Binary data corruption due to a Brownian agent. II. Two dimensions, competing agents, and generalized couplings

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This work is a continuation of our previous investigation of binary data corruption due to a Brownian agent [Phys. Rev. E **59**, 5172 (1999)]. We extend our study in three main directions which allow us to make closer contact with real bistable systems. These are (i) a detailed analysis of two dimensions, (ii) the case of competing agents, and (iii) the cases of asymmetric and quenched random couplings. Most of our results are obtained by extending our original phenomenological model, and are supported by extensive numerical simulations. [S1063-651X(99)07208-6]

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I. INTRODUCTION

This paper is the second part of a two-stage investigation into the statistics of an active random walker (Brownian agent) in a bistable medium. This is but one example of the myriad of systems in which a random walker interacts in some way with its environment [1,2]. We consider systems in which the Brownian agent (BA) performs a pure, unbiased random walk in a medium composed of elements which may take one of two values. On visiting a given element, the BA has a certain probability of switching the value of that element. In our first paper [3] [hereafter referred to as data corruption I (DCI)] we motivated our investigation into such processes using the example of data corruption caused by a Brownian agent (i.e., the elements were taken to be bits of binary data), and we outlined the possible applications of our results to describing soft error production in small-scale memory devices [4,5]. The disturbance of a bistable medium by a BA is also related to reversible chemical kinetics by a high mobility catalyst, and disordering of biatomic structures by a wandering agent [6,7], such as an anion or cation vacancy in NaCl, or an impurity in a semiconductor compound (e.g., Zn in GaAs). One may also view this process as a nonconserved cousin of magnetic disordering via spin exchange with a wandering vacancy [8-10]. Last, but not least, the analysis of this process yields a deeper understanding of the statistics of random walks.

In DCI we studied the simplest possible process, namely, a single BA disordering a bistable system, with a switching probability which is independent of the value of the element. One of our main concerns in DCI was to construct a phenomenological continuum model of the process. This enabled us to find results independent of microscopic details, and also to study coarse-grained quantities, such as the probability distribution for the local ''density of disorder'' which is less easily defined on a lattice. Our results were derived mostly for the case of one spatial dimension.

In order to make contact with a wider range of processes

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it is necessary to consider a more general model. This is the aim of the present work. We shall extend our original investigation in three directions. First, we shall present a careful analysis of two spatial dimensions. We find that the phenomenological model predicts the correct asymptotic behavior, despite the need to introduce some form of regularization. We present results for the mean density of disorder $\overline{\rho}$, and using some special properties of the continuum description, we shall also derive an approximate form for the probability distribution of the density of disorder. Second, we shall consider a system containing more than one BA, thereby inducing "competition" as each BA interferes with the disorder created by the others. The main result here is that the disordering efficacy (i.e., the global amount of disorder due to Nagents as compared to one agent) is massively reduced for dimensions less than two, whereas in precisely two dimensions, each BA eventually becomes independent (in that the disorder it creates is not reordered by other BA's). We shall present calculations based on the continuum model to make these statements quantitative. Third, we shall consider two kinds of generalized couplings between the BA and the bistable medium: asymmetric switching probabilities, and quenched random switching probabilities. We shall argue that these generalized couplings may be modeled within the continuum limit by simple generalizations of our original model, and we shall derive some basic consequences, which for the case of quenched randomness are particularly interesting (see the outline below). In all cases, we shall support our results by numerical simulations of the underlying lattice model.

A more detailed outline follows: In Sec. II we present a recapitulation of the results of DCI. We shall briefly describe the lattice model in general spatial dimension d, and present it in the form of a master equation. We then describe the associated continuum theory, and we state without derivation some pertinent results previously derived from this continuum model in DCI. In Sec. III we concentrate on calculating $\bar{\rho}$ in two spatial dimensions. Given that two is the critical dimension of the process, the results are modulated by logarithmic corrections. Therefore, great care is needed to compare different theoretical predictions and numerical results. We shall present (briefly) four alternative methods of

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calculation, and show that they all predict the same asymptotic behavior, and agree with the numerical simulations once the strong corrections to scaling are included. Using these results, we are also able to reconstruct the probability distribution for the density of disorder approximately. In Sec. IV we study the case of more than one BA in the context of a generalized version of our continuum model. We shall derive an exact integral expression (for asymptotically large times) for the disordering efficacy $\sigma_N(d)$ for arbitrary N. First we concentrate on d=1. We evaluate $\sigma_N(1)$ for N = 2,3, and 4, and also extract its functional form for $N \ge 1$. We then extend our study to arbitrary spatial dimension d < 2, and evaluate $\sigma_2(d)$ and $\sigma_N(d)$ for large N. We use these results to analytically continue to two dimensions, thereby avoiding the use of a microscopic regularization. In Sec. V we consider the system in d=1 with generalized couplings. First we study asymmetric rates, so, for example, in the data corruption process the BA will have different probabilities to switch $0 \rightarrow 1$ and $1 \rightarrow 0$. We propose to model this using a simple extension of the original continuum theory, based on the idea that relative to the nonzero "background disorder" the dynamics of the system are the same as the symmetric case. Second, we consider quenched random couplings. In this case we argue that $\overline{\rho}$ picks up logarithmic corrections in time, while the global amount of disorder remains unaffected. We shall also discuss the quenched average of the distribution function of disorder density, and show that it is very sensitive to the distribution of the couplings. We shall support our results by numerical simulations which are described in detail in Sec. VI. We end the paper with a summary of our results and our conclusions.

II. RECAPITULATION

In this section we give a very brief review of the main ideas and some of the results contained in DCI in order to place the present work in a proper context. The process of a BA in a bistable medium is first modeled on a hypercubic lattice of dimension d. The position of the BA is denoted by a lattice vector **R**. In a time step δt the BA has a probability p to move to one of its 2d nearest neighbor sites. In making such a jump, there is a probability q that the element on the site departed from is switched. The elements are described by spin variables $\sigma_{\mathbf{r}}$ (where **r** denotes a discrete lattice vector) which may take the values ± 1 . The spin variables encode the information about the disordering process. For example in the data corruption process we label uncorrupted bits (of value 1) by spin +1 and corrupted bits (of value 0) by -1. Thus, we shall often use the terms "magnetization" density" and "global magnetization" which may be simply translated as "density of disorder" and "total amount of disorder."] This process is illustrated in Fig. 1 for d=2 and p = q = 1. We can define the dynamics via the probability distribution $P(\mathbf{R}, \{\sigma_{\mathbf{r}}\}, t)$, which is the probability that at time t, the BA is at position \mathbf{R} and the spins have values given by $\{\sigma_r\}$. This distribution evolves according to a master equation [11] which takes the form

	1	1	1	1	1
	1	1	1	1	1
	1	1	1	1	1
,	1	1	1	1	1
1	1	1	1	1	1
1	1	1	1	1	1

FIG. 1. Illustration of the data corruption process for d=2, with p=q=1. The initial uncorrupted state is shown on the left, with the BA represented by the filled circle. On the right we show a typical walk of ~ 20 steps. The BA switches a bit with each visit, so those bits visited an even number of times are restored to their original value.

$$P(\mathbf{R}, \{\sigma_{\mathbf{r}}\}, t+\delta t)$$

$$= (1-p)P(\mathbf{R}, \{\sigma_{\mathbf{r}}\}, t) + \frac{p(1-q)}{2d} \sum_{\mathbf{l}} P(\mathbf{R}+\mathbf{l}, \{\sigma_{\mathbf{r}}\}, t)$$

$$+ \frac{pq}{2d} \sum_{\mathbf{l}} P(\mathbf{R}+\mathbf{l}, \dots, -\sigma_{\mathbf{R}+\mathbf{l}}, \dots, t), \qquad (1)$$

where $\{l\}$ represents the 2*d* orthogonal lattice vectors (which have magnitude *l*).

The most direct quantities to extract from the master equation are marginal averages, the simplest of which is the magnetization density given by

$$\Theta(\mathbf{r}_1, \mathbf{R}, t) \equiv \operatorname{Tr}_{\sigma} \sigma_{\mathbf{r}_1} P(\mathbf{R}, \{\sigma_{\mathbf{r}}\}, t).$$
(2)

Averaging the master equation over the spin variables gives

$$\Theta(\mathbf{r}, \mathbf{R}, t + \delta t) - \Theta(\mathbf{r}, \mathbf{R}, t)$$

$$= \frac{p}{2d} \sum_{\mathbf{l}} \left[\Theta(\mathbf{r}, \mathbf{R} + \mathbf{l}, t) - \Theta(\mathbf{r}, \mathbf{R}, t) \right]$$

$$- \frac{pq}{d} \Theta(\mathbf{r}, \mathbf{r}, t) \sum_{\mathbf{l}} \delta_{\mathbf{r}, \mathbf{R} + \mathbf{l}}.$$
(3)

At this stage a continuum limit (in both space and time) may be taken of the above equation, which yields

$$\partial_t \Theta(\mathbf{r}, \mathbf{R}, t) = \frac{D}{2} \nabla_{\mathbf{R}}^2 \Theta(\mathbf{r}, \mathbf{R}, t) - \lambda \Theta(\mathbf{r}, \mathbf{R}, t) \Delta_{\mathbf{l}}(\mathbf{r} - \mathbf{R}).$$
(4)

Two parameters have appeared: the effective diffusion constant of the BA given by $D = 2l^2p/\delta t$, and an effective coupling between the BA and the spins given by $\lambda \propto pql^d/\delta t$. This continuum equation for Θ has the form of a diffusion equation with a sink potential $\Delta_{\mathbf{l}}(\mathbf{r})$ which is a strongly localized function with lateral extent *l* and normalized to unity. In the naive continuum limit, this function may be taken to be a *d*-dimensional Dirac delta function. However, for *d* ≥ 2 , it is necessary to smear this function in order to regularize the theory.

In DCI an alternative continuum description was obtained by viewing the process as a stochastic cellular automaton. The process is then defined in terms of the position $\mathbf{R}(t)$ of the BA (which is now an independent stochastic process), and the coarse-grained density of disorder (or magnetization density) which is defined in a small region of space at a specific time, and is a functional of $\mathbf{R}(t)$. In some sense, one may view this in the same spirit as a Langevin description of a stochastic process described at a more fundamental level by a master equation. Taking the continuum limit of this description yields first a simple Langevin equation for the position of the BA,

$$\frac{d\mathbf{R}}{dt} = \boldsymbol{\xi}(t), \tag{5}$$

where $\xi(t)$ is a noise term, each component of which is an uncorrelated Gaussian random variable with zero mean, [i.e., $\xi_i(t)$ is a white noise process]. The correlator of ξ is given by

$$\left\langle \xi_i(t)\xi_i(t')\right\rangle = D'\,\delta_{i,i}\delta(t-t').\tag{6}$$

Here and henceforth, angled brackets indicate an average over the noise (or equivalently the paths of the BA). The BA is chosen to reside initially at the origin: $\mathbf{R}(0) = \mathbf{0}$.

The evolution of the magnetization density ϕ is described by

$$\partial_t \phi(\mathbf{r},t) = -\lambda' \phi(\mathbf{r},t) \Delta_l (\mathbf{r} - \mathbf{R}(t)).$$
(7)

This equation may be integrated to give the explicit functional solution

$$\phi(\mathbf{r},t) = \exp\left[-\lambda' \int_0^t dt' \Delta_l(\mathbf{r} - \mathbf{R}(t'))\right].$$
(8)

The above solution is obtained for an initial condition $\phi(\mathbf{r},0) = 1$, which we shall use exclusively. In terms of the original lattice model it corresponds to choosing all the spins to have the initial value of +1, so that we measure the subsequent disorder of the system by counting the number of minus spins in the system. [Although of no relevance to the coarse-grained description, we mention here that it is often convenient to choose the spin at the origin (i.e., the initial BA position) to be -1, so that for p=q=1 all spins have value +1 after the first jump of the BA.]

In DCI we showed that the continuum descriptions (4) and (7) are indeed equivalent with the identifications D = D' and $\lambda = \lambda'$. [This was proven by considering the former as an imaginary time Schrödinger equation [12], and writing the solution of the latter as an imaginary time Feynman path integral [13]]. This ends our review of the modelization of the process—the reader is referred to DCI for further details and discussions.

In DCI we exclusively used the continuum description (8) to generate results for various average quantities. The simplest quantity to consider is the mean magnetization density given by $m(\mathbf{r},t) = \langle \phi(\mathbf{r},t) \rangle$ [which is equivalent to the sum of $\Theta(\mathbf{r},\mathbf{R},t)$ over the BA position **R**]. For d < 2 there is no necessity for regularization, and we replaced the sink function Δ_l by the *d*-dimensional Dirac δ function. The (temporal) Laplace transform for $m(\mathbf{r},t)$ was found to have the exact form

$$\hat{m}(\mathbf{r},s) = \frac{1}{s} \left[1 - \frac{\lambda \hat{g}(\mathbf{r},s)}{1 + \lambda \hat{g}(\mathbf{0},s)} \right], \tag{9}$$

where $\hat{g}(\mathbf{r},s)$ is the Laplace transform of the diffusion equation Green function

$$g(\mathbf{r},t) = (2\pi Dt)^{-d/2} \exp(-r^2/2Dt).$$
(10)

This exact result allows one to extract a great deal of statistical information about the process. First one may simply invert the Laplace transform to find the average magnetization density (or average density of disorder relative to $\frac{1}{2}$) as a function of **r** and *t*. Explicit forms are given in DCI for d=1. We note here that the average magnetization density at the origin decays for long times in d=1 as

$$m(0,t) = \left(\frac{2D}{\pi\lambda^2 t}\right)^{1/2} \left[1 + O\left(\frac{D}{\lambda^2 t}\right)\right].$$
 (11)

The continuum solution has the important property that $\langle \phi(\mathbf{r},t;\lambda)^n \rangle = \langle \phi(\mathbf{r},t;n\lambda) \rangle$. This allows us to utilize the exact solution (9) to reconstruct the probability density for the magnetization density. We define \mathcal{P} via

$$\mathcal{P}(\boldsymbol{\phi}, \mathbf{r}, t) = \langle \, \delta(\boldsymbol{\phi} - \boldsymbol{\phi}_{\mathbf{R}}(\mathbf{r}, t)) \rangle, \tag{12}$$

where $\phi_{\mathbf{R}}(\mathbf{r},t)$ is the stochastic field solution given in Eq. (8). As explained in DCI, one may solve for this distribution exactly. In particular, for d=1 the probability distribution for the magnetization density at the origin takes the form

$$\mathcal{P}(\phi,0,t) = \frac{1}{(\pi t)^{1/2}} \frac{1}{\hat{\lambda}\phi} \exp\left[-\frac{(\ln(\phi))^2}{4\hat{\lambda}^2 t}\right], \quad (13)$$

which is a log-normal distribution [and where we have defined $\hat{\lambda} = \lambda/(2D)^{1/2}$]. This is interesting, as it indicates the extreme nature of the fluctuations in this system. For instance, the typical value of the magnetization density can be found from the above expression to decay exponentially, i.e., $\phi_{typ} \sim \exp(-\hat{\lambda}^2 t/2)$, whereas the mean density decays as $1/\hat{\lambda}\sqrt{t}$ as given in Eq. (11).

Another interesting quantity which may be extracted from $m(\mathbf{0},t)$ is the average global magnetization defined (relative to its initial value) as

$$M(t) = \int d^d r [\langle \phi(\mathbf{r}, 0) \rangle - \langle \phi(\mathbf{r}, t) \rangle].$$
(14)

As shown in DCI, for d < 2 this quantity obeys the exact relation

$$\frac{dM(t)}{dt} = \lambda m(\mathbf{0}, t).$$
(15)

Thus, for large times in one dimension we have $M(t) \sim (Dt)^{1/2}$ independent of the coupling λ . In other words, the total amount of disorder created by a single BA on average increases as $(Dt)^{1/2}$, and is (rather surprisingly) independent of the coupling between the BA and the spins (for times larger than D/λ^2).

III. TWO-DIMENSIONAL SYSTEMS

In this section we shall present a careful analysis of the case of two dimensions. The simple random walk is recurrent for dimensions $d \leq 2$, whereas for d > 2 the walker has a probability less than unity for ever returning to its starting point [1]. This basic fact from the theory of random walks has an immediate implication for our data corruption problem. The nonrecurrent nature of random walks for d>2 implies that the BA will continually corrupt new regions of the system, and rarely revisit sites which it has previously corrupted. Thus the relative density of disorder (or average magnetization density) at the origin m(0,t) will decay to a nonzero (and nonuniversal) value, and the total amount of disorder (or average global magnetization) M(t) will increase linearly in time, with a nonuniversal prefactor. For practical applications, in which one wishes to limit the disordering capabilities of the BA, the first requirement is to restrict the geometry of the system to a dimension $d \leq 2$. So d=2 is the critical dimension of the problem, and because of this we can expect logarithmic corrections to modulate the leading order results, and also to cause long crossover times, thus making numerical results more difficult to interpret.

In DCI we studied some general properties of higher dimensional systems, and we also derived an approximate form for $m(\mathbf{0},t)$ in d=2 using a crude form of regularization. Rough agreement was found between this result and simulations, but no quantitative data analysis was performed. In this section we shall rederive the form of m(0,t) more carefully. The reasons for this are threefold: first, so that we can have confidence in the leading order result, and also obtain some idea of the subleading corrections; second, to provide insight into the relation between the discrete and continuum approaches in two dimensions (which is important given that the latter must be regularized for d=2; and third, to allow us to construct the form of the probability distribution of the magnetization density [for which we need as much information about $m(\mathbf{0},t)$ as possible]. We shall use four different methods (which will each be described with brevity) to derive the form of m(0,t). Each has its strong and weak points, as we shall see.

A. Lattice calculation from Eq. (3)

Referring to the equation of motion (3) for the marginal average, we set d=2, and for convenience we set the hopping probability p=1, giving

$$\Theta(\mathbf{r},\mathbf{R},t+\delta t) = \frac{1}{4} \sum_{\mathbf{l}} \Theta(\mathbf{r},\mathbf{R}+\mathbf{l},t) - \frac{q}{2} \Theta(\mathbf{r},\mathbf{r},t) \sum_{\mathbf{l}} \delta_{\mathbf{r},\mathbf{R}+\mathbf{l}}.$$
(16)

As an initial condition we take the BA to be located at the origin, and all spins to be +1, except the spin at the origin which is taken to be -1. [This convention is useful for q = 1, so that all spins have value +1 after one time step. However, if $q \ll 1$, it is more convenient to take the spin at the origin to be initially +1, as the chance of it flipping after one time step is small. We stress that these different choices for the initial value of the spin at the origin have no effect on the asymptotic properties of the system, and only serve to smooth the magnetization density in the immediate vicinity

of the origin.] Thus $\Theta(\mathbf{r}, \mathbf{R}, 0) = \delta_{\mathbf{R}, \mathbf{0}}(1 - 2\delta_{\mathbf{r}, \mathbf{0}})$. The solution of Eq. (16) may be attained by discrete Fourier and Laplace transform. Defining the former via

$$\widetilde{\Theta}(\mathbf{r},\mathbf{k},t) = \sum_{\mathbf{R}} \Theta(\mathbf{r},\mathbf{R},t) e^{i\mathbf{k}\cdot\mathbf{R}}$$
(17)

and the latter via

$$\hat{\Theta}(\mathbf{r},\mathbf{R},z) = \sum_{n=0}^{\infty} \Theta(\mathbf{r},\mathbf{R},n\,\delta t)z^n,$$
(18)

it is fairly straightforward to diagonalize Eq. (16) to the form

$$\hat{\Theta}(\mathbf{r},\mathbf{k},z) = \frac{(1-2\,\delta_{\mathbf{r},\mathbf{0}}) - 2qzf(\mathbf{k})e^{i\mathbf{k}\cdot\mathbf{r}}\hat{\Theta}(\mathbf{r},\mathbf{r},z)}{1-zf(\mathbf{k})},\quad(19)$$

where $f(\mathbf{k}) = [\cos k_1 l + \cos k_2 l]/2$. One may now solve the above equation self-consistently for $\hat{\Theta}(\mathbf{r},\mathbf{r},z)$ by inverting the discrete Fourier transform. One has

$$\hat{\Theta}(\mathbf{r},\mathbf{r},z) = \frac{(1-2\,\delta_{\mathbf{r},\mathbf{0}})\int d\mathbf{k}\,e^{-i\mathbf{k}\cdot\mathbf{r}}D(\mathbf{k},z)}{1+2\,qz\int d\mathbf{k}f(\mathbf{k})D(\mathbf{k},z)},\qquad(20)$$

where $D(\mathbf{k}) = [1 - zf(\mathbf{k})]^{-1}$ and the momentum integrals are over the two-dimensional Brillouin zone. The average magnetization density at the origin is given by summing $\Theta(\mathbf{0}, \mathbf{R}, t)$ over \mathbf{R} , which is equivalent to the zero Fourier mode $\widetilde{\Theta}(\mathbf{0}, \mathbf{0}, t)$. Thus, substituting Eq. (20) into Eq. (19), after some rearrangement we have

$$\sum_{\mathbf{R}} \hat{\Theta}(\mathbf{0}, \mathbf{R}, z) = \frac{-1}{(1-z)} \left[\frac{1 + q'(1-z) \int d\mathbf{k} D(\mathbf{k}, z)}{1 + q' \int d\mathbf{k} D(\mathbf{k}, z)} \right],$$
(21)

where $q' \equiv 2q/(1-2q)$. [The function $\int d\mathbf{k} D(\mathbf{k},z)$ is very well known in the theory of random walks [1], and is the discrete Laplace transform of the probability of a random walker to return to its starting point after *n* steps.] Finally we must inverse Laplace transform the above equation. For large *n* we can extract the asymptotic form of the average magnetization by invoking the Tauberian theorem [1]. In this case we need the form of the Laplace transform as $z \rightarrow 1$. Using [1]

$$\int d\mathbf{k}D(\mathbf{k},z) \sim \frac{1}{\pi} \ln\left(\frac{8}{1-z}\right) [1+O(1-z)], \qquad (22)$$

from the Tauberian theorem we have the asymptotic result

$$\sum_{\mathbf{R}} \Theta(\mathbf{0}, \mathbf{R}, n\,\delta t) = \frac{-1}{(q'/\pi) \ln n + C(q)}.$$
 (23)

The constant C(q) is not accessible from the Tauberian theorem, although it could in principle be calculated from a careful inverse Laplace transform of Eq. (21). In our simulations we generally take the switching probability q to be unity, which implies q' = -2 and thus

$$\sum_{\mathbf{R}} \Theta(\mathbf{0}, \mathbf{R}, n\,\delta t) = \frac{1}{(2/\pi) \ln n - C(1)}.$$
 (24)

In Sec. VI we shall make a direct comparison of this result with numerical simulations in two dimensions. It is also interesting to note from Eq. (21) that setting the switching probability q = 1/2 gives $\sum_{\mathbf{R}} \Theta(\mathbf{0}, \mathbf{R}, n \, \delta t) = -\delta_{n,0}$. In other words, the average value of the spin at the origin remains at zero after one jump of the agent. This "maximal uncertainty" for $q = \frac{1}{2}$ only holds at the origin, and spins at other lattice sites will have a positive mean for all times. As a final note, if $q \ll 1$ it is more convenient to choose the initial value of the spin at the origin to be +1 in which case we find, from the foregoing analysis,

$$\sum_{\mathbf{R}} \Theta(\mathbf{0}, \mathbf{R}, n\,\delta t) = \frac{1}{(2q/\pi)\ln n + C(q)}.$$
 (25)

B. Diffusion equation (4) with smeared sink function

We now wish to solve for the coarse-grained magnetization density $m(\mathbf{0},t)$ within some continuum limit. In this subsection we accomplish this by solving the diffusion equation (4) with a smeared sink function. This calculation is the closest in spirit to the lattice calculation since (4) is the direct continuum analog of the discrete equation (3) solved in Sec. III A. Setting $\mathbf{r}=\mathbf{0}$, Eq. (4) takes the form

$$\partial_t \Theta(\mathbf{0}, \mathbf{R}, t) = \frac{D}{2} \nabla_{\mathbf{R}}^2 \Theta(\mathbf{0}, \mathbf{R}, t) - \lambda \Theta(\mathbf{0}, \mathbf{R}, t) \Delta_l(\mathbf{R}).$$
(26)

The simplest finite range form to take for the sink function Δ_l is a radial function which is zero outside a radius *l*, and (through normalization) equal to $(1/\pi l^2)$ within this radius. It is also convenient to smear the initial condition of Θ in the same way.

The above equation may be solved by standard boundary value techniques. For the sake of brevity, we shall present no calculational details. The Laplace transform of the magnetization density at the origin is found to be (in the limit of $l \rightarrow 0$)

$$\hat{m}(\mathbf{0},s) = \frac{1}{s} \left[\frac{1}{w(\tilde{\lambda}) - \tilde{\lambda} \ln(sl'^2)} \right],$$
(27)

where $\tilde{\lambda} = \lambda/2 \pi D$, $l' = l(2/D)^{1/2}$, and $w(x) = \sqrt{x}I_0(2\sqrt{x})/I_1(2\sqrt{x})$, which approaches unity for small x ($I_n(z)$ is the modified Bessel function [14]). It is important to mention that this expression is not valid for Re(s)>1/l', and thus the apparent pole is an artifact of the limit $l \rightarrow 0$. We refer the reader to Appendix A, in which this Laplace transform is inverted. For large t the result takes the form

$$m(\mathbf{0},t) = \frac{1}{w(\tilde{\lambda}) + \tilde{\lambda} \ln(t/\tau)},$$
(28)

where $\tau = l'^2 e^{-\gamma}$, and $\gamma = 0.57721...$ is Euler's constant.

Thus we have arrived at our result, and indeed the leading order behavior has the same functional form as Eq. (25) derived in Sec. III A. A closer comparison of the leading terms indicates that $\lambda \propto q$ for $q \ll 1$ as expected on physical grounds. In the present continuum calculation, we have also extracted an explicit form for the subleading term [i.e., $w(\tilde{\lambda})$], which is nonuniversal and depends on the precise form of the smeared sink function.

C. Infinite order perturbation theory of Eq. (8)

In DCI all results were obtained from an infinite order perturbation expansion of Eq. (8). In two dimensions each term in the expansion is divergent. Using the time cutoff regularization scheme allows one to extract the dominant contribution from each term, and to resume the series. The result is

$$m(\mathbf{0},t) = \frac{1}{1 + \tilde{\lambda} \ln(t/t_0)}.$$
(29)

It is interesting to compare this with Eq. (28) obtained in Sec. III B. They are seen to agree [for small λ , in which case $w(\tilde{\lambda}) \approx 1$] if we make the identification $t_0 = \tau (=2l^2e^{-\gamma}/D)$. This indeed supports the role of t_0 as a microscopic correlation time of the noise, since it is seen to correspond to the time taken for the diffusion process to correlate a microscopic region of size $\sim l^2$.

D. Analytic continuation from Eq. (9)

As a fourth method of extracting the form of m(0,t) for d=2 we briefly mention analytic continuation from d<2. As discussed in Sec. II, no regularization is required in the perturbation expansion method for d<2, and one retrieves the exact result

$$\hat{m}(\mathbf{0},s) = \frac{1}{s} \left[\frac{1}{1 + \lambda \hat{g}(\mathbf{0},s)} \right], \tag{30}$$

where

$$\hat{g}(\mathbf{0},s) = \frac{\Gamma(1-d/2)}{(2\pi D)^{d/2} s^{1-d/2}},$$
(31)

and $\Gamma(z)$ is the gamma function [14]. Using the result $\Gamma(z) \sim 1/z$ for $z \rightarrow 0$, for $\epsilon \equiv 2 - d \rightarrow 0$ we have

$$\hat{m}(\mathbf{0},s) = \frac{1}{s} \left[\frac{1}{1 - \tilde{\lambda} \ln(s/s_0)} \right], \tag{32}$$

where $s_0 = e^{-2/\epsilon}$.

On comparing this result with Eq. (27) obtained in Sec. III B, we see that they are equivalent [for small λ , in which case $w(\tilde{\lambda}) \approx 1$] if we make the identification $1/\epsilon = \ln(1/l')$. This relation has only a formal meaning, as there is no physical sense in continuing dimensionality. However, it is instructive to learn that the leading order result may be re-

trieved intact from analytic continuation. Indeed, we shall use this tool in Sec. IV to study competing agents in two dimensions.

E. Calculation of the probability distribution

To complete this section, we shall briefly discuss the probability distribution of the magnetization density at the origin. As mentioned in Sec. II, the continuum theory (8) has the property $\langle \phi(\mathbf{0},t;\lambda)^n \rangle = \langle \phi(\mathbf{0},t;n\lambda) \rangle$. Thus knowledge of the λ dependence of the first moment allows one to reconstruct all the moments of the magnetization density, and thus the probability density $\mathcal{P}(\phi, 0, t)$ for this quantity. The method for retrieving \mathcal{P} from the moments is given in detail in DCI for the case of d=1. A similar calculation suffices for d=2, so long as one has an "accurate" form for the mean magnetization density. In the present section we have attacked this problem from four different directions, and have arrived at agreement for the asymptotically dominant term for m(0,t). However, as noted in DCI, it is necessary to have more information than this in order to construct \mathcal{P} correctly. If one tries to calculate \mathcal{P} using $m(\mathbf{0},t;\lambda) \sim 1/\tilde{\lambda} \ln t$ one obtains a distribution $\mathcal{P} \sim m(\mathbf{0}, t) / \phi$ which is singular at $\phi = 0$. The subleading correction to $m(0,t;\lambda)$ is crucial to determine the distribution correctly. Our results from Secs. III A-III D are in agreement with regard to the time dependence, but differ in the way in which λ appears in the subleading term. We prefer Eq. (28), in that an explicit form $w(\tilde{\lambda})$ appears. However, we have been unable so far to reconstruct \mathcal{P} using this form due to the complicated nature of w. For small λ , $w(\tilde{\lambda}) \approx 1$, and Eq. (28) then coincides with the less controlled results (29) and (32) of Secs. III C and III D, respectively. The reconstruction of \mathcal{P} is possible from these forms, and one finds

$$\mathcal{P}(\boldsymbol{\phi}, \mathbf{0}, t) = \boldsymbol{\beta}(t) \boldsymbol{\phi}^{\boldsymbol{\beta}(t)-1}, \tag{33}$$

with $\beta(t) = 1/\tilde{\lambda} \ln(t/t_0)$. As $t \to \infty$ this distribution approaches the form $m(0,t)/\phi$, but is never singular for finite times. In Sec. VI we shall describe our attempts to measure \mathcal{P} for two-dimensional systems. Our numerical results are in surprisingly good agreement with Eq. (33) above.

IV. COMPETING AGENTS

In this section we shall analyze the effects of many BA's within the system. We shall assume the BA's to be noninteracting, in the sense that they are unaware of each other's immediate presence. The nontrivial statistics reside in the fact that the disordering effects of the BA's statistically interact via the overlap of the BA histories. As we have already seen, a single BA interferes with the previous disorder it has created, such that the amount of disorder does not simply increase linearly in time. This effect is more severe when more than one BA is present, as each BA can disturb the disorder that another BA has previously created.

We measure the strength of this interference via a quantity called the "disordering efficacy" of the agents, defined as

$$\sigma_N(d) \equiv \lim_{t \to \infty} \frac{M^{(N)}(t)}{M^{(1)}(t)},\tag{34}$$

where $M^N(t)$ is the average global disorder created by N agents. If the BA's were truly independent (in terms of the disorder they create), then we would expect $\sigma_N = N$. As we shall see, for d < 2 the value of σ_N is strongly reduced below this value. However, for d=2 this value is recovered, but only in the deep asymptotic regime $(t \ge e^N)$.

The extension to many BA's is easily modeled within the continuum theory. We introduce *N* random walkers, each of which is described by a position vector $\mathbf{R}_{\alpha}(t)$, $\alpha = 1, 2, ..., N$. Since the BA's are independent, we have

$$\frac{d\mathbf{R}_{\alpha}}{dt} = \boldsymbol{\xi}_{\alpha}(t), \qquad (35)$$

where $\boldsymbol{\xi}_{\alpha}(t)$ are independent Gaussian white noise sources with zero mean. The equation of motion for the coarsegrained magnetization density $\boldsymbol{\phi}^{(N)}$ takes the form

$$\partial_t \phi^{(N)}(\mathbf{r},t) = -\lambda \phi^{(N)}(\mathbf{r},t) \sum_{\alpha=1}^N \Delta_l(\mathbf{r} - \mathbf{R}_{\alpha}(t)), \quad (36)$$

with solution

$$\phi^{(N)}(\mathbf{r},t) = \prod_{\alpha=1}^{N} \exp\left[-\lambda \int_{0}^{t} dt' \Delta_{l}(\mathbf{r} - \mathbf{R}_{\alpha}(t'))\right]. \quad (37)$$

On averaging over the paths of the N agents, we have the particularly simple result

$$m^{(N)}(\mathbf{r},t) \equiv \langle \phi^{(N)}(\mathbf{r},t) \rangle = m^{(1)}(\mathbf{r},t)^{N}.$$
 (38)

Thus the average global magnetization is given by

$$M^{(N)}(\mathbf{r},t) \equiv \int d^{d}r [m^{(N)}(\mathbf{r},0) - m^{(N)}(\mathbf{r},t)] = \int d^{d}r \\ \times [1 - m^{(1)}(\mathbf{r},t)^{N}].$$
(39)

Scaling the spatial coordinate by the diffusion length scale $(2Dt)^{d/2}$, from Eqs. (34) and (39) we have

$$\sigma_N(d) = \lim_{t \to \infty} \frac{\int d^d r [1 - m^{(1)} (\mathbf{r}/(2Dt)^{1/2}, t)^N]}{\int d^d r [1 - m^{(1)} (\mathbf{r}/(2Dt)^{1/2}, t)]}.$$
 (40)

To proceed with the calculation it is convenient to first perform the large-*t* limit. We concentrate on d < 2, and therefore replace $\Delta_l(\mathbf{r})$ by the Dirac δ -function. From Eqs. (9) and (10) it is a fairly straightforward matter to show that

$$m(r) \equiv \lim_{t \to \infty} m^{(1)} (\mathbf{r}/(2Dt)^{1/2}, t) = \frac{2}{\Gamma(1 - d/2)} \int_0^r du \, u^{1 - d} e^{-u^2}.$$
(41)

We may use this result to explicitly evaluate the denominator of Eq. (40), giving us

TABLE I. The predicted values of $\sigma_N(1)$ for N=2,3, and 4, from Eq. (45), compared with our numerical simulations.

Ν	$\sigma_N(1)_{ m theory}$	$\sigma_N(1)_{ m simul}$
2	1.414	1.42(1)
3	1.662	1.66(1)
4	1.835	1.84(2)

$$\sigma_N(d) = d\Gamma(1 - d/2) \int_0^\infty dr \, r^{d-1} (1 - m(r)^N).$$
(42)

We refer the reader to Appendix B, where it is demonstrated that the above expression may be recast in the more useful form

$$\sigma_N(d) = \frac{2N(N-1)}{\Gamma(1-d/2)} \int_0^\infty dr \, r^{1-d} e^{-2r^2} m(r)^{N-2} \qquad (43)$$

for N > 1. We note that the result for two agents follows immediately from this expression, and we have $\sigma_2 = 2^{d/2}$. This result is striking. For the case of d = 1, we see that two agents create only $\sqrt{2}$ as much disorder as one agent. Also, assuming we may continue this result to exactly two dimensions, we find that $\sigma_2(2)=2$, i.e., two agents create disorder independently. In the remainder of this section we shall use Eq. (43) for two purposes. First, we shall concentrate on d= 1 and evaluate the exact values of $\sigma_3(1)$ and $\sigma_4(1)$, which may be used to compare with numerical simulations. Second, we shall evaluate the integral for large N for arbitrary $d \in [0,2]$ using a saddle point method. This calculation will make clear the tremendous difference in the large-N behavior of $\sigma_N(d)$ for d < 2 and d = 2. We end the section with a simple scaling argument which helps us to understand these analytic results.

For d=1, expression (41) is simply $m(x) = \operatorname{erf}(x)$, where $\operatorname{erf}(z)$ is the error function [14]. Thus from Eq. (43) we have

$$\sigma_N(1) = \frac{2N(N-1)}{\pi^{1/2}} \int_0^\infty dx \ e^{-2x^2} \operatorname{erf}(x)^{N-2}.$$
 (44)

We are able to evaluate these integrals for N=3 and 4. The details may be found in Appendix C. The results are

$$\sigma_{2}(1) = \sqrt{2}, \quad \sigma_{3}(1) = \frac{6\sqrt{2}}{\pi} \sin^{-1}\left(\frac{1}{\sqrt{3}}\right),$$
$$\sigma_{4}(1) = \frac{12\sqrt{2}}{\pi} \sin^{-1}\left(\frac{1}{3}\right). \tag{45}$$

The numerical values of these expressions are presented in Table I along with the results from our computer simulations (the details of which may be found in Sec. VI). Excellent agreement is found.

We now return to the case of arbitrary $d \in [0,2]$ and consider the limit of large *N*. For the sake of generality, we study the integral

$$Q_{N}(\beta,d) = \int_{0}^{\infty} dr \, r^{1-d} e^{-\beta r^{2}} m(r)^{N}, \qquad (46)$$

with m(r) as given above in Eq. (41). We can recover the disordering efficacy via

$$\sigma_N(d) = \frac{2N(N-1)}{\Gamma(1-d/2)} Q_{N-2}(2,d).$$
(47)

We wish to implement a saddle point calculation, so we rewrite Eq. (46) as

$$Q_N(\beta, d) = \int_0^\infty dr \, r^{1-d} e^{-F_N(r)}, \tag{48}$$

where

$$F_N(r;\beta,d) = \beta r^2 - N \ln(m(r)). \tag{49}$$

The saddle point is defined via $dF_N/dr|_{r=r_0}=0$, which for r_0 yields the transcendental equation

$$\beta r_0^d = \frac{N e^{-r_0^2}}{m(r_0) \Gamma(1 - d/2)}.$$
(50)

For $N \ge 1$, it is easy to see that $r_0 \sim (\ln N)^{1/2} \ge 1$, so that $F(r_0) \ge 1$ and the saddle point method is self-consistently justified.

For the sake of brevity we give no details of the saddle point calculation. The result is

$$Q_N(\beta, d) = \frac{\Gamma(\beta)\Gamma(1 - d/2)^{\beta}}{2} \frac{(\ln N)^{(\beta - 1)d/2}}{N^{\beta}} \times \left[1 - \frac{(\beta - 1)d^2}{4} \frac{\ln(\ln N)}{\ln N} + O\left(\frac{1}{\ln N}\right)\right].$$
(51)

Combining Eqs. (47) and (51) then gives us the final large-*N* result

$$\sigma_N(d) = \Gamma(1 - d/2) (\ln N)^{d/2} \left[1 - \frac{d^2}{4} \frac{\ln(\ln N)}{\ln N} + O\left(\frac{1}{\ln N}\right) \right].$$
(52)

In particular, for d=1 the disordering efficacy increases as $(\ln N)^{1/2}$ with strong logarithmic corrections. We also note that for real-valued d<2, the disordering efficacy increases as $(\ln N)^{d/2}$. In other words, the BA's overlap very strongly in their disordering for all d<2.

The main reason for calculating $\sigma_N(d)$ for real-valued d < 2 was to attempt to analytically continue the result to d = 2, which is otherwise difficult to calculate due to the presence of an explicit regulator. However, if we simply set d = 2 in our expression for $\sigma_N(d)$ above, we find that the apparent behavior $\sigma_N(2) \sim \ln N$ is multiplied by the infinite constant. This gives us the hint that the large-*N* behavior for d=2 is stronger than $\ln N$, but offers no more information than that. We can, however, trace the failure of the saddle point calculation for d=2 back to the saddle point equation (50) for r_0 . For large *N* and d < 2 we have the leading result

 $r_0 \sim [\ln \epsilon N]^{1/2}$, where $\epsilon = 2 - d$ [and we have obtained this result from Eq. (50) by taking the small- ϵ and large-r limit of m(r)]. Thus for fixed $\epsilon > 0$ we may always take N large enough to create a saddle point at a large value of r_0 , in which case the result (52) is valid. However for fixed N (no matter how large), taking $\epsilon \rightarrow 0$ squeezes the saddle point back into the origin, in which case the saddle point method is of no use.

Therefore we cannot use Eq. (52) to continue to d=2 analytically. However, a simpler method may be used to extract the result. We refer the reader to the original expressions for $\sigma_N(d)$ and m(r) as given in Eqs. (43) and (41), respectively. Taking $\epsilon \ll 1$ and $N \gg 1$, these equations take the forms

$$\sigma_N(d) = \epsilon N^2 \int_0^\infty dr \, r^{\epsilon - 1} e^{-2r^2} m(r)^N \tag{53}$$

and

$$m(r) = \epsilon \int_0^r du \ u^{\epsilon - 1} e^{-u^2}.$$
 (54)

On studying Eq. (53) we see that as $\epsilon \rightarrow 0$ the integral is dominated by small *r*. Thus we require the small-*r* form for m(r) which is easily extracted from Eq. (54) to be $m(r) \approx r^{\epsilon}$. We now break the integral in Eq. (53) into two pieces. The first piece encompasses the range $r \in (0,1)$ so that we can neglect the Gaussian factor and substitute the small *r* form of m(r) to find the leading order result

$$\sigma_N^I(d) = \epsilon N^2 \int_0^1 dr \ r^{N\epsilon - 1} = N.$$
(55)

The second piece encompasses the range $r \in (1,\infty)$, and for large enough *N* will have the asymptotic form $\sigma_N^{II}(d) \sim (\ln N)/\epsilon$. To summarize, the contribution $\sigma_N^{I}(d)$ dominates for $N \leq 1/\epsilon$, and the contribution $\sigma_N^{II}(d)$ dominates for $N \geq 1/\epsilon$. Thus for d=2 we have the result $\sigma_N(2)=N$ for all *N*. This is consistent with the analytic continuation to d=2 of the exact result $\sigma_2(d)=2^{d/2}$ found earlier.

These results may be understood in the following way. Consider N agents in a system of d dimensions. Since each agent performs a random walk, the amount of available space in which we can expect to find the agents has a volume $\sim t^{d/2}$. Also, we know from our previous results that the amount of disorder created by a single agent increases as $t^{d/2}$ for d < 2, but as $t/\ln t$ for d = 2. Thus for d < 2 the available space for N agents and the amount of disorder created by a single agent both scale as $t^{d/2}$, which means that there is certain to be interference between the agents, which will become severe for large N as we have seen. However for d=2 the amount of available space for N agents scales logarithmically faster with t than does the amount of disorder created by a single agent. Thus for times large enough such that $N \ll \ln t$, we expect the disorder created by the agents to become statistically independent and thus $\sigma_N(2) = N$. Note that the "independence time" grows exponentially with the number of agents, making a numerical observation difficult for even moderate values of N. Similar arguments have been made in the context of "the number of distinct sites visited by N random walkers'' [15]. In Sec. VI we shall present results for N=2,3, and 4 in two dimensions. The disordering efficacy is seen to approach its expected value slowly.

V. GENERALIZED COUPLINGS

In all of our work so far we have taken the coupling between the BA and the spins to be symmetric and spatially homogeneous. At least one of these properties is likely to be absent in a practical application of our model. Asymmetry in the switching probability is likely in data corruption since the states of a bit are not physically encoded in a symmetric way. Also, in chemical kinetic applications, the reaction rates between two chemical species are unlikely to be symmetric. Spatial homogeneity of the couplings is also an idealized situation. Generally there will be quenched random fluctuations in the switching rates of different sites, and it is important to quantify the effect of this randomness on the results obtained so far. In the next two subsections we shall consider these important generalizations in turn.

A. Asymmetric rates

Consider the lattice model described in Sec. II with the additional property that the probability of flipping a spin depends on the value of the spin. If the BA leaves a site with spin +1, we flip that spin with probability q^+ , whereas if the BA leaves a site with spin -1 we flip that spin with probability q^- . At a microscopic level the model is now considerably more complicated as the transition rates for flipping depend explicitly on the values of the spins. We shall not construct the master equation for this case. Rather, we shall try to construct an analog of the continuum model (7) using a simple physical idea.

The main effect of the asymmetric couplings is to favor one type of spin over the other. We make the approximation that in a coarse-grained model the dynamics of relaxation to this "background magnetization" are the same as the relaxation to zero magnetization in the symmetric model. We therefore write the analog of Eq. (7) for asymmetric rates in the form

$$\partial_t \phi(\mathbf{r},t) = -\lambda [\phi(\mathbf{r},t) - m_{\rm eq}] \Delta_l (\mathbf{r} - \mathbf{R}(t)), \qquad (56)$$

where $m_{eq} = (q^- - q^+)/(q^- + q^+)$. From this simple model it is easy to show that

$$\phi_A(\mathbf{r},t) = m_{\rm eq} - (1 - m_{\rm eq})\phi_S(\mathbf{r},t), \qquad (57)$$

where the subscripts A and S indicate "asymmetric" and "symmetric" respectively. Integrating this equation over space, and using definition (14) of the average global magnetization, we find

$$M_A(t) = (1 - m_{\rm eq})M_S(t) = \frac{2}{(1 + q^{-}/q^{+})}M_S(t), \quad (58)$$

which is an intuitively acceptable result.

B. Quenched random rates

We now consider an alternative generalization in which the rates are symmetric, but spatially inhomogeneous. At the microscopic level, this is modeled by attaching to each lattice position \mathbf{r} a quenched random variable $q_{\mathbf{r}} \in (0,1)$ drawn from some distribution $S(\{q_{\mathbf{r}}\})$. The random variable $q_{\mathbf{r}}$ gives the probability of switching in the event that the BA visits the site \mathbf{r} . In the continuum theory, we model these quenched random couplings by a simple extension of Eq. (7). We generalize the coupling parameter λ to a spatially inhomogeneous random function $\lambda(\mathbf{r})$, with a distribution $S[\lambda]$. This does not affect the solvability of the model due to the locality of the continuum theory, so that we have the explicit solution

$$\phi(\mathbf{r},t;\lambda) = \exp\left[-\lambda(\mathbf{r})\int_{0}^{t} dt' \Delta_{l}(\mathbf{r}-\mathbf{R}(t'))\right], \quad (59)$$

where we have emphasized the dependence of ϕ on λ .

There is an enormous variety in the forms of the quenched disorder distribution S that one may study. A comprehensive analysis is beyond the scope of the present work. In our simulations (to be described in Sec. IV), we have chosen to study switching rates $q_r \in (0,1)$ which are independent random variables drawn from a uniform distribution. There is more than one way in which such a distribution can manifest itself at the continuum level. Depending on the extent to which one coarse grains the lattice model, the continuum theory will have a distribution of couplings which is either very similar to the lattice distribution (i.e., uncorrelated and uniform), or more Gaussian in nature (due to the central limit theorem). We expect different physics according to the behavior of the distribution $S[\lambda]$ as $\lambda \rightarrow 0$. Roughly speaking, if S has zero weight for small λ , we expect the quenched disorder to be irrelevant to the system for large times. Conversely, if S has nonzero weight for small λ , then the disorder will play a role for arbitrarily late times. In order to exemplify this latter case, we consider as an example the distribution

$$S[\lambda] = \prod_{\mathbf{r}} S_{\text{loc}}(\lambda(\mathbf{r})), \qquad (60)$$

where S_{loc} is a local (or "on-site") distribution function which we take to be uniform:

$$S_{\rm loc}(\lambda) = (1/\lambda_m) H(\lambda_m - \lambda) H(\lambda), \tag{61}$$

where H(x) is the Heaviside step function [16]. Henceforth we shall concentrate on d=1 for simplicity.

Due to the way in which the quenched random couplings enter the continuum theory, we can easily generalize our earlier exact results. Let us first concentrate on the magnetization density at the origin, averaged over both the BA trajectories, and the distribution of λ . The first average is performed in the usual way using Laplace transforms (see DCI) and we have the analog of Eq. (9)

$$\hat{m}(0,s;\lambda) = \frac{1}{s} \left[\frac{1}{1 + \lambda(0)\hat{g}(0,s)} \right].$$
(62)

On averaging this density over the distribution of random couplings (61), we have

$$\langle \hat{m}(0,s;\lambda) \rangle_{S} = \left(\frac{2D}{\lambda_{m}^{2}s}\right)^{1/2} \ln \left[1 + \left(\frac{\lambda_{m}^{2}}{2Ds}\right)^{1/2}\right].$$
 (63)

Inverse Laplace transforming yields the asymptotic result

$$\langle m(0,t;\lambda)\rangle_{S} = \left(\frac{D}{2\pi\lambda_{m}^{2}t}\right)^{1/2} \ln\left(\frac{\lambda_{m}^{2}t}{2D}\right).$$
 (64)

Therefore, the decay in the homogeneous case (of $1/\sqrt{t}$ behavior) for the average magnetization density is slowed by a logarithmic factor due to the presence of the quenched random couplings. It is interesting that this logarithmic slowing down does not affect the average of the global magnetization. This quantity may be explicitly calculated using the method of infinite order perturbation theory described in DCI. The result is that the leading order term for large times is independent of the coupling, and is unaffected by the average over S. Thus the average over the random couplings will in no way affect the leading order behavior of M(t) $\sim (Dt)^{1/2}$. The average of the global magnetization is a very robust quantity, and is unaffected by scalings of the homogeneous couplings (as seen in Sec. II), and by making the couplings, quenched random variables. We have performed numerical simulations for systems with quenched random couplings and report our findings toward the end of Sec. IV.

Finally, we briefly discuss the changes that can occur to the probability distribution \mathcal{P} of the local magnetization density at the origin, when quenched random couplings are introduced. From Sec. II we have the exact result (13) for this function in d=1: it is a log-normal distribution. The same form will hold in the present case, but now the parameter λ is to be averaged over $S[\lambda]$. We average over the uncorrelated uniform distribution given above in Eq. (61). Simple integration gives

$$\langle \mathcal{P} \rangle_{S} = \frac{1}{2(\pi t)^{1/2} \hat{\lambda}_{m} \phi} \mathbf{E}_{1} \left(\frac{(\ln \phi)^{2}}{4 \hat{\lambda}_{m}^{2} t} \right), \tag{65}$$

where $\hat{\lambda}_m = \lambda_m / (2D)^{1/2}$, and $E_1(z)$ is the exponential integral [14]. This distribution behaves very much like the pure log-normal distribution for $\phi \ll 1$. However, we have the interesting result that the distribution is singular at $\phi = 1$. Explicitly, for $\psi \equiv 1 - \phi \ll 1$ we have

$$\langle \mathcal{P} \rangle_{S} \sim \frac{\ln[\psi^{2}/4\hat{\lambda}_{m}^{2}t]}{2(\pi t)^{1/2}\hat{\lambda}_{m}}.$$
 (66)

This logarithmic singularity stems from the fact that if $S[\lambda]$ has a nonzero weight for arbitrarily small λ , there will be a nonvanishing subset of systems for which almost no corruption occurs. For large times, this subset will appear as a pronounced "peak" in the distribution $\langle \mathcal{P} \rangle_S$ for $\phi \rightarrow 1$.

VI. NUMERICAL SIMULATIONS

In this section we describe the details of our numerical simulations, and present figures of our data to support the various theoretical claims made in the preceding sections. The simulations are performed on a d-dimensional lattice



FIG. 2. Plot of 1/m(0,t) (the inverse of the average magnetization density at the origin), as a function of $\ln t$ for d=2. The lower solid line is the asymptotic prediction, while the upper solid line (which is partially obscured by the data) is the prediction with corrections to scaling included. The inset shows the difference between 1/m(0,t) and its asymptotic form, plotted against $1/\ln t$. The data may be fitted to a straight line, thereby yielding the corrections to scaling explicitly.

(with *d* either 1 or 2) and follow the lattice rules described in Sec. II. We set the hopping probability p = 1, so that in each time step the BA is moved randomly to one of its 2*d* nearest neighbor sites. In doing so, the spin it leaves behind is flipped with probability *q* (which is set to unity unless otherwise stated). Systems are taken large enough so that the BA never touches the boundary during the lifetime of the simulation. We generally average over between 10^5 and 10^7 realizations, depending on the desired precision of the simulation.

In Sec. III we calculated the asymptotic form of the average magnetization density at the origin from an exact lattice calculation. As noted, the subleading term is not obtainable from the Tauberian theorem. However, it is clear from the various calculations in Sec. III that the corrections decay as $1/(\ln t)^2$. We have checked this in the following way. From our simulations we measure $m(\mathbf{0},t) [= \sum_{\mathbf{R}} \Theta(\mathbf{0},\mathbf{R},n\,\delta t)]$. We take its inverse and subtract the predicted asymptotic result of $(2/\pi) \ln n$ [in order to compare with Eq. (24)]. The resulting data is plotted against $1/\ln n$, which should yield a straight line, the y intercept of which fixes C(1). As seen in the inset of Fig. 2, the data indeed confirm this expectation, and we find $C(1) \simeq -0.742$. In Fig. 2, we plot the inverse of the data against ln t, along with the predicted asymptotic result. The large difference between the curves indicates the strong role of the corrections (which are of order 10% even for $t \sim 10^6$).

We have also attempted to measure the probability distribution $\mathcal{P}(\phi, \mathbf{0}, t)$ in d=2, as calculated in Sec. III E. In DCI we attempted to measure this quantity in d=1, and met with limited success. The difficulty lies in the fact that the distribution only makes sense for a coarse-grained magnetization density (since a single spin always has a bimodal distribution). Thus to construct \mathcal{P} numerically we must measure the magnetization density for a patch of spins. If the patch is too



FIG. 3. Plot of the probability distribution \mathcal{P} as a function of ϕ for d=2, for times $t=10^3$ (diamonds), $t=10^4$ (plusses), and $t=10^5$ (squares). Note the qualitative similarity (as time proceeds) between these different histograms, and the theoretical prediction [Eq. (33)].

small, the resulting histogram will have too few bins to have any smooth structure, but if the patch is too large, the time scales required to allow the agent to wander far away from the patch and return many times become numerically prohibitive. Thus we must compromise, and we use a square patch of 121 spins. The agent is started on the boundary of the patch to avoid the patch being internally decimated by the transient motion of the agent. The predicted result for \mathcal{P} is given in Eq. (33). It shows three types of behavior set by a crossover time t^* defined by $\beta(t^*) = 1$. For $t \ll t^*$ the distribution has most weight near $\phi = 1$, and then for $t \sim t^*$ the distribution becomes uniform over the entire interval of ϕ . For late times $t \ge t^*$, the distribution approaches the form $\sim 1/\phi$. Our numerical measurement of \mathcal{P} is shown in Fig. 3 for these three time regimes. Good qualitative agreement is found. [Note that the measured distribution has nonzero weight for negative ϕ due to the modest patch size, and is also suppressed near $\phi = 1$ due to internal decimation of the patch magnetization.

In Sec. IV we presented an analysis of the continuum theory with N agents. We described the interference between the agents by a number termed the "disordering efficacy" $\sigma_N(d)$. In particular we calculated the exact value for $\sigma_N(1)$ for N = 2,3, and 4, as given in Eq. (45). We also predicted the large-N form of $\sigma_N(1)$, along with the exact form of the strong logarithmic corrections, as shown in Eq. (52). In two dimensions we argued that $\sigma_N(2) = N$ for all N, but only for times $t \ge e^N$. We have performed numerical simulations of the many agent system in order to test these results. The microscopic rule we use is that there is no hard-core exclusion between the agents, and that for each time step the Nagents are in turn moved to a randomly chosen nearest neighbor site. A spin which is occupied by two agents, say, will thus (for q = 1) be flipped twice in that time step. In Fig. 4 we show the evolution of the ratio of the average global magnetization for N agents as compared to one agent, for d= 1. Asymptotically this ratio is the disordering efficacy by definition. Results are shown for N=2.3, and 4. The curves asymptote to constants as expected, the values of which are



FIG. 4. Plot of the ratio of the average global magnetization for N agents as compared to a single agent, for (in ascending order) N=2, 3, and 4, in d=1. The solid lines are the theoretical predictions [Eq. (45)] for the disordering efficacies, which are the asymptotes of this ratio.

compared to the theoretical predictions. Excellent agreement (to within less than 2%) is found. The numerical values are also given in Table I in section IV. We have measured the disordering efficacy for higher values of N in d=1, and plot our results on a logarithmic scale in Fig. 5. Also shown is the theoretical prediction (52) for large N (where we have included both the leading and sub-leading terms). Again, excellent agreement is found. As argued in Sec. IV, the crossover times in the two-dimensional many agent system are very large, growing exponentially with N. We have attempted to measure $\sigma_N(2)$ numerically, but we have been unable to reach large enough times to see the ratio of the average global magnetizations reach a constant value. However, we may still compare our results to the theoretical prediction, by plotting the predicted value of the asymptote (=N) minus the measured ratio, against $1/\ln t$. If the predicted asymptote is correct, the data so plotted should converge to the origin. In Fig. 6 we present such a plot for N



FIG. 5. Plot of the disordering efficacy σ_N as a function of *N* in d=1. The solid line is the theoretical prediction (52) (including the strong logarithmic corrections).



FIG. 6. Plot of the ratio of the average global magnetization for N agents as compared to a single agent, subtracted from N, as a function of $1/\ln t$ for (in ascending order) N=2, 3, and 4, in d=2. The data in each case are apparently heading for the origin, thus supporting the theoretical prediction $\sigma_N = N$.

=2,3, and 4. The resulting curves appear to be heading for the origin, and therefore constitute numerical support for the prediction $\sigma_N(2)=N$.

In the first part of Sec. V we considered a system with asymmetric switching rates, and argued for the very simple phenomenological description given by the equation of motion (56). This equation may be solved and one obtains the simple relationship (57) between the magnetization density for asymmetric rates and that for symmetric rates. In particular, this result may be used to derive the intuitive result (58) for the proportionality of the average global magnetization for the asymmetric and symmetric cases $[M_A(t)]$ and $M_S(t)$, respectively]. We have performed simulations of the system with asymmetric rates in d = 1. In Fig. 7 we show the ratio of $M_A(t)/M_S(t)$ for different choices of the rates q^+ and q^- . It is seen that the ratio gradually tends to a constant, and that the constant is in agreement with the theoretical prediction (58). We have also tried to test Eq. (57) at the level of the average magnetization density. We have found that the ratio



FIG. 7. Plot of the ratio $M_A(t)/M_S(t)$ vs time for systems in d=1 with asymmetric rates. In ascending order, the values of the rates are $(q^-, q^+) = (0.5, 0.2)$, (0.5, 0.4), (0.4, 0.5), and (0.2, 0.5). The solid lines are the theoretical predictions obtained from Eq. (58).



FIG. 8. Plot of the ratio of magnetization densities $m_A(0,t)/m_S(0,t)$ for d=1 with asymmetric rates $q^-=0.2$ and $q^+=0.8$. The ratio is measured for different patch sizes, the patches having 9, 19, and 101 spins, respectively, from top to bottom. The solid line is the theoretical prediction from Eq. (57).

of the densities is constant for all but the shortest times, which already indicates that the form of the time decay of m(0,t) (relative to its equilibrium value) is insensitive to the asymmetry in the rates. However, we have found that this constant is sensitive to whether we measure the densities for an individual spin, or for a patch. The value of the constant decreases as we increase the patch size. In Fig. 8 we plot the ratio of magnetization densities for asymmetric and symmetric systems, for various patch sizes with the rates $q^+=0.8$ and $q^-=0.2$. As we see, the ratio seems to be approximately constant, and the value decreases as the patch size is increased. Reasonable agreement is found with the prediction of Eq. (57) for the largest patch of 101 spins. This indicates that Eq. (57) is likely to be correct, but only after a large degree of coarse graining.

Finally we turn to the predictions of Sec. V B, which concern the system with symmetric quenched random rates. We investigated this situation numerically in d=1 by measuring the global magnetization and the magnetization density. We perform the averaging in batches; that is, we use the same set of quenched rates for a batch of 1000 systems, which are then averaged over their different BA histories. Then we repeat this for N_b batches (with $N_b \sim 10^3$), thus averaging over different realizations of the quenched rates. We choose the quenched couplings $\{q_r\}$ to be uncorrelated random variables drawn from a uniform distribution in the range [0,1]. In Fig. 9 we show the average of the global magnetization in a system with quenched rates, and compare it to the same quantity in a "pure" system. The two curves become identical for late times (following a power law growth $\sim \sqrt{t}$) indicating that the average global magnetization is insensitive to the quenched rates, as predicted. Turning to the average magnetization density, we found at first that our results could not be fitted to the theoretical prediction (64). However, a closer analysis revealed that great care must be taken in making the comparison. The point is that one wishes to compare theory and simulation for long times when the asymptotic form given in Eq. (64) becomes valid]. However, it turns out that there is a new crossover time t_f within the numerical simulations, beyond which the theoretical prediction is expected to break down. This crossover time emerges



FIG. 9. Plot of the average global magnetization vs time for systems with quenched and pure (homogeneous) switching rates. For long times the functions become identical, growing $\sim \sqrt{t}$.

due to insufficient averaging over batches. In the simulation we average over N_b batches, and thus select N_b couplings from the uniform distribution. There will be a smallest selected value $q_f \sim 1/N_b$ of the coupling, and thus, from this finite sampling of S, we cannot discern if we are really using a uniform distribution with nonzero weight all the way down to q=0, or a distribution with nonzero weight only down to $q = q_f$. In the latter case, one can shown (from either the lattice or continuum theories) that the logarithmic slowing down will vanish after a time $t_f \sim 1/q_f^2$. Thus in our simulations we can only expect to see the logarithmic slowing down for times much less than $t_f \sim N_h^2$. So, in order to make this time window as large as possible it is important to perform as many batch averages as possible, at the expense of averaging over BA histories within a given batch. To make a quantitative comparison to theory, we have calculated the exact form of the logarithmic slowing down from the lattice theory, using the methods described in Sec. III A. After averaging over a uniform distribution of the rates, from the Tauberian theorem we have the exact asymptotic result

$$\left\langle \sum_{R} \Theta(0,R,n\,\delta t) \right\rangle_{S} \sim \frac{\ln(n)}{2(2\,\pi n)^{1/2}} \left[1 + O\left(\frac{1}{\ln n}\right) \right]. \tag{67}$$

In Fig. 10 we plot the average magnetization density, and the above asymptotic result. Good agreement is found, and the difference may be well fitted against the slow logarithmic corrections. Note, for times longer than those shown on the plot, the data fail to agree with the predicted result due to crossover into the regime $t \ge t_f$ in which finite sampling effects dominate.

VII. CONCLUSIONS

This paper has been devoted to a continuation of our previous study DCI [3] of the statistical properties of data corruption due to a Brownian agent (BA). In order to make closer contact to potential physical realizations of this process, we have extended our original study in three directions:



FIG. 10. Plot of the magnetization density at the origin (averaged over both BA histories and the quenched rates) vs time. The solid line is the asymptotic form of the theoretical prediction [Eq. (67)] with no free parameters.

(i) a careful examination of two dimensions, which is the critical dimension for the model, (ii) the competition induced by more than one agent; and (iii) the case of generalized couplings between the BA and its environment. The reader is referred to Sec. VI, which effectively summarizes our results.

The most important conclusion to emerge from this investigation is the robustness of the phenomenological model (7) in accounting for the properties of the data corruption process under a wide range of conditions. The fact that this process may be modeled by such a simple continuum theory gives one confidence in applying similar phenomenological models to more complicated physical situations involving Brownian agents.

As mentioned in Sec. I, the process of disordering of a bistable medium by a Brownian agent has putative applications to data corruption, homogeneous catalysis, and the effect of impurities in biatomic crystals. It remains to be seen whether one can find a solid application of the models studied in DCI and the present work. By having explored more challenging situations such as two dimensions, many agents, and asymmetric and quenched random couplings, we are able to seriously address the practical issues involved in making the connection between the rich statistical properties of our model, and their potential existence in the real world.

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APPENDIX A

In this appendix we outline the procedure of taking the inverse Laplace transform of Eq. (27). Consider the function H(t), the Laplace transform of which has the form

$$\hat{H}(s) = \frac{1}{s[a-b\ln s]},\tag{A1}$$

for sufficiently small $\operatorname{Re}(s)$ such that the apparent pole at $s = e^{a/b}$ may be disregarded (*a* and *b* are positive constants). The inverse Laplace transform is given by the Bromwich integral [16]

$$H(t) = \int_C \frac{ds}{2\pi i} \frac{e^{st}}{s[a-b\ln s]},$$
 (A2)

where the contour C is to be taken up the imaginary axis (note the integration around the simple pole at the origin gives a contribution of zero). The important singularity is the branch point at the origin. We take the cut along the negative real axis and deform the contour around this cut. Integrating across the cut then gives us

$$H(t) = \frac{1}{b} \int_0^\infty \frac{dx}{x} \left[\frac{e^{-xt}}{(\ln x - a/b)^2 + \pi^2} \right].$$
 (A3)

We are interested in the asymptotic form of this integral for large t. In this limit the integral is dominated by small x. We therefore expand the denominator to give

$$H(t) = \frac{1}{b} \int_0^\infty \frac{dx}{x} \frac{e^{-xt}}{(\ln x)^2} \left[1 + \frac{2a}{b\ln x} + O\left(\frac{1}{(\ln x)^2}\right) \right].$$
(A4)

By splitting the range of integration into two pieces $[x \in (0,1/t) \text{ and } x \in (1/t,\infty)]$, we systematically evaluate the terms in the above expansion to find

$$H(t) = \frac{1}{b \ln t} - \frac{(a + \gamma b)}{(b \ln t)^2} + O\left(\frac{1}{(\ln t)^3}\right),$$
(A5)

where $\gamma = 0.57721...$ is Euler's constant. Using the appropriate form for the constants *a* and *b* we arrive at Eq. (28) correct up to $O(1/(\ln t)^3)$.

APPENDIX B

In this appendix we indicate the steps leading from Eq. (42) to Eq. (43). First we factorize the integrand to give

$$\sigma_N(d) = d\Gamma(1 - d/2) \sum_{n=0}^{N-1} J_n(d),$$
(B1)

where

$$J_n(d) = \int_0^\infty dr \ r^{d-1} f(r) m(r)^n,$$
 (B2)

with $f(r) \equiv 1 - m(r)$. We then integrate the above expression by parts using

$$\int^{r} du \, u^{d-1} f(u) = \frac{r^{d} f(r)}{d} - \frac{e^{-r^{2}}}{d\Gamma(1 - d/2)} + \text{const} \,.$$
(B3)

The resulting integral contains a factor of re^{-r^2} which allows one to integrate by parts a second time, thus yielding, after some algebra (for n > 0),

with

$$K_n(d) = \frac{2n(n+1)}{d\Gamma(1-d/2)^2} \int_0^\infty dr \, r^{1-d} e^{-2r^2} m(r)^{n-1}.$$
 (B5)

Substitution of Eq. (B4) into Eq. (B1) yields the result (43), as shown in the main text.

APPENDIX C

In this appendix we indicate the evaluation of $\sigma_N(1)$ as given by Eq. (44) for N=3 and 4. For N=3 we require

$$\sigma_3(1) = \frac{12}{\pi^{1/2}} \int_0^\infty dx \, e^{-2x^2} \operatorname{erf}(x).$$
 (C1)

We introduce the integral

$$I(\beta) \equiv \pi^{1/2} \int_0^\infty dx \ e^{-\beta x^2} \operatorname{erf}(x).$$
 (C2)

Using integration by parts, it is straightforward to show that $I(\beta)$ satisfies the differential equation

$$2\beta \frac{dI}{d\beta} + I = -\frac{1}{(1+\beta)}.$$
 (C3)

This differential equation may then be integrated [using the boundary condition $I(\infty)=0$] to give

$$I(\beta) = \frac{1}{2\beta^{1/2}} \int_{\beta}^{\infty} d\beta' \frac{1}{\beta'^{1/2}(1+\beta')} = \frac{1}{\beta^{1/2}} \sin^{-1} \left(\frac{1}{(1+\beta)^{1/2}}\right).$$
(C4)

Using this result, along with the fact that $\sigma_3(1) = (12/\pi)I(2)$, we have the leftmost result of Eq. (45) as shown in the main text.

The evaluation of $\sigma_4(1)$ proceeds along similar grounds. We define a function

$$L(\beta) = \pi^{1/2} \int_0^\infty dx \, e^{-\beta x^2} \text{erf}(x)^2,$$
 (C5)

which may be shown to satisfy the differential equation

$$2\beta \frac{dL}{d\beta} + L = -\frac{2}{(1+\beta)(2+\beta)^{1/2}}.$$
 (C6)

This equation may be integrated, and the resulting integral may be calculated by elementary methods to yield

$$L(\beta) = \frac{1}{\beta^{1/2}} \sin^{-1} \left(\frac{1}{1+\beta} \right).$$
 (C7)

Using this result, along with the fact that $\sigma_4(1) = (24/\pi)L(2)$, we have the rightmost result of Eq. (45) as shown in the main text.

A similar method may be used to simplify the integrals for higher N, but we have been unable to evaluate these integrals in closed form. Naturally, the integral in Eq. (44) may be evaluated numerically for a given N to any desired precision.

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