Persistence in the Voter model: continuum reaction-diffusion approach

M Howard†, C Godrèche‡§

†Center for Chaos and Turbulence Studies, The Niels Bohr Institute, Blegdamsvej 17, 2100 Copenhagen Ø, Denmark
‡Service de Physique de l’État Condensé, CEA-Saclay, 91191 Gif sur Yvette, France
§Laboratoire de Physique Théorique et Modélisation, Université de Cergy-Pontoise, France

Abstract

We investigate the persistence probability in the Voter model for dimensions \( d \geq 2 \). This is achieved by mapping the Voter model onto a continuum reaction–diffusion system. Using path integral methods, we compute the persistence probability \( r(q, t) \), where \( q \) is the number of “opinions” in the original Voter model. We find \( r(q, t) \sim \exp[-f_2(q)(\ln t)^2] \) in \( d = 2 \); \( r(q, t) \sim \exp[-f_d(q)t^{(d-2)/2}] \) for \( 2 < d < 4 \); \( r(q, t) \sim \exp[-f_4(q)t/\ln t] \) in \( d = 4 \); and \( r(q, t) \sim \exp[-f_d(q)t] \) for \( d > 4 \). The results of our analysis are checked by Monte Carlo simulations.

The Voter model is a simple stochastic model which exhibits interesting dimension dependent properties [1]. Whilst, in one dimension, it is equivalent to the Glauber-Ising model at zero temperature, its properties depart from this model in higher dimensions. On each site of a \( d \)–dimensional lattice, opinions of a voter, or values of a spin \( \sigma = 1, 2, \ldots, q \), are initially distributed randomly. Between \( t \) and \( t + dt \) a site is picked at random. The voter on this site takes the opinion of one of its \( 2d \) neighbours, also chosen at random. This model is equivalent to a system of coalescing random walkers [2]. Therefore, due to the recurrence properties of random walks, the interactions between
walkers are strong in low dimensions \((d \leq 2)\), and progressively decrease in higher dimensions. As a consequence, for \(d \leq 2\) a consensus of opinion is approached as \(t \to \infty\), or in other words, the system coarsens. This contrasts with the behaviour in higher dimensions where disagreements persist indefinitely \([1]\).

In this letter we compute the persistence probability \(r(q, t)\), i.e. the probability that a given voter never changed his opinion up to time \(t\), in arbitrary dimension \(d \geq 2\), using field theoretical methods. Persistence probabilities have been investigated for a number of coarsening systems, including the Glauber-Ising chain \([2, 3, 4]\), exhibiting in all cases an algebraic decay with time. By contrast, it was found numerically that the persistence probability of the 2\(d\) Voter model with \(q = 2\) had the behaviour \(\exp[-\text{const.}(\ln t)^2]\) \([5]\).

This change in behaviour with dimension motivated the present work. Field theoretical methods are well adapted for the study of the role of dimension in simple physical systems, hence their use is natural in the present context.

The first step for the study of persistence consists in mapping the Voter model onto its dual particle system, i.e. a system of particles on a \(d\)-dimensional lattice, with the reaction \(A + A \rightarrow A\), fermionic occupation rules and particle injection at the origin. As soon as one particle leaves that site another particle is immediately injected there in its place. This reaction-diffusion model is hereafter referred to as model I. Defining \(Q(m, t)\) as the probability that, after a time \(t\) there are \(m\) particles in the system, one has \(r(q, t) = \sum_{m=1}^\infty Q(m, t) q^{1-m}\) \([3, 4]\). After a time \(t\) the injected particles will have penetrated a distance of order \(t^{1/2}\) from the source. At greater distances there are almost no particles present, whereas behind the front a steady state is reached. Hence in order to calculate the probability \(r(q, t)\), one can proceed as follows: (i) Compute

\[
s(q, L) = \sum_{m=1}^{N} P^I(m, L) q^{1-m},
\]

where \(P^I(m, L)\) is the steady state probability of finding \(m\) particles within a distance \(L\) of the point of injection (for \(t^{1/2} \gg L\)), and where \(N = \mathcal{O}(L^d)\) is the number of sites within that radius. (ii) Now compute the persistence probability using finite size scaling: \(r(q, t) \sim s(q, L \sim t^{1/2})\).

Unfortunately a computation of the \(P^I(m, L)\) is still too difficult. Therefore we replace model I with a slightly modified variant, which we will call
model II. We now have multiple (bosonic) site occupancy with a finite reaction rate $\lambda$, and particle input at the origin with a rate $J$. However the relevant properties of both models will turn out to be dominated by their (low density) behaviour far from the injection point. This means that, in the large $L$ limit, the distributions $P^I$ and $P^H$ actually become very similar, hence model II can be used for the persistence calculation. Evidence for this claim will be given later.

The mean field equation for model II, with particle density $\rho(x,t)$, is given by (for details see [6, 7])

$$\partial_t \rho = D \nabla^2 \rho - \lambda \rho^2 + J \delta^d(x),$$

(2)

where $D$ is the diffusion constant. Well behind the diffusive front, but far from the injection point, we can insert a steady-state power law ansatz into (2), giving

$$\rho(x) \sim \begin{cases} 
2(4 - d)D/(\lambda x^2), & (d < 4) \\
2D/(\lambda x^2 \ln x), & (d = 4) \\
\text{const.} \ x^{2-d}, & (d > 4).
\end{cases}$$

(3)

Note that for $d \leq 4$, the asymptotic form of $\rho(x)$ does not depend on $J$, an indication of the similarity between models I and II. The change in behaviour at $d = 4$ is due to the role of the reaction term in (2). For large $x$, this reaction term is marginal in $d = 4$, whereas for $d > 4$, it can be asymptotically ignored, the density instead decaying in accordance with Laplace’s equation. Hence $d = 4$ is, in a sense, a critical dimension. Nevertheless the behaviour of the average density around $d = 4$ can be described adequately within mean field theory. However in low spatial dimensions ($d \leq d_c = 2$), the mean field equation (2) is no longer correct, since it does not include fluctuation effects. We therefore map model II to a field theory which systematically includes these fluctuations, and permits an analysis of the probability distribution $P^H(m, L)$ in arbitrary dimension. Following the work of [8, 9], and using the formalism of [10], the master equation for model II may be rewritten as a second quantized Hamiltonian

$$\hat{H} = -D \sum_{n.n.} \{ \hat{a}_i^\dagger (\hat{a}_j - \hat{a}_i) \} - \lambda \sum_i \{(1 - \hat{a}_i^\dagger \hat{a}_i)\hat{a}_i^\dagger \hat{a}_i^2 \} - J\{\hat{a}_1^\dagger - 1\}. $$

(4)

The first term describes diffusion of the $A$ particles, the second term describes their coagulation reaction, and the final term results from stochastic
(Poissonian) particle injection at the origin (site 1). Defining \( P(n_1, \ldots, n_N) \) to be the probability that the system is in a configuration \( \{n_1, \ldots, n_N\} \) \( (n_i \) is the occupation number for site \( i \)\), then the steady state probability \( P^{II}(m, L) \) of finding \( m \) particles in the system is

\[
P^{II}(m, L) = \sum_{\{n\}} P(n_1, \ldots, n_N) \delta(\sum_j n_j, m). 
\]

(5)

In the second quantized formalism, where \( |\Psi\rangle \) is the system state ket, this becomes

\[
P^{II}(m, L) = \langle 0 | \exp(\sum_i \hat{a}_i) \delta(\sum_j n_j, m) |\Psi\rangle
\]

\[
= \frac{1}{m!} \sum_{n_1 \ldots n_N}^{m!} \frac{m!}{n_1! \ldots n_N!} \langle 0 | \hat{a}^{n_1}_1 \ldots \hat{a}^{n_N}_N |\Psi\rangle.
\]

Identifying this last expression as a multinomial expansion yields

\[
P^{II}(m, L) = \langle 0 | \frac{1}{m!} (\sum_i \hat{a}_i)^m |\Psi\rangle.
\]

(7)

Hence the generating function for the probabilities \( P^{II}(m, L) \) is given by

\[
G(y) = \sum_{m=0}^{\infty} y^m P^{II}(m, L) = \langle 0 | \exp(y \sum_i \hat{a}_i) |\Psi\rangle = \langle | \exp((y - 1) \sum_i \hat{a}_i) |\Psi\rangle,
\]

(8)

with the definition \( \langle | \equiv \langle 0 | \exp(\sum_i \hat{a}_i) \). Performing the mapping to a path integral \([8, 9, 10]\) yields the continuum action

\[
S = \int d^4x \int dt \left( \bar{a}(\partial_t - \nabla^2)a + \lambda \bar{a}a^2 + \lambda \bar{a}^2a^2 - J\bar{a}\delta^d(x) \right),
\]

(9)

where we have absorbed the diffusion constant \( D \) into a rescaling of time and of the couplings \( \lambda \) and \( J \). Furthermore the continuum limit of (8) is

\[
G(y) = \left\langle \exp \left[ (y - 1) \int d^d x \ a(x) \right] \right\rangle,
\]

(10)

with the average performed with respect to the action \( S \). From (8) we have

\[
P^{II}(m, L) = \left. \frac{1}{m!} \frac{d^m}{dy^m} G(y) \right|_{y=0}
\]

(11)
which leads to the continuum expression
\[ P^{\Pi}(m, L) = \frac{1}{m!} \left\langle \int d^d x_1 a(x_1) \ldots \int d^d x_m a(x_m) \exp \left[ - \int d^d x \ a(x) \right] \right\rangle, \] (12)
and hence, from (1),
\[ s(q, L) \sim G(q^{-1}) = \left\langle \exp \left[ - (1 - q^{-1}) \int d^d x \ a(x) \right] \right\rangle, \] (13)
where we have used the equivalence between models I and II in the large \( L \) limit, mentioned previously.

The reaction rate \( \lambda \) has naive scaling dimension \( k_2 - d \). This predicts a critical dimension \( d_c = 2 \) at or below which we must renormalize the theory. A very similar case has been studied in [11]. Hence the dimensionless renormalized coupling evaluated at the normalization point \( s = \kappa^2 \) is given by \( g_R = \kappa^{-\epsilon} \lambda_R \), where \( \epsilon = 2 - d \) and
\[ \lambda_R = \lambda(1 + \lambda I_d(\kappa))^{-1}, \quad I_d(\kappa) = 2^{d/2} \int \frac{d^d p}{(2\pi)^d} \frac{1}{s + p^2} \bigg|_{s=\kappa^2}. \] (14)
The beta function is then given by \( \beta(g_R) \equiv \kappa \partial_\kappa g_R = g_R^2 / 2\pi \) in \( d = 2 \). This is exact to all orders in \( g_R \). The field \( \langle a \rangle \) now satisfies the renormalization group equation
\[ \left( \frac{\kappa}{\kappa} \frac{\partial}{\partial \kappa} + \beta(g_R) \frac{\partial}{\partial g_R} \right) \langle a \rangle(x, g_R, \kappa) = 0. \] (15)
The solution of this equation is given by
\[ \langle a \rangle(x, g_R, \kappa) = (\kappa x)^{-d} \langle a \rangle(\kappa^{-1}, \tilde{g}_R, \kappa) \quad (d \leq 2), \] (16)
where the running coupling \( \tilde{g}_R \) acquires the universal asymptotic form
\[ \tilde{g}_R \sim \frac{2\pi}{\ln x} \left[ 1 + O((\ln x)^{-1}) \right] \quad (d = 2). \] (17)
The leading order (tree) diagrams for the density are exactly equivalent to the mean field rate equations [10]. Hence inserting the mean field expression (3) into the RG equation (16) yields the exact leading asymptotic expression
\[ \langle a(x) \rangle \sim \frac{2 \ln x}{\pi x^2} \left( 1 + O((\ln x)^{-1}) \right) \quad (d = 2). \] (18)
We note that this renormalization procedure is only needed at or below the
upper critical dimension. For \( d > 2 \) the mean field results in (3) become
qualitatively correct.

The next step is to turn the expression (13) for the persistence proba-
bility into a cumulant expansion (see [12] for a similar case),

\[
\langle \exp \left[ - (1 - q^{-1}) \int d^d x \ A(x) \right] \rangle \sim (19)
\]

\[
\exp \left[ -(1 - q^{-1}) \int d^d x \langle A(x) \rangle + \frac{1}{2} (1 - q^{-1})^2 \int d^d x \int d^d x' \langle A(x) A(x') \rangle_\text{c} - \cdots \right].
\]

We now consider the behaviour of \( s(q, L) \) in different dimensions.

\( \bullet d = 2 \). We use (13) to derive the leading behaviour of the first term in the
cumulant expansion. Hence for a system of radius \( L \),

\[
\int d^2 x \langle A(x) \rangle \sim \int d^2 x \, \frac{2 \ln x}{\pi x^2} \sim 2 (\ln L)^2,
\]

where the next order term is suppressed by \( (\ln L)^{-1} \). This result implies that
almost all of the particles in the system are situated very far from the point of
injection. This confirms that the behaviour of the model close to the injection
point is unimportant when considering the large \( L \) limit. Substituting (20)
into (13) gives

\[
s(q, L) \sim \exp \left[ - h_2(q) (\ln L)^2 + h_2^{(2)}(q) \ln L + \cdots \right]
\]

(21)

\[
h_2(q) = 2(1 - q^{-1}) + 4 C_2 (1 - q^{-1})^2 + \cdots,
\]  

(22)

where we assume that all the higher order cumulants also have leading order
contributions of \( \mathcal{O}(\ln L)^2 \). Simulations of the reaction-diffusion system have
confirmed this expectation (see below). Hence for the Voter model, using
finite size scaling \( L \sim t^{1/2} \) in (21), we have

\[
r(q, t) \sim \exp \left[ - f_2(q) (\ln t)^2 + f_2^{(2)}(q) \ln t + \cdots \right], \quad (d = 2)
\]

(23)

\[
f_2(q) = (1/2)(1 - q^{-1}) + C_2 (1 - q^{-1})^2 + \cdots.
\]  

(24)

This analysis shows that the unusual logarithmic form of this decay is due
to the fact that \( d = 2 \) is a critical dimension.
• $2 < d \leq 4$. Using the mean field expressions (3) yields the leading behaviour of the first term in the cumulant expansion

$$\int d^d x \langle a(x) \rangle \propto \begin{cases} L^{d-2}/\lambda, & (2 < d < 4) \\ L^2/\lambda \ln L, & (d = 4). \end{cases}$$

(25)

Hence almost all of the particles are again situated far from the injection site. Furthermore, dimensional analysis can be combined with an analysis of the diagrams for the higher order cumulants to show that each term in the cumulant expansion scales identically to (25). We assume that a logarithm is present in each term of the cumulant expansion for $d = 4$, although this analysis only explicitly shows the presence of a logarithm in the first term. Using finite size scaling, we then have

$$r(q, t) \sim \begin{cases} \exp[-f_d(q)t^{(d-2)/2} + \cdots], & (2 < d < 4) \\ \exp[-f_4(q)t/\ln t + \cdots], & (d = 4) \end{cases}$$

(26)

$$f_d(q) = B_d(1-q^{-1}) + C_d(1-q^{-1})^2 + \cdots.$$  

(27)

Note that the cumulant amplitudes $B_d, C_d, \ldots$ are no longer universal.

• $d > 4$. We again use (3) for the first term in the cumulant expansion,

$$\int d^d x \langle a(x) \rangle \propto L^2.$$  

(28)

For $d > 4$ the reaction process is asymptotically irrelevant since the particles no longer “see” each other. Since the (non-Poissonian) reaction process is not then present, the diagrams for the higher order cumulants actually vanish. Hence we have

$$r(q, t) \sim \exp[-f_d(q)t + \cdots] \quad (d > 4)$$

(29)

$$f_d(q) = B_d(1-q^{-1}).$$  

(30)

In summary, we find that, at large times and in any dimension, $r(q, t) \sim \exp[-f(q)\langle m(t) \rangle]$, where $\langle m(t) \rangle$ is the average number of particles in the dual reaction–diffusion models at time $t$ (see also [13]). Note that in $d = 1$, $\langle m(t) \rangle$ grows logarithmically in time, in accord with the algebraic fall off of the persistence probability.

Let us finally report on simulations. We measured the persistence probability $r(q, t)$ for the Voter model, as a function of time, with $q = 2, \ldots, \infty$.  

7
The results for $d = 2$ on a system of size $(1000)^2$ are shown in figure 1, where a decay $r(q, t) \sim \exp[-f_2(q)(\ln t)^2]$ is clearly seen, in agreement with both our predictions and the simulations for $q = 2$ in [3]. A plot of $f_2(q)$ extracted numerically is given in figure 2. It shows qualitative agreement with the prediction given in [24]. In particular for increasing values of $q$, $f_2(q)$ progressively departs from the first predicted term (dashed line on the figure).

Simulations performed on the 3d Voter model of size $(100)^3$ do not show the expected decay $r(q, t) \sim \exp[-f_3(q)t^{1/2}]$ at large times. Instead, the numerical results indicate an effective exponent for $t$ slightly larger than 1/2. This is due to large subleading corrections which make extraction of the leading behaviour difficult.

We now turn to simulations of the dual model I. In $d = 2$ we measured $\langle m \rangle = \int d^d x \langle a(x) \rangle$, the average number of particles in the stationary state, for system sizes $L = 10, \ldots, 80$, averaged over $2 \times 10^6$ MC steps (figure 3). The data are consistent with $\langle m \rangle \sim 2(\ln L)^2$ plus corrections of $O(\ln L)$, in agreement with [20]. This plot also provides evidence for the equivalence between models I and II, since our simulations are for model I whereas the theory was developed for model II.

We also examined the behaviour in $d = 2$ of the first three higher order cumulants for the $a(x)$ field (integrated over the system size as in [20]). They all scale with $L$ in the same way as the first cumulant $\int d^d x \langle a(x) \rangle = \langle m \rangle$. This confirms that, in principle, each term in the cumulant expansion is equally “relevant” for the persistence calculation. Nevertheless the numerical prefactors for each term appear to decrease fairly rapidly indicating that the first few terms are the most significant.

Finally, in $d = 3$ we measured $\langle m \rangle$ as a function of $L$, for $L = 10, \ldots, 40$. For large sizes, instead of the prediction $\langle m \rangle \sim L$ (see [23]), the data show an apparent exponent of about 1.2, indicating again the presence of large subleading corrections.

References

[1] Liggett T M 1985 Interacting Particle Systems (Springer)
[2] Derrida B, Bray A and Godrèche C 1994 *J. Phys. A: Math. Gen.* **27** L357

[3] Derrida B 1995 *J. Phys. A: Math. Gen.* **28** 1481

[4] Derrida B, Hakim V, Pasquier V 1995 *Phys. Rev. Lett.* **75** 751

[5] Ben-Naim E, Frachebourg L and Krapivsky P L 1996 *Phys. Rev. E* **53** 3078

[6] Cheng Z, Redner S and Leyvraz F 1989 *Phys. Rev. Lett.* **62** 2321

[7] Krapivsky P L 1993 *Physica A* **198** 157

[8] Doi M 1976 *J. Phys. A: Math. Gen.* **9** 1465, 1479

[9] Peliti L 1985 *J. Physique* **46** 1469

[10] Lee B P 1994 *J. Phys. A: Math. Gen.* **27** 2633

[11] Howard M and Cardy J 1995 *J. Phys. A: Math. Gen.* **28** 3599

[12] Cardy J 1995 *J. Phys. A: Math. Gen.* **28** L19

[13] Dornic I and Godrèche C, in preparation
Figure 1: \( \ln r(q, t) \) plotted against \((\ln t)^2\) for \(q = 2, 3, 4, 5, 6, 7, 8, 10, 14, 20, 40, 10^6\) (from top to bottom), in the 2d Voter model (system size \((1000)^2\)).
Figure 2: Plot of the numerical data for $f_2(q)$, extracted from figure 1, versus $(1-q^{-1})$ (dots). Also illustrated is the lowest order theoretical approximation to $f_2(q)$ (dashed line).
Figure 3: Plot of $\langle m \rangle$ versus $\ln L$ for system sizes $L = 10, \ldots, 80$, averaged over $2 \times 10^6$ MC steps (for model I). Solid line is a quadratic best fit to the data, where the coefficient of the quadratic term is found to be close to 1.9.