. Stanley, and W. Klein, *Phys. Rev.* **21**:1223 (1980).
10. H. Gould and J. Tobochnik, *An Introduction to Computer Simulation Methods* (Addison-Wesley, New York, 1988), Vol. 2, p. 440.
11. C.-K. Hu, *Phys. Rev. B* **46**:6592 (1992).
12. C.-K. Hu, *Phys. Rev. Lett.* **69**:2739 (1992).
13. C.-K. Hu, *J. Phys. A: Math. Gen.* **27**:L813 (1994).
14. C.-K. Hu, *Chin. J. Phys. (Taipei)* **32**:519 (1994).
15. C.-K. Hu, *Phys. Rev. B* **51**:3922 (1995).
16. J. L. Cardy, *J. Phys. A: Math. Gen.* **25**:L201 (1992).
17. M. Sahimi and H. Rassamdana, *J. Stat. Phys.* **78**:1157-1164 (1995).
18. A. Aharony and J. P. Hovi, *Phys. Rev. Lett.* **72**:1941 (1994).
19. R. M. Ziff, *Phys. Rev. Lett.* **72**:1942 (1994).

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