

Majority-vote model in one-dimensional on directed small-world networks

Abstract

We investigate the critical properties of the Majority-Vote model (MVM) one-dimensional (1D) on directed small-world networks. The MVM is studied by applying the Monte Carlo method. We calculate the critical points, as well as the critical exponent's ratio, β/ν . We find that MVM presents identical exponents to the Ising model one-dimensional on directed Small-World networks (DSW). Our results are in agreement with the Grinstein criterion for models with up and down symmetry on regular lattices

Keywords: Majority-vote, networks, monte Carlo simulations

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Introduction

Grinstein et al.¹ argue that non-equilibrium spin systems on square lattices (SL) with up-down symmetry belongs to the same class of universality of the Lenz-Ising model in two dimensions (2D). This hypothesis was endorsed for some non-equilibrium models on other regular lattices.²⁻⁸

In 1992 Oliveira⁹ proposed the a non-equilibrium model known as MVM which disobeys the detailed balance. The update of the MVM follows a Markov sequence of stochastic dynamics with local rules and with up-down symmetry. In 2D, on a square lattice, the MVM presents a continuous phase transition with critical exponents identical⁹ of the Ising model¹⁰. Sousa¹¹ and Brenda¹² studied the Ising model and MVM on DSW random lattices, respectively. The exponents obtained in both models are identical and in agreement with the conjecture suggested by Grinstein et al.¹

In this paper, we consider the MVM in 1D on DSW networks and perform an extensive computer simulation study of the MVM. To extract the critical exponents, we applied finite-size scaling (FSS) techniques. Monte Carlo simulations of this system were performed using a master equation to update the spins. Here, there is a continuous phase transition for $0 < p \leq 1$, where p is the rewiring probability. Besides, the calculated critical exponents for $p=0.1$ do not belong to the same universality class as the 2D Ising model.

We consider the non-equilibrium MVM on DSW networks by a set of spins variables σ_i assuming values values ± 1 located on every node i of a DSW networks with L sites, where L is the length of a linear chain. The small-world network in one-dimension is built from a regular network with two closest neighbors, connected to L nodes and J neighbors. In this network, each node is randomly reconnected with n edges with probability p . When $p = 0$ for the network it is regular (received no long-range connection), but for $0 < p < 1$ the network is small world (short-range links) and $p = 1$ random network (long-range connections), as shown in Figure 1.

In the MVM on a network, the system dynamics traditionally is as follows: We assign a spin variable σ_i with values ± 1 at each node of the net. At each step, we try to spin- flip a node. The flip is accepted with probability

$$w_i = \frac{1}{2} \left(1 - (1-2q) \sigma_i \cdot S \left(\sum_j \sigma_j \right) \right) \quad (1)$$

where $S(x)$ is the sign of x if $x \neq 0$, $S(x)=0$ if $x=0$. To calculate w_i our sum runs over the ($J=2$) nearest neighbours of spin i on the network. In this model, we add a long-range connection connecting to another site k with $p=0.1$. This connection is only one way, that is, the site k does not send back a connection to the site i . w_i means that with probability $(1-q)$ the spin will adopt the same state as the majority of its neighbours. The control noise parameter $q (0 \leq q \leq 1)$ works like the temperature in the Ising model: the smaller the value of q , the greater the likelihood of parallel alignment with the local majority. The simulations have been performed on different DSW networks sizes comprising a number $L=5000, 10000, 20000, 40000, 60000, 80000, 120000, 160000, 200000$, and 2600000 of sites. For each L size quenched averages over the connectivity disorder are approximated by averaging over independent realizations. For each simulation, we have started with a stable configuration of spins. We ran 3×10^5 Monte Carlo steps (MCS) per spin with 2×10^5 configurations discarded for thermalization using a random-number generator.

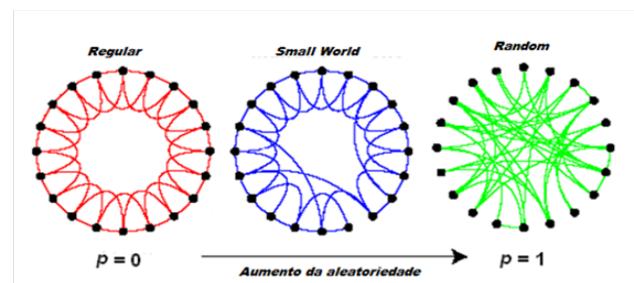


Figure 1 Random networks.

Results and discussions

The molar magnetization, $m = \sum_i \sigma_i / L$, were measured. From magnetization, we can obtain other measures such as the average magnetization, susceptibility and the fourth-order Binder cumulant,

$$m = [\langle m \rangle]_{av}, \quad (2)$$

$$\chi(q) = \frac{N}{T} \left[\langle m^2 \rangle_{av} - \langle m \rangle_{av}^2 \right]_{av} \quad (3)$$

$$U_4(q) = 1 - \frac{\langle m^4 \rangle_{av}}{3[\langle m \rangle]_{av}^2}, \quad (4)$$

in the above equations $\langle \dots \rangle$ stands for thermodynamic averages and $[\dots]_{av}$ for averages over different realizations.

In order to calculate the exponents of these models, we apply finite-size scaling (FSS) theory. We then expect, for large system sizes, an asymptotic FSS behavior of the form

$$m = L^{-\beta/v} f_m(x) [1 + \dots], \quad (5)$$

$$\chi = L^{\gamma/v} f_\chi(x) [1 + \dots], \quad (6)$$

Where β and γ are the usual critical exponents, and $f_i(x)$ are FSS functions with

$$x = (q - q_c)^{1/v} \quad (7)$$

being the scaling variable. The dots in the brackets $[1 + \dots]$ indicate corrections-to-scaling terms. We calculated the error bars from the fluctuations among the different realizations. Therefore, from the size dependence of m and χ we obtain the exponents ratios β/v and γ/v respectively. The susceptibility at its maximum also scales as $L^{\gamma/v}$. Moreover, the value of $T^0 = T_c(L)$ for which χ has a maximum scales with the lattice size as

$$T_c(L) = T_c + bL^{-1/v}$$

In this way, Eq. 7 may be used to get $1/v$.⁷

In the Figure 2, we plot the magnetization, Binder Cumulante, and susceptibility versus the noise parameter q for sizes $L=5M, 10M, 20M, 30M, 40M, 60M, 80M, 120M, 160M, 200M$, and $260M$ and rewiring probability $p=0.1$. The shape of these figures indicates that this model exhibits a continuous phase transition.

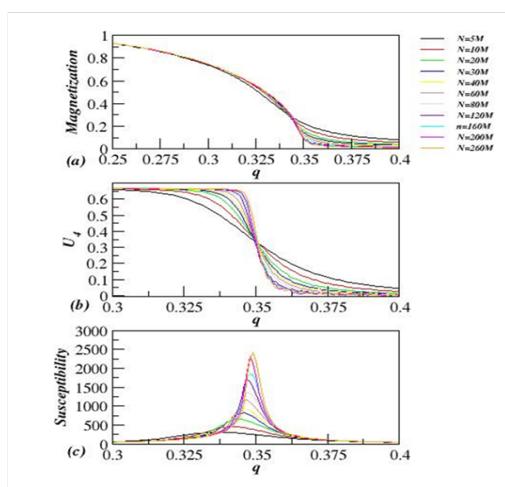
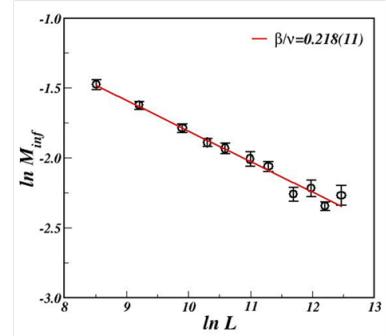


Figure 2 Plot of the magnetization (A) Binder Cumulant (U_4) (B) and susceptibility χ (C) as a function of the noise parameter q and for sizes $L=5M, 10M, 20M, 30M, 40M, 60M, 80M, 120M, 160M, 200M$, and $260M$ with rewiring probability $p=0.1$. Here $1M=1000$.

In the Figure 3, we plot logarithm of the magnetization at q_c versus $\ln L$ for $p=0.01$ and of the eq. (5), we obtain the exponents ratio $\beta/v = 0.218(11)$.

Figure 3 Plot of the logarithm of the magnetization at q_c as a function of the



logarithm of L noise parameter $p=0.1$.

In the Figure 4, we plot logarithm of the susceptibility χ at q_c and χ_{\max} versus $\ln L$. Of the eq.(6), we obtain the exponents ratio $\gamma/v_{q_c} = 0.535(7)$, and $\gamma/v_{\chi_{\max}} = 0.533(8)$ for $p = 0.1$.

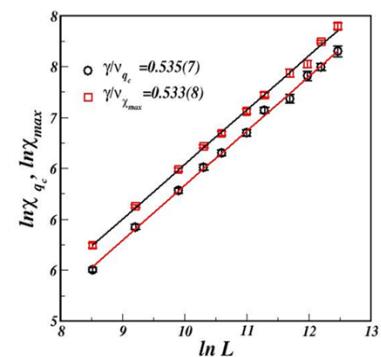


Figure 4 Plot of the logarithm of the susceptibility χ_{q_c} and χ_{\max} versus $\ln L$ and noise parameter $p=0.1$.

In the Figure 5, we plot the log-log of $[q_c(L) - q_c]$ versus L and the eq. (7), we obtain the exponents ratio $1/v = 0.51(3)$.

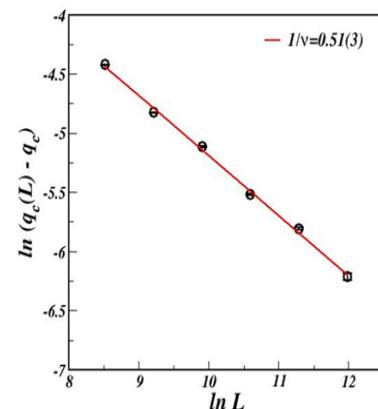


Figure 5 Plot of the $\ln[q_c(L) - q_c]$ versus $\ln L$ and noise parameter $p=0.1$.

Conclusion

In the present work, we have shown that, by considering the ferromagnetic MVM in one-dimension on DSW networks there is a continuous phase transition. The exponents ratio $\beta / \nu = 0.218(11)$, $\gamma / \nu_{qc} = 0.535(7)$ and $1 / \nu = 0.51(3)$ for $p=0.1$ indicate that they are identical from Ising model in one-dimension on DSW networks.¹³ Therefore, our results agree with the Grinstein criterion for DSW networks.

Acknowledgments

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Conflict of interest

The author states that there is no conflict of interest.

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