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Enhanced rare-region effects in the contact process with long-range correlated disorder

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We investigate the nonequilibrium phase transition in the disordered contact process in the presence of long-range spatial disorder correlations. These correlations greatly increase the probability for finding rare regions that are locally in the active phase while the bulk system is still in the inactive phase. Specifically, if the correlations decay as a power of the distance, the rare-region probability is a stretched exponential of the rare-region size rather than a simple exponential as is the case for uncorrelated disorder. As a result, the Griffiths singularities are enhanced and take a non-power-law form. The critical point itself is of infinite-randomness type but with critical exponent values that differ from the uncorrelated case. We report large-scale Monte Carlo simulations that verify and illustrate our theory. We also discuss generalizations to higher dimensions and applications to other systems such as the random transverse-field Ising model, itinerant magnets, and the superconductor-metal transition.

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I. INTRODUCTION

The effects of quenched spatial disorder on phase transitions have been a topic of great interest for several decades. Initially, research concentrated on classical (thermal) transitions for which many results can be obtained by using perturbative methods adapted from the theory of phase transitions in clean systems (see, e.g., Ref. [1]).

Later, it became clear, however, that many transitions are dominated by the nonperturbative effects of strong, rare disorder fluctuations and the rare spatial regions that support them. Such rare regions can be locally in one phase while the bulk system is in the other. The resulting slow dynamics leads to thermodynamic singularities, now known as the Griffiths singularities [2,3], not just at the transition point but in an entire parameter region around it. Griffiths singularities at generic classical (thermal) phase transitions are very weak and probably unobservable in experiment [4]. In contrast, at many quantum and nonequilibrium phase transitions, the rare regions lead to strong Griffiths effects characterized by nonuniversal power-law singularities of various observables. The critical point itself is of exotic infinite-randomness type and characterized by activated rather than power-law dynamical scaling. This was first demonstrated in the random-transverse field Ising chain using a strong-disorder renormalization group [5,6] as well as heuristic optimal fluctuation arguments and computer simulations [7–9]. Similar power-law Griffiths singularities were also found at the nonequilibrium transition of the disordered contact process [10-12] and at many other quantum and nonequilibrium transitions. In some systems, the rare-region effects are even stronger and destroy the sharp phase transition by smearing [13]. Recent reviews and a classification of rare-region effects can be found, e.g., in Refs. [14].

The majority of the literature on rare regions and Griffiths singularities focuses on uncorrelated disorder. In many physical situations, we can expect, however, that the disorder is correlated in space, for example if it caused by charged impurities. It is intuitively clear that sufficiently long-ranged spatial disorder correlations must enhance the rare-region effects because they greatly increase the probability for finding large atypical rare regions. Rieger and Igloi [15] studied a random transverse-field Ising chain with power-law disorder correlations. They indeed found that sufficiently long-ranged correlations change the universality class of the transition. They also predicted that the Griffiths singularities take the same power-law form as in the case of uncorrelated disorder, but with changed exponents.

In this paper, we investigate the nonequilibrium phase transition in the disordered one-dimensional contact process with power-law disorder correlations by means of optimal fluctuation theory and computer simulations. Our paper is organized as follows. We define the contact process with correlated disorder in Sec. II. In Sec. III, we develop our theory of the nonequilibrium phase transition and the accompanying Griffiths phase. Specifically, we show that the probability of finding a large rare region is a stretched exponential of its size rather than a simple exponential as for uncorrelated disorder. As a result, the Griffiths singularities are enhanced and take a non-power-law form. The critical point itself is of infinite-randomness type but its exponents differ from the uncorrelated case. Section IV is devoted to Monte Carlo simulations that verify and illustrate our theory. In Sec. V, we generalize our results to higher dimensions and other physical systems. We also discuss the relation between the present work and Ref. [15]. We conclude in Sec. VI.

II. CONTACT PROCESS WITH CORRELATED DISORDER

The contact process [16] is a prototypical nonequilibrium many-particle system which can be understood as a model for the spreading of an epidemic. Consider a one-dimensional regular lattice of L sites. Each site can be in one of two states, either inactive (healthy) or active (infected). The time evolution of the contact process is given by a continuous-time Markov process during which active lattice sites infect their nearest neighbors or heal spontaneously. Specifically, an active site becomes inactive at rate μ , while an inactive site becomes active at rate $n\lambda/2$ where n is the number of its active nearest neighbors. The healing rate μ and the infection rate λ are the external control parameters of the contact process. Without loss of generality, μ can be set to unity, thereby fixing the unit of time.

The qualitative behavior of the contact process is easily understood. If healing dominates over infection, $\mu \gg \lambda$, the epidemic eventually dies out completely, i.e., all lattices sites become inactive. At this point, the system is in a fluctuationless state that it can never leave. This absorbing state constitutes the inactive phase of the contact process. In the opposite limit, $\mu \ll \lambda$, the infection never dies out (in the thermodynamic limit $L \to \infty$). The system eventually reaches a steady state in which a nonzero fraction of lattices sites is active. This fluctuating steady state constitutes the active phase of the contact process. The active and inactive phases are separated by a nonequilibrium phase transition in the directed percolation universality class [17-19]. The order parameter of this absorbing-state transition is given by the steady state density $\rho_{\text{stat}} = \lim_{t \to \infty} \rho(t)$ which is the long-time limit of the density of infected sites at time t,

$$\rho(t) = \frac{1}{L} \sum_{i} \langle n_i(t) \rangle. \tag{1}$$

Here, $n_i(t)$ is the occupation of site *i* at time *t*, i.e., $n_i(t) = 1$ if the site is infected and $n_i(t) = 0$ if it is healthy. $\langle ... \rangle$ denotes the average over all realizations of the Markov process.

So far, we have discussed the clean contact process for which λ and μ are spatially uniform. Quenched spatial disorder is introduced by making the infection rate λ_i of site *i* and/or its healing rate μ_i random variables. The correlations of the randomness can be characterized by the correlation function

$$G_{\lambda}(i,j) = [\lambda_i \lambda_j]_{\text{dis}} - [\lambda_i]_{\text{dis}} [\lambda_j]_{\text{dis}}, \qquad (2)$$

where $[\ldots]_{dis}$ denotes the disorder average. The correlation function G_{μ} of the healing rates μ_i can be defined analogously. The existing literature on the disordered contact process mostly considered the case of uncorrelated disorder, $G_{\lambda}(i,j) \sim$ $G_{\mu}(i,j) \sim \delta_{ij}$. In the present paper, we are interested in long-range correlations whose correlation function decays as a power of the distance r_{ij} between the two sites,

$$G_{\lambda}(i,j) \sim G_{\mu}(i,j) \sim r_{ij}^{-\gamma}, \qquad (3)$$

for large r_{ij} . For our analytical calculations we will often use a correlated Gaussian distribution

$$P_G(\lambda_1, \dots, \lambda_L) \sim \exp\left[-\frac{1}{2}\sum_{i,j}(\lambda_i - \bar{\lambda})A_{ij}(\lambda_j - \bar{\lambda})\right]$$
 (4)

of average $\bar{\lambda} = [\lambda_i]_{\text{dis}}$ and covariance matrix $(A^{-1})_{ij} = G_{\lambda}(i, j)$ [20]. Alternatively, we will also use a correlated binary distribution in which λ_i can take values λ and $c\lambda$ with overall probabilities (1 - p) and p, respectively. Here, p and c are constants between 0 and 1.

III. THEORY

A. Rare-region probability

The Griffiths phase in the disordered contact process is caused by rare large spatial regions whose effective infection rate is larger than the bulk average $\bar{\lambda}$. For weak disorder and outside the asymptotic critical region, the effective infection rate can be approximated by

$$\lambda_{RR} \approx \frac{1}{L_{RR}} \sum_{i \in RR} \lambda_i.$$
 (5)

To estimate how the probability distribution of λ_{RR} depends on the rare-region size L_{RR} , we start from the correlated Gaussian (4), introduce λ_{RR} as a new variable, and then integrate out all other random variables. For large L_{RR} and up to subleading boundary terms, this leads to the distribution

$$P(\lambda_{RR}, L_{RR}) \sim \exp\left[-\frac{L_{RR}}{2\tilde{G}(L_{RR})}(\lambda_{RR} - \bar{\lambda})^2\right], \quad (6)$$

where $\tilde{G}(L_{RR})$ is the sum over the correlation function

$$\tilde{G}(L_{RR}) \sim \sum_{j=0}^{L_{RR}/2} G_{\lambda}(0,j).$$
(7)

Two cases need to be distinguished, depending on the value of the decay exponent γ in the correlation function (3). If $\gamma > 1$, the sum $\tilde{G}(L_{RR})$ converges in the limit $L_{RR} \to \infty$. The probability distribution of the effective infection rate λ_{RR} thus takes the asymptotic form

$$P(\lambda_{RR}, L_{RR}) \sim \exp\left[-\frac{1}{2b^2} L_{RR} \left(\lambda_{RR} - \bar{\lambda}\right)^2\right], \qquad (8)$$

where *b* is a constant. This form is identical to the result for uncorrelated or short-range correlated disorder (and agrees with the prediction of the central limit theorem). For $0 < \gamma < 1$, in contrast, the sum $\tilde{G}(L_{RR})$ behaves as $L_{RR}^{1-\gamma}$ for large L_{RR} . Consequently, the probability distribution of λ_{RR} reads

$$P(\lambda_{RR}, L_{RR}) \sim \exp\left[-\frac{1}{2b^2} L_{RR}^{\gamma} (\lambda_{RR} - \bar{\lambda})^2\right].$$
(9)

This is a stretched exponential decay in L_{RR} rather than the simple exponential obtained in (8). In other words, for $0 < \gamma < 1$, the probability for finding a large deviation of λ_{RR} from the average $\bar{\lambda}$ decays much more slowly with rare-region size than in the uncorrelated case.

We have also considered a correlated binary disorder distribution instead of the Gaussian (4). In this case, rare regions can be defined as regions of L_{RR} consecutive sites having the larger of the two infection rates. For uncorrelated disorder, the probability for finding such a region decays as a simple exponential of its size L_{RR} . We have confirmed numerically that the corresponding probability for the power-law correlations (3) with $0 < \gamma < 1$ follows a stretched exponential

$$w(L_{RR}) \sim \exp\left(-cL_{RR}^{\gamma}\right) \tag{10}$$

with the same exponent γ as in Eq. (9).

B. Griffiths phase

We now use the results of Sec. III A to analyze the time evolution of the density of active sites $\rho(t)$ in the Griffiths phase on the inactive side of the nonequilibrium transition. This calculation is a generalization to the case of correlated disorder of the approach of Refs. [21,22].

The rare-region contribution to $\rho(t)$ can be obtained by summing over all regions that are locally in the active phase,

i.e., all regions having $\lambda_{RR} > \lambda_c$. For the correlated Gaussian distribution (4), $\rho(t)$ reads

$$\rho(t) \sim \int_{\lambda_c}^{\infty} d\lambda_{RR} \int_0^{\infty} dL_{RR} P(\lambda_{RR}, L_{RR}) \\ \times L_{RR} \exp[-t/\tau(\lambda_{RR}, L_{RR})].$$
(11)

Here, $P(\lambda_{RR}, L_{RR})$ is the rare-region distribution (8) or (9), depending on the value of γ , and $\tau(\lambda_{RR}, L_{RR})$ denotes the lifetime of the rare region. It can be estimated as follows. As the rare region is locally in the active phase, $\lambda_{RR} > \lambda_c$, it can only decay via an atypical coherent fluctuation of all its sites. The probability for this to happen is exponentially small in the rare-region size [10], resulting in an exponentially large lifetime,

$$\tau(\lambda_{RR}, L_{RR}) = t_0 \exp\left[aL_{RR}\right],\tag{12}$$

where t_0 is a microscopic time scale. The coefficient *a* vanishes at $\lambda_{RR} = \lambda_c$ and increases with increasing λ_{RR} , i.e., the deeper the region is in the active phase, the larger *a* becomes. Because *a* has the dimension of an inverse length, it scales as ξ_{\perp}^{-1} (where ξ_{\perp} is the correlation length) according to finite-size scaling [23],

$$a = a' (\lambda_{RR} - \lambda_c)^{\nu_{0\perp}}.$$
 (13)

Note that $v_{0\perp}$ is the *clean* correlation length exponent unless the rare region is very close to criticality (inside the narrow asymptotic critical region) [24].

In the long-time limit $t \gg t_0$, the integral (11) can be solved in saddle-point approximation. The saddle point equations read

$$\frac{\partial}{\partial L_{RR}} \left[\frac{L_{RR}^{\gamma}}{2b^2} (\lambda_{RR} - \bar{\lambda})^2 + \frac{t}{t_0} e^{-a'(\lambda_{RR} - \lambda_c)^{\nu_{0\perp}} L_{RR}} \right] = 0, \quad (14)$$

$$\frac{\partial}{\partial\lambda_{RR}} \left[\frac{L_{RR}^{\gamma}}{2b^2} (\lambda_{RR} - \bar{\lambda})^2 + \frac{t}{t_0} e^{-a'(\lambda_{RR} - \lambda_c)^{\nu_{0\perp}} L_{RR}} \right] = 0, \quad (15)$$

and yield the saddle point values

$$\lambda_{sp} - \lambda_c = \frac{\gamma \nu_{0\perp}}{2 - \gamma \nu_{0\perp}} (\lambda_c - \bar{\lambda}), \qquad (16)$$

$$L_{sp} \sim (\lambda_c - \bar{\lambda})^{-\nu_{0\perp}} \ln(t/t_0).$$
(17)

Equations (14) to (17) apply to the long-range correlated case $\gamma < 1$; the corresponding relations for the short-range correlated case follow by formally setting $\gamma = 1$.

For the method to be valid, λ_{sp} must be within the integration range of the integral (11). The bulk system is in the inactive phase implying $\bar{\lambda} < \lambda_c$. Moreover, the clean correlation length exponent of the one-dimensional contact process takes the value $\nu_{0\perp} \approx 1.097$ [25]. Consequently, the saddle-point value λ_{sp} is larger than λ_c , as required. Inserting the saddle-point values into the integrand yields

$$\rho(t) \sim \exp\left[-\frac{1}{z'}\left(\ln\frac{t}{t_0}\right)^{\gamma}\right],\tag{18}$$

where

$$z' \sim (\lambda_c - \bar{\lambda})^{\gamma \nu_{0\perp} - 2} \tag{19}$$

plays the role of a dynamical exponent in the Griffiths phase. In the short-range correlated case, γ is formally 1. Thus, Eq. (18)

reproduces the well-known power-law Griffiths singularity of density in this case [10–12]. In contrast, in the long-range correlated case, $\gamma < 1$, the decay of the density is slower than any power. Long-range disorder correlations thus lead to a qualitatively enhanced Griffiths singularity.

The above derivation started from the correlated Gaussian distribution (4). However, an analogous calculation can be performed for a correlated binary distribution by combining the rare-region probability (10) with the rare-region life-time (12). Solving the resulting integral over L_{RR} in saddle-point approximation leads to the same functional form (18) of the Griffiths singularity, with

$$z' = a^{\gamma}/c. \tag{20}$$

If the rare regions are not in the active phase but right at the critical point, their decay time depends on their size via the power law $\tau(\lambda_c, L_{RR}) \sim L_{RR}^{z_0}$ rather than the exponential (12). Here, $z_0 \approx 1.581$ is the clean dynamical exponent. For a correlated binary disorder distribution, this can be achieved by tuning the stronger of the two infection rates to the clean critical value. Repeating the saddle-point integration for this case gives a stretched exponential density decay

$$\ln \rho(t) \sim -t^{\gamma/(\gamma+z_0)}.$$
(21)

As before, the short-range correlated case is recovered by formally setting $\gamma = 1$.

Griffiths singularities in other quantities can be derived in an analogous manner. Consider, for example, systems that start from a single active site in an otherwise inactive lattice. In this situation, the central quantity is the survival probability $P_s(t)$ that measures how likely the system is to be still active (i.e., to contain at least one active site) at time t. For directed percolation problems such as the contact process, the survival probability behaves in the same way as the density of active sites [17]. Thus, the time dependencies (18) and (21) derived for $\rho(t)$ also hold for $P_s(t)$.

We emphasize that the dependencies of the Griffiths dynamical exponent z' on the distance from criticality given in (19) and (20) hold outside the asymptotic critical region of the disordered contact process. The analysis of the critical region itself requires more sophisticated methods that will be discussed in the next section.

C. Critical point

After discussing the Griffiths phase, we now turn to the critical point of the disordered contact process itself. The contact process with spatially *uncorrelated* disorder features an exotic infinite-randomness critical point in the universality class of the (uncorrelated) random transverse-field Ising chain [11,12]. Is this critical point stable or unstable against the long-range power-law disorder correlations (3)? According to Weinrib and Halperin's generalization [26] of the Harris criterion, power-law disorder correlations are irrelevant if the decay exponent γ fulfills the inequality

$$\gamma > 2/\nu_{\perp}^{\rm unc},\tag{22}$$

where ν_{\perp}^{unc} is the correlation length exponent for uncorrelated disorder. If this inequality is violated, the correlations are relevant, and the critical behavior must change. The correlation

length exponent of the contact process with uncorrelated disorder takes the value $\nu_{\perp}^{unc} = 2$ [5,11]. The long-range correlations are thus irrelevant if $\gamma > 1$ and relevant if $\gamma < 1$. Interestingly, this is the same criterion as we derived for the Griffiths phase in Secs. III A and III B.

What is the fate of the transition in the long-range correlated case $\gamma < 1$? As long-range correlations tend to further enhance the disorder effects, we expect the critical behavior to be of infinite-randomness type, but with modified critical exponents that produce stronger singularities. In the strong-disorder regime close to criticality, the behavior of the contact process is identical to that of a random transverse-field Ising chain as both are governed by the same strong-disorder renormalization group recursion relations [5,11]. Note that the application of these recursions is justified even in the presence of disorder correlations provided that the distributions of the logarithms of μ and λ become infinitely broad. The transverse-field Ising chain with long-range correlated disorder was solved by Rieger and Igloi [15] who mapped the problem onto fractional Brownian motion. They found an exact result for the tunneling exponent ψ which relates correlation length ξ_{\perp} and correlation time ξ_t via $\ln(\xi_t/t_0) \sim \xi_{\perp}^{\psi}$. For $\gamma > 1$, it takes the uncorrelated value $\psi = 1/2$ while it is given by $\psi = 1 - \gamma/2$ for $\gamma < 1$. The correlation length exponent ν_{\perp} takes the value 2 for $\gamma > 1$ as for uncorrelated disorder. For $\gamma < 1$, it reads $\nu_{\perp} = 2/\gamma$ in agreement with general arguments by Weinrib and Halperin [26]. A third exponent is necessary to define a complete set; Rieger and Igloi numerically calculated the scale dimension β/ν_{\perp} of the order parameter and found it to decay continuously from its uncorrelated value $(3 - \sqrt{5})/4$ (taken for all $\gamma > 1$) to 0 (for $\gamma = 0$).

A qualitative understanding of these results in the context of the contact process can be obtained from simple arguments based on the strong-disorder recursion relations [11] even though a closed form solution of the renormalization group does not exist for the case of long-range correlated disorder [27]. Imagine performing a (large) number of strongdisorder renormalization group steps, iteratively removing the largest decay rates μ_i and infection rates λ_i . The resulting chain will consist of surviving sites (representing clusters of original sites) whose effective decay rate can be estimated as

$$\mu_{\rm eff} = C_{\mu} \frac{\mu_1 \dots \mu_L}{\lambda_1 \dots \lambda_{L-1}} \tag{23}$$

and long bonds with effective infection rates

$$\lambda_{\rm eff} = C_{\lambda} \frac{\lambda_1 \dots \lambda_L}{\mu_1 \dots \mu_{L-1}},\tag{24}$$

where *L* is the size of the cluster or bond. In the strongdisorder limit, the prefactors C_{μ} and C_{λ} provide subleading corrections only. $\ln \mu_{\text{eff}}$ and $\ln \lambda_{\text{eff}}$ can thus be understood as the displacements of correlated random walks,

$$\ln \mu_{\rm eff} \sim \sum_{i=1}^{L-1} \ln(\mu_i/\lambda_i), \qquad \ln \lambda_{\rm eff} \sim \sum_{i=1}^{L-1} \ln(\lambda_i/\mu_i). \quad (25)$$

Right at criticality, these random walks have to be (asymptotically) unbiased because healing and infection remain competing in the limit $L \rightarrow \infty$. The typical values $\ln \mu_{typ}$ and $\ln \lambda_{typ}$ of the cluster healing and infection rates can be

estimated from the variance of the random walk displacements giving

$$\ln \mu_{\rm typ}| \sim |\ln \lambda_{\rm typ}| \sim \sqrt{L \, \tilde{G}(L)} \sim \begin{cases} L^{1/2} & (\gamma > 1) \\ L^{1-\gamma/2} & (\gamma < 1) \end{cases}$$
(26)

for large L. Here, $\tilde{G}(L)$ is the sum over the disorder correlation function defined in Eq. (7). This estimate thus reproduces the values of ψ quoted above [28].

Moving away from criticality introduces a bias into the random walks. The crossover from critical to off-critical behavior occurs when the displacement due to the bias becomes larger than the displacement (26) due to the randomness. The bias term scales as $|\lambda - \lambda_c|L$. We thus obtain a crossover length

$$L_x \sim \begin{cases} |\lambda - \lambda_c|^{-2} & (\gamma > 1) \\ |\lambda - \lambda_c|^{-2/\gamma} & (\gamma < 1) \end{cases}$$
(27)

in agreement with the quoted values of v_{\perp} .

IV. MONTE CARLO SIMULATIONS

A. Overview

We now turn to large-scale Monte Carlo simulations of the one-dimensional contact process with power-law correlated disorder. We use the same numerical implementation of the contact process as in earlier studies with uncorrelated disorder in one, two, three, and five dimensions in Refs. [12,22,29,30]. It is based on an algorithm suggested by Dickman [31]: The simulation starts at time t = 0 from an initial configuration of active and inactive sites and consists of a sequence of events. During each event an active site *i* is chosen at random from a list of all N_a active sites. Then a process is selected, either infection of a neighbor with probability $\lambda_i/(1 + \lambda_i)$ or healing with probability $1/(1 + \lambda_i)$. For infection, either the left or the right neighbor are chosen with probability 1/2. The infection succeeds if this neighbor is inactive. The time is then incremented by $1/N_a$.

Using this algorithm, we have simulated long chains for times up to $t = 10^7$. All production runs use $L = 2^{20} \approx 10^6$ sites with periodic boundary conditions, and the results are averages over large numbers of disorder configurations; precise data will be given below.

The random infection rates λ_i are drawn from a correlated binary distribution in which λ_i can take values λ and $c\lambda$ with overall probabilities (1 - p) and p, respectively. Here, p and c are constants between 0 and 1. To generate these correlated random variables, we employ the Fourier-filtering method [32]. It starts from uncorrelated Gaussian random numbers u_i and turns them into correlated Gaussian random numbers v_i characterized by the (translationally invariant) correlation function $G_{\lambda}(i, j)$. This is achieved by transforming the Fourier components \tilde{u}_q of the uncorrelated random numbers according to

$$\tilde{v}_q = [\tilde{G}(L,q)]^{1/2} \tilde{u}_q, \qquad (28)$$

where $\tilde{G}(L,q)$ is the Fourier transform of $G_{\lambda}(i,j)$. We parametrize our long-range correlations by the function

$$G_{\lambda}(i,j) = [1 + (i-j)^2]^{-\gamma/2}$$
(29)

with periodic boundary conditions using the minimum image convention. Simulations are performed for $\gamma = 1.5, 0.8, 0.6$, and 0.4. To arrive at binary random variables, the correlated Gaussian random numbers v_i then undergo binary projection: the infection rate λ_i takes the value λ ("strong site") if v_i is greater than a composition-dependent threshold and the value $c\lambda$ with 0 < c < 1 ("weak site") if v_i is less than the threshold. We chose a concentration p = 0.8 of weak sites and a strength c = 0.2 in all simulations. While the binary projection changes the details of the disorder correlations, the functional form of the long-distance tail remains unchanged.

Most of our simulations are spreading runs that start from a single active site in an otherwise inactive lattice; we monitor the survival probability $P_s(t)$, the number of sites $N_s(t)$ of the active cluster, and its (mean-square) radius R(t). Within the activated scaling scenario [11,12] associated with an infiniterandomness critical point, these quantities are expected to display logarithmic time dependencies,

$$P_s \sim \left[\ln(t/t_0)\right]^{-\delta},\tag{30}$$

$$N_s \sim \left[\ln(t/t_0)\right]^{\bar{\Theta}},\tag{31}$$

$$R \sim [\ln(t/t_0)]^{1/\psi}.$$
 (32)

The exponents $\overline{\delta}$ and $\overline{\Theta}$ can be expressed in terms of the scale dimension β/ν_{\perp} of the order parameter and the tunneling exponent ψ as $\overline{\delta} = \beta/(\nu_{\perp}\psi)$ and $\overline{\Theta} = 1/\psi - 2\overline{\delta}$ [12].

B. Results: Critical behavior

We start by considering the case $\gamma = 1.5$. According to the theory laid out in Sec. III, the power-law disorder correlations are irrelevant for $\gamma > 1$. We therefore expect the critical behavior for $\gamma = 1.5$ to be identical to that of the random contact process with uncorrelated disorder which features an infinite-randomness critical point in the universality class of the (uncorrelated) random transverse-field Ising chain [11,12]. Its critical exponents are known exactly, their numerical values read $\beta = 0.38197$, $\nu_{\perp} = 2$, $\psi = 0.5$, $\bar{\delta} = 0.38197$, and $\bar{\Theta} = 1.2360$ [5,6].

To test these predictions, we analyze the time evolution of P_s , N_s , and R in Fig. 1. Specifically, the figure presents plots of $P_s^{-1/\bar{\delta}}$, $N_s^{1/\bar{\Theta}}$, and R^{ψ} vs $\ln(t)$ using the theoretically predicted exponent values. In such plots, the critical time dependencies (30)–(32) correspond to straight lines independent of the unknown value of the microscopic time scale t_0 . The plots show that the data for infection rate $\lambda = 11.44$ follow the predicted time dependencies (30)–(32) over more than four orders of magnitude in time. We thus identify $\lambda_c = 11.44$ (6) as the critical infection rate (the number in brackets is an estimate of the error of the last digit); and we conclude that the critical behavior for $\gamma = 1.5$ is indeed identical to that of the contact process with uncorrelated disorder.

We now turn to $\gamma < 1$, for which the long-range correlations are expected to change the critical behavior. A complete set of exponents is not known analytically in this case; the data analysis is therefore more complicated than for $\gamma > 1$. As we do have an analytical value for the tunneling exponent,



FIG. 1. (Color online) Time evolution of the number of active sites N_s , the survival probability P_s , and the radius of the active cloud *R* for the disordered contact process with power-law disorder correlations characterized by a decay exponent $\gamma = 1.5$. The data are averages over up to 40 000 samples with 100 individual runs per sample. The critical exponents are fixed at their uncorrelated values $\psi = 0.5$, $\bar{\delta} = 0.38197$, and $\bar{\Theta} = 1.2360$.

 $\psi = 1 - \gamma/2$, we can graph R^{ψ} vs ln(*t*), to find the critical point. Figure 2 shows the corresponding plot for $\gamma = 0.4$. The data at $\lambda = 11.6$ follow the predicted time dependence (32) for more than three orders of magnitude in time. We thus identify $\lambda_c = 11.6(2)$ as the critical infection rate. Analogous plots for $\gamma = 0.8$ and 0.6 give infection rates of $\lambda_c = 11.3(2)$ and $\lambda_c = 11.4(2)$, respectively.

Alternatively, we can employ a version of the method used in Refs. [29,30] that allows us to eliminate the unknown microscopic time scale t_0 from the analysis. It is based on the observation that t_0 takes the same value in all of the quantities (because it is related to the basic energy scale of the underlying renormalization group). Thus, if we plot $N_s(t)$ vs $P_s(t)$, the critical point corresponds to power-law behavior, and t_0 drops out. The same is true for other combinations of observables. Specifically, by combining Eqs. (30), (31), and (32), we see that $N_s/P_s^2 \propto R$ at criticality. Thus, identifying straight lines in



FIG. 2. (Color online) Time evolution of the radius of the active cloud *R* for $\gamma = 0.4$. The data are averages over about 30 000 samples with 100 individual runs per sample. The tunneling exponent is set to its analytical value $\psi = 1 - \gamma/2 = 0.8$.

plots of N_s/P_s^2 vs *R* allows us to find the critical point without needing a value for t_0 . Figure 3 shows such a plot for $\gamma = 0.8$; and we have created analogous plots of $\gamma = 0.6$ and 0.4. They give the same critical infection rates, $\lambda_c = 11.3(2)$ (for $\gamma = 0.8$), $\lambda_c = 11.4(2)$ (for $\gamma = 0.6$), and $\lambda_c = 11.6(2)$ (for $\gamma = 0.4$) as the plots of R^{ψ} vs ln(*t*). Interestingly, within their numerical errors, λ_c does not depend on the decay exponent γ of the disorder correlations.

Once the critical point is identified, we can verify and/or find critical exponents by analyzing the time evolutions of P_s , N_s , and R. Figure 4 displays $P_s^{-1/\delta}$, $N_s^{1/\Theta}$, and R^{ψ} vs $\ln(t)$ at criticality for $\gamma = 0.8$. The tunneling exponent ψ is set to its theoretical value $1 - \gamma/2$ while δ and Θ are determined from the data by requiring that the corresponding curves become straight lines for large times. The data follow the predicted logarithmic time dependencies (30), (31), and (32) over about four orders of magnitude in time. This not only confirms the theoretical value of ψ , it also allows us to extract estimates the scale dimension β/ν_{\perp} of the order parameter from both $\bar{\delta}$ and $\bar{\Theta}$. We have performed the same analysis also for $\gamma = 0.6$ and $\gamma = 0.4$.



FIG. 3. (Color online) N_s/P_s^2 vs *R* for a correlation decay exponent $\gamma = 0.8$. The data are averages over about 20 000 samples with 100 individual runs per sample. The maximum time is 10⁶ for all curves except the critical one, $\lambda = 11.3$, for which it is 10⁷.



FIG. 4. (Color online) $N_s^{1/\bar{\Theta}}$, $P_s^{-1/\bar{\delta}}$, and R^{ψ} vs ln(*t*) at criticality for a correlation decay exponent $\gamma = 0.8$. Here, $\psi = 0.6$ is set to its theoretical value while $\bar{\delta} = 0.269$ and $\bar{\Theta} = 0.982$ are determined from the data by requiring that the corresponding curves become straight lines for large times.

The resulting exponent values are summarized in Table I. The uncertainty of $\overline{\delta}$ and $\overline{\Theta}$ can be roughly estimated from the hyperscaling relation $\overline{\Theta} + 2\overline{\delta} = 1/\psi$. The exponents for $\gamma = 0.6$ and $\gamma = 0.4$ fulfill this relation in good approximation (less than 4% difference between the left and the right sides). For $\gamma = 0.8$, the agreement is not quite as good. As $\gamma = 0.8$ is close to the marginal value of 1, this may be caused by a slow crossover from the short-range correlated fixed point to the long-range correlated one.

The values of the scale dimension of the order parameter, β/ν_{\perp} , are in reasonable agreement with those calculated by Rieger and Igloi from the average persistence of a Sinai random walker (see inset of Fig. 1 of Ref. [15]).

To obtain a complete set of exponents, we also analyze offcritical data. Figure 5 shows a double-logarithmic plot of P_s vs R for decay exponent $\gamma = 0.8$. The plot allows us to determine the crossover radius R_x at which the survival probability of slightly off-critical curves has dropped to half of its critical value. According to scaling, the crossover radius must depend on the distance from criticality via $R_x \sim |\lambda - \lambda_c|^{-\nu_{\perp}}$. The inset of Fig. 5 shows that our data indeed follow this power law with the predicted exponent $\nu_{\perp} = 2/\gamma = 2.5$.

TABLE I. Critical exponents of the one-dimensional contact process with power-law correlated disorder. The exponents v_{\perp} and ψ (above the horizontal line) are known analytically, as are all exponents in the short-range case $\gamma > 1$. The exponents $\overline{\delta}$ and $\overline{\Theta}$ for $\gamma < 1$ stem from fits of our data. The scale dimension β/v_{\perp} of the order parameter can be extracted from both $\overline{\delta}$ and $\overline{\Theta}$, the data in the table are averages of the two values.

Exponent	$\gamma > 1$	$\gamma = 0.8$	$\gamma = 0.6$	$\gamma = 0.4$
$\overline{\nu_{\perp}}$	2	2.5	3.33	5
ψ	0.5	0.6	0.7	0.8
$\bar{\delta}$	0.3820	0.27	0.20	0.13
$\bar{\Theta}$	1.2360	0.98	0.98	1.01
eta/ u_{ot}	0.1910	0.18	0.14	0.10



FIG. 5. (Color online) Double-log plot of P_s vs R for decay exponent $\gamma = 0.8$ and several infection rates λ at and below the critical rate $\lambda_c = 11.3$. The dash-dotted line shows $P_s/2$ for $\lambda = \lambda_c$. The crossing points of the dash-dotted line with the off-critical data determines the crossover radius R_x . Inset: R_x vs $|\lambda - \lambda_c|$. The solid line is a power-law fit to $R_x \sim |\lambda - \lambda_c|^{-\nu_{\perp}}$ with an exponent $\nu_{\perp} = 2.5$.

C. Results: Griffiths phase

We now turn to the Griffiths phase $\lambda_{c0} \leq \lambda < \lambda_c$ where $\lambda_{c0} \approx 3.298$ is the critical infection rate of the clean contact process containing only "strong" sites (p = 0).

Right at the clean critical point, $\lambda = \lambda_{c0}$, the time evolution of the survival probability is predicted to follow the stretched exponential (21) in the long-time limit. Our corresponding data for $\gamma = 0.8$, 0.6, and 0.4 are plotted in Fig. 6. For all γ , the data indeed follow stretched exponentials over more than six orders of magnitude in P_s . The exponent y decreases with decreasing γ , as predicted in (21). The actual numerical values of y are somewhat larger than the prediction $y = \gamma/(\gamma + z_0)$. We attribute this to the fact that, due to the rapid decay of P_s ,



FIG. 6. (Color online) Time evolution of the survival probability P_s at the clean critical infection rate $\lambda_{c0} = 3.298$ for decay exponents $\gamma = 0.8$, 0.6, and 0.4. The data are averages over 2×10^4 to 10^5 samples with at least 10^4 individual runs per sample. The experimental values y_{Ex} are determined by requiring that the respective curves become straight lines for large times, implying a stretched exponential time dependence, $\ln P_s \sim t^y$. The theoretical values follow from Eq. (21) which gives $y_{Th} = \gamma/(\gamma + z_0)$.



FIG. 7. (Color online) Double-log plot of the survival probability P_s vs time t for decay exponent $\gamma = 1.5$ at several infection rates inside the Griffiths phase, $\lambda_{c0} < \lambda < \lambda_c$. The data are averages over up to 40 000 samples with 100 individual runs per sample.

the data are taken at rather short times ($t \leq 10^3$). Thus, they probably have not reached the true asymptotic regime, yet.

We now move into the bulk of the Griffiths phase, $\lambda_{c0} < \lambda < \lambda_c$. Here, we wish to contrast the conventional powerlaw Griffiths singularity with the unusual non-power-law form (18). Figure 7 shows the survival probability as a function of time for a decay exponent $\gamma = 1.5$ and several infection rates inside the Griffiths phase. After initial transients, all data follow power laws (represented by straight lines) over serval orders of magnitude in P_s and/or t. For $\gamma = 1.5$, we thus find the same type of power-law Griffiths singularity as in the case of uncorrelated or short-range correlated disorder.

In the long-range correlated regime, $\gamma < 1$, we expect the survival probability to follow Eq. (18) rather than a power law. This prediction is tested in Fig. 8 which shows of P_s vs t for decay exponent $\gamma = 0.4$. In the double-logarithmic plot in the inset, all data show pronounced upward curvatures rather than



FIG. 8. (Color online) Survival probability P_s vs time t for decay exponent $\gamma = 0.4$ at several infection rates inside the Griffiths phase, $\lambda_{c0} < \lambda < \lambda_c$, plotted such that Eq. (18) yields straight lines (the values of t_0 are fit parameters). The data are averages over up to 20 000 samples with 100 individual runs per sample. Inset: Double-log plot of the same data to test for power-law behavior.

the straight lines expected for power laws. In contrast, when plotted as $\ln P_s$ vs $[\ln(t/t_0)]^{\gamma}$ (where t_0 is a fit parameter) in the main panel of the figure, all curves become straight for sufficiently long times implying that the long-time behavior of P_s indeed follows Eq. (18).

We have produced analogous plots for decay exponents $\gamma = 0.6$ and 0.8. As γ decreases from 1 towards 0, the upward curvature in the double-logarithmic plots becomes bigger, reflecting stronger and stronger deviations from power-law behavior, as expected. In contrast, Eq. (18) describes the long-time behavior of all data very well, confirming our theory.

V. GENERALIZATIONS

A. Higher dimensions

It this section, we generalize our results to the contact process in higher dimensions d > 1. The theory of Sec. III A can be easily adapted, yielding the rare-region distribution

$$P(\lambda_{RR}, L_{RR}) \sim \begin{cases} \exp\left[-\frac{1}{2b^2} L_{RR}^d (\lambda_{RR} - \bar{\lambda})^2\right] & (\gamma > d) \\ \exp\left[-\frac{1}{2b^2} L_{RR}^\gamma (\lambda_{RR} - \bar{\lambda})^2\right] & (\gamma < d) \end{cases}$$
(33)

This means that the functional form of the rare-region distribution is identical to the case of uncorrelated disorder as long as $\gamma > d$. For $\gamma < d$, the probability for finding a rare region decays more slowly with its size. In terms of the volume L_{RR}^d , it is given by a stretched exponential rather than a simple one.

Using this result, we now repeat the calculation of Sec. III B for general d. For $\gamma < d$, the resulting long-time behavior of the density of active sites in the Griffiths phase reads

$$\rho(t) \sim \exp\left[-\frac{d}{z'} \left(\ln\frac{t}{t_0}\right)^{\gamma/d}\right]$$
(34)

with

$$z' \sim d(\lambda_c - \bar{\lambda})^{\gamma \nu_{0\perp} - 2}.$$
(35)

As in one dimension, the decay described by Eq. (34) is slower than any power. For $\gamma > d$, in contrast, we find the usual power-law behavior. Equation (34) also holds for a correlated binary distribution with $z' = d a^{\gamma/d}/c$. The behavior right at the boundary of the Griffiths phase (when the stronger of the two infection rates of the binary distribution is tuned to the clean critical value) takes the form (21) for all dimensions.

The behavior of the critical point itself will again be of infinite-randomness type, but for a sufficiently small correlation decay exponent $\gamma < 2/\nu_{\perp}^{unc}$, the critical exponents will differ from those of the contact process with uncorrelated disorder (which were found numerically in Ref. [29] for two dimensions and in Ref. [30] for three dimensions). The correlation length exponent will take the value $\nu_{\perp} = 2/\gamma$ [26]; other exponents need to be found numerically [15].

It is interesting to compare the relevance criteria of the long-range correlations in the Griffiths phase and at the critical point. In one dimension, the long-range correlations become relevant for $\gamma < 1$ both in the Griffiths phase and at criticality (because the correlation length exponent of the one-dimensional contact process with uncorrelated disorder has the value $\nu_{\perp}^{unc} = 2$, saturating the Harris criterion). In dimensions d > 1, the two criteria differ. The uncorrelated correlation length exponent is larger than 2/d [29,30]. Thus, the long-range correlations do not become relevant for $\gamma < d$ but only if $\gamma < 2/\nu_{\perp}^{unc} < d$. In contrast, the long-range correlations become relevant for $\gamma < d$ in the Griffiths phase. Consequently, for d > 1, we expect a (narrow) range of decay exponents γ for which the long-range correlations are relevant in the Griffiths phase but irrelevant at criticality. The fate of the system in this subtle regime remains a task for the future.

B. Other systems

The theory of Secs. III A and III B and its generalization to higher dimensions have produced enhanced *non-power-law* Griffiths singularities for sufficiently long-ranged disorder correlations. Are these results restricted to the contact process or do they apply to other systems as well? In this section, we show that they hold for a broad class of systems in which the characteristic energy or inverse time scale of a rare region depends exponentially on its volume (class B of the rare-region classification of Refs. [14,33]). In addition to the contact process, this class contains, e.g., the random transverse-field Ising model, Hertz's model of the itinerant antiferromagnetic quantum phase transition, and the pairbreaking superconductor-metal quantum phase transition.

To demonstrate the enhanced Griffiths singularities, we generalize the calculation of the rare-region density of states developed in Ref. [21] to the case of our power-law correlated disorder. Consider a disordered system with rare regions whose characteristic energy ϵ depends on their volume via

$$\epsilon(\lambda_{RR}, L_{RR}) = \epsilon_0 \exp\left[-aL_{RR}^d\right]. \tag{36}$$

Here, ϵ_0 is a microscopic energy scale, and $a = a'(\lambda_{RR} - \lambda_c)^{dv_{0\perp}}$ with λ representing the parameter that tunes the system through the phase transition. In the contact process, $\epsilon = 1/\tau$ is the inverse life time of a rare region; in the transverse-field Ising model, it represents its energy gap. We can derive a rare-region density of states by summing over all values of λ_{RR} and L_{RR} ,

$$\tilde{\rho}(\epsilon) \sim \int_{\lambda_c}^{\infty} \int_0^{\infty} dL_{RR} P(\lambda_{RR}, L_{RR}) \,\delta[\epsilon - \epsilon(\lambda_{RR}, L_{RR})]$$
(37)

with the Gaussian rare-region probability $P(\lambda_{RR}, L_{RR})$ from Eq. (33). After carrying out the integral over L_{RR} with the help of the δ function, the remaining λ_{RR} integral can be performed in saddle-point approximation in the limit $\epsilon \rightarrow 0$. For $\gamma < d$, the resulting density of states takes the form

$$\tilde{\rho}(\epsilon) \sim \frac{1}{\epsilon} \exp\left[-\frac{d}{z'} \left(\ln\frac{\epsilon_0}{\epsilon}\right)^{\gamma/d}\right]$$
(38)

with z' given by Eq. (35). For $\gamma > d$, in contrast, we recover the usual power-law behavior $\tilde{\rho}(\epsilon) \sim \epsilon^{d/z'-1}$. If we start from correlated binary disorder rather than a Gaussian distribution, we arrive at the same expression (38) for the density of states with $z' = da^{\gamma/d}/c$. Equation (38) shows that the Griffiths singularities are qualitatively enhanced for $\gamma < d$ as the density of states diverges as $1/\epsilon$ times a function that is slower than any power law.

Griffiths singularities in other observables can be calculated from appropriate integrals of $\tilde{\rho}(\epsilon)$. For example, our results for the density of active sites in the contact process can be reproduced by $\rho(t) \sim \int d\epsilon \ \tilde{\rho}(\epsilon) \exp(-\epsilon t)$. In the case of the random transverse-field Ising model, we can calculate (see, e.g., Ref. [14]) the temperature dependence of observables such as the entropy $S(T) \sim \int_0^T d\epsilon \ \tilde{\rho}(\epsilon)$, the specific heat $C(T) = T(\partial S/\partial T)$, and the susceptibility $\chi(T) \sim (1/T) \int_0^T d\epsilon \ \tilde{\rho}(\epsilon)$. For $\gamma < d$, we find

$$S(T) \sim C(T) \sim T\chi(T) \sim \exp\left[-\frac{d}{z'}\left(\ln\frac{\epsilon_0}{T}\right)^{\gamma/d}\right].$$
 (39)

Analogously, the magnetization in a longitudinal field H scales as

$$M(H) \sim \exp\left[-\frac{d}{z'} \left(\ln\frac{\epsilon_0}{H}\right)^{\gamma/d}\right].$$
 (40)

Let us compare these results with those obtained in Ref. [15]. Equations (38), (39), and (40) yield Griffiths singularities that are qualitatively stronger than power laws. In contrast, Rieger and Igloi obtained the usual power-law Griffiths singularities, albeit with changed exponents. We believe that this discrepancy arises from the fact that Rieger and Igloi assumed that the probability for finding a strongly coupled cluster of size L_{RR} in *d* dimensions takes the same functional form, $\exp(-cL_{RR}^d)$, as for uncorrelated disorder. Our calculations show that this assumption is justified for $\gamma > d$. For $\gamma < d$, however, the rare-region probability decays as $\exp(-cL_{RR}^{\gamma})$, i.e., more slowly than in the uncorrelated case.

VI. CONCLUSIONS

To summarize, we have studied the effects of long-range spatial disorder correlations on the critical behavior and the Griffiths singularities in the disordered one-dimensional contact process. As long as the correlations decay faster as $1/r_{ij}$ with the distance r_{ij} between the sites, the correlations are irrelevant both at criticality and in the Griffiths phase. This means that both the critical and the Griffiths singularities are identical to those of the contact process with uncorrelated disorder. If the correlations decay more slowly than $1/r_{ij}$, the universality class of the critical point changes, and the Griffiths singularities take an enhanced, non-power-law form.

What is the reason for the enhanced singularities? As positive spatial correlations imply that neighboring sites have similar infection rates, it is intuitively clear that sufficiently region behaves as a stretched exponential of its size (rather than the simple exponential found for uncorrelated and short-range correlated disorder). Note that similar stretched exponentials have also been found in the distributions of rare events in long-range correlated time series [34,35].

Our theory of the Griffiths phase is easily generalized to higher dimensions. In general dimension d, the rare-region probability decays exponentially with the rare-region volume as long as the disorder correlations decay faster than $1/r_{ij}^d$. As a result, the Griffiths singularities take the usual power-law form. For correlations decaying slower than $1/r_{ij}^d$, the rare-region probability becomes a stretched exponential of the volume, leading to enhanced, non-power-law Griffiths singularities.

Moreover, as shown in Sec. V B, the theory is not restricted to the contact process. It holds for all systems for which the characteristic energy (or inverse time) of a rare region depends exponentially on its volume, i.e., for all systems in class B of the rare-region classification of Refs. [14,33]. The random transverse-field Ising model is a prototypical example in the class. Our theory predicts that the character of its Griffiths singularities changes from the usual power-law behavior for correlations decaying faster than $1/r_{ij}^d$ to the enhanced non-power-law forms (39) and (40) for correlations decaying slower than $1/r_{ij}^d$.

What about systems in the other classes, class A and class C, of the rare-region classification of Refs. [14,33]? The rare regions in systems belonging to class A have characteristic energies that decrease as a power of their sizes. Using this power-law dependence rather than the exponential (36) in the calculation of Sec. V B yields an exponentially small density of states. We conclude that rare-region effects in class A remain very weak, even in the presence of long-range disorder correlations. Rare regions in systems belonging to class C can undergo the phase transition by themselves, independently from the bulk system. This results in a smearing of the global phase transition. Svoboda et al. [36] considered the effects of spatial disorder correlations on such smeared phase transitions. They found that even short-range correlations can have dramatic effects and qualitatively change the behavior of observable quantities compared to the uncorrelated case. This phenomenon may have been observed in $Sr_{1-x}Ca_xRuO_3$ [37].

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