

Impact of contrarians and intransigents in a kinetic model of opinion dynamicsNuno Crokidakis,^{1,*} Victor H. Blanco,¹ and Celia Anteneodo^{1,2}¹*Departamento de Física, PUC-Rio, Rio de Janeiro, Brazil*²*National Institute of Science and Technology for Complex Systems, Rio de Janeiro, Brazil*

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In this work we study opinion formation on a fully connected population participating of a public debate with two distinct choices, where the agents may adopt three different attitudes (favorable to either one choice or to the other, or undecided). The interactions between agents occur by pairs and are competitive, with couplings that are either negative with probability p or positive with probability $1 - p$. This bimodal probability distribution of couplings produces a behavior similar to the one resulting from the introduction of Galam's contrarians in the population. In addition, we consider that a fraction d of the individuals are intransigent, that is, reluctant to change their opinions. The consequences of the presence of contrarians and intransigents are studied by means of computer simulations. Our results suggest that the presence of inflexible agents affects the critical behavior of the system, causing either the shift of the critical point or the suppression of the ordering phase transition, depending on the groups of opinions to which the intransigents belong. We also discuss the relevance of the model for real social systems.

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I. INTRODUCTION

Models of opinion formation have been studied by physicists since the 1980s and are now part of the new branch of physics called sociophysics [1]. This recent research area uses tools and concepts of statistical physics to describe some aspects of social and political behavior [2]. From the theoretical point of view, opinion models are interesting to physicists because they present order-disorder transitions, scaling and universality, among other typical features of physical systems [2].

Following the success of the Ising model to capture the essential physics of complex systems, several opinion models have been proposed based on ± 1 (i.e., spin-1/2) state variables [2]. The first paper that considered the Ising model to describe a social system was proposed by Galam [3]. The spin-spin coupling of the Ising Hamiltonian represents the agent-agent interaction, whereas the magnetic field represents the effects of propaganda. Moreover, local (or individual) fields are introduced that reflect agent preference toward each orientation (or opinion). Depending on the strength of the local fields, the system may reach full consensus toward one of the two possible opinions $+1$ or -1 , or a state in which both opinions coexist. In the last 30 years many other opinion models based on Ising variables have been proposed [1,2]. Among them, we highlight the voter model [4,5], the majority-rule models [6–8], the Sznajd model [9], and the Continuous Opinion and Discrete Actions (CODA) model [10]. Besides the affinity by either one of two distinct opinions or attitudes, one can also consider the possibility that individuals may remain undecided [11,12]. This more realistic situation, that we will consider here can be associated to spin-1 systems, in which the state variables can assume also a null value besides ± 1 .

To make the models even more realistic, other psychosocial ingredients can be taken into account. The so-called *contrarians* are agents who always have the opposite opinion to that of

the majority of the surrounding agents [6]. The consideration of such agents affects opinion dynamics, and their impact on opinion formation has been studied in a series of models [11–17]. Another category of agents are the *intransigents*, whose stubbornness or inflexibility makes them reluctant to change their opinions. This class of agents was first introduced in Ref. [18] and they received later the name inflexible agents or just inflexibles in Ref. [19]. After these works, many other papers considered the effect of inflexibles in opinion dynamics [20–24].

In this work we study a three-state kinetic model of opinion formation, in which the dynamics evolves according to pairwise competitive interactions and where both contrarian and inflexible features are considered. Our results suggest that the presence of inflexible agents affects the critical behavior of the system, causing either the shift of the critical point or the suppression of the phase transition, depending on the opinion group to which the intransigents belong.

This work is organized as follows. In Sec. II we present the microscopic rules that define the model. The numerical results are discussed in Sec. III, and our conclusions are presented in Sec. IV.

II. MODEL

Our model is based on kinetic exchange opinion models [11,25]. A population of N agents is defined on a fully connected graph, i.e., each agent can interact with all others, which characterizes a mean-field-like scheme. Each individual i ($i = 1, 2, \dots, N$) carries one of three possible opinions or attitudes at a given time step t , represented by $o_i(t) = +1, -1$, or 0 . This scenario mimics any polarized public debate, for example, an electoral process with two different candidates A and B , where each agent (or elector) votes for the candidate A (opinion $+1$), for the candidate B (opinion -1), or remains undecided (opinion 0). In addition, there is a fraction d of agents that are averse to change their opinions, the so-called inflexible agents.

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Each interaction occurs between two given agents i and j , such that j will influence i . The following rules govern the dynamics.

- (1) A pair of agents (i, j) is randomly chosen.
- (2) If i is an inflexible agent nothing occurs, because he or she cannot be persuaded to change their opinion.
- (3) On the other hand, if i is not an inflexible agent, his or her opinion in the next time step $t + 1$ will be updated according to

$$o_i(t + 1) = \text{sgn}[o_i(t) + \mu_{ij} o_j(t)], \quad (1)$$

where the sign function is defined such that $\text{sgn}(0) = 0$ and the couplings $\{\mu_{ij}\}$ are given by the discrete bimodal probability distribution

$$F(\mu_{ij}) = p \delta(\mu_{ij} + 1) + (1 - p) \delta(\mu_{ij} - 1). \quad (2)$$

Notice that the above rules impose that for an agent to shift from state $o_i = +1$ to $o_i = -1$ or vice versa it must pass by the intermediate state $o_i = 0$. The above process is repeated N times, which defines one time step in the simulations. The pairwise couplings may be either negative (with probability p) or positive (with probability $1 - p$), such that p represents the fraction of negative couplings [11]. In other words, a disorder is introduced in the system, and we will consider that the stochastic random variables μ_{ij} can be either quenched (fixed in time) or annealed (changing with time), as in Res. [11,12]. The influence of one individual over another does not need to be reciprocal (i.e., not necessarily $\mu_{ij} = \mu_{ji}$), however, whether interactions are symmetric or not does not affect the results. The intransigents (a fraction d of the population) are randomly selected at the beginning of the simulation, maintaining that character throughout the dynamics, as considered in the Galam model [19].

In the absence of intransigents, there is a nonequilibrium order-disorder phase transition at a critical fraction $p_c = 1/4$ [11]. For $p < p_c$ one of the extreme opinions $+1$ or -1 dominates the system, with consensus states occurring only for $p = 0$, i.e., in the absence of negative interactions. On the other hand, for $p \geq p_c$ the system is in a disordered “paramagnetic” phase characterized by the coexistence of the three opinions, with the fraction of each opinion being $1/3$. Furthermore, it has already been argued [11,12] that negative couplings produce a similar effect to that of the introduction of the Galams’ contrarians [6] since the main consequence of such negative couplings is such that interacting agents with the same opinions move to the undecided state (opinion 0). In this sense, our model contains both contrarian and inflexible features.

In the simulations, we have considered two kinds of random couplings $\{\mu_{ij}\}$, quenched and annealed, as well as two kinds of updating schemes, synchronous (or parallel) and asynchronous (or sequential) updates. The systems were prepared in fully disordered initial states, i.e., we started all simulations with an equal fraction of each opinion ($1/3$ for each one). In the next section we will present our results.

III. RESULTS

We analyze the critical behavior of the system, in analogy to magnetic spin systems, by computing the order parameter

$$O = \left\langle \frac{1}{N} \left| \sum_{i=1}^N o_i \right| \right\rangle, \quad (3)$$

where $\langle \dots \rangle$ denotes a disorder or configurational average. It is sensitive to the unbalance between extreme opinions. Notice that O plays the role of the “magnetization per spin” in magnetic systems. In addition, we also consider the fluctuations χ of the order parameter (or “susceptibility”)

$$\chi = N (\langle O^2 \rangle - \langle O \rangle^2) \quad (4)$$

and the Binder cumulant U , defined as [26]

$$U = 1 - \frac{\langle O^4 \rangle}{3 \langle O^2 \rangle^2}. \quad (5)$$

We analyzed three distinct cases, according to whether the inflexible agents are (i) chosen independently of their initial opinions; (ii) chosen only among the agents with extreme (± 1) opinions; or (iii) restricted to a given group of opinion ($o = +1$ or -1 or 0). In the following subsections, we will present each case separately.

A. Uniformly distributed inflexible agents

In this case, the fraction d of inflexible agents is randomly selected at the beginning of the simulation, independently of their opinions. In Fig. 1 we exhibit the results for the order parameter, Eq. (3), versus the fraction p of negative couplings for typical values of d . We display, as examples, the outcomes for annealed variables $\{\mu_{ij}\}$ with synchronous updates [Fig. 1(a)], and quenched variables $\{\mu_{ij}\}$ with asynchronous updates [Fig. 1(b)], for a population of $N = 1000$ agents. Note the strong impact of the change of the parameter d on the behavior of the order parameter O . Furthermore, given a fixed value of d , the curve of O is not affected by the nature of the random variables μ_{ij} nor by the kind of update scheme used. Consensus states are obtained only in a very specific case: In the absence of both intransigents ($d = 0$) and negative interactions ($p = 0$). In other words, the maximal value of the order parameter O is smaller than one for all values of $d > 0$, independently of p . In real systems, full consensus, with $O = 1$, occurs in particular situations where a government exerts a social control through propaganda or policies that lead to a full acceptance of the *status quo*, while collective states with $O < 1$ represent more “democratic” frequently observed situations [20,27]. Thus, in this sense the inclusion of inflexible agents makes the model more realistic.

For sufficiently large d the system is always found in a disordered (paramagnetic) phase, but for small values of d the system orders at specific points that depend on d . To locate the critical points $p_c(d)$ numerically, we have performed simulations for different population sizes N . Thus, the transition points $p_c(d)$ are estimated, for each value of d , from the crossing of the Binder cumulant curves for the different sizes [26]. In addition, a finite-size scaling (FSS) analysis was performed to obtain an estimate of the critical

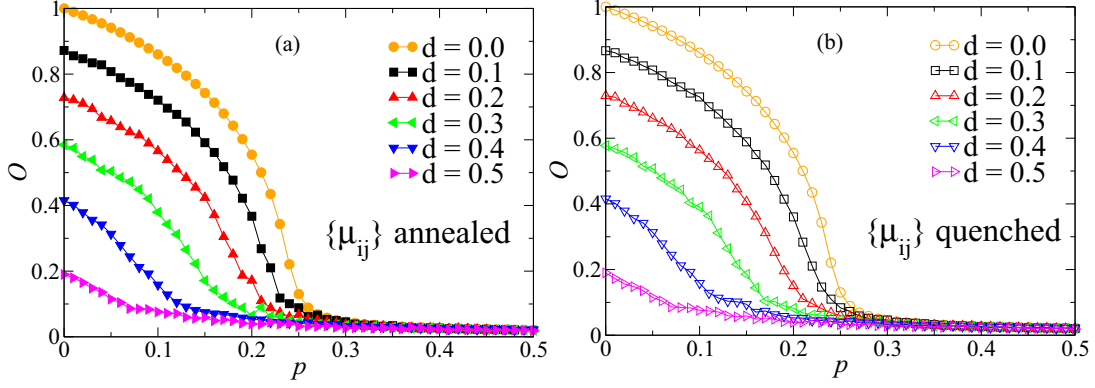


FIG. 1. (Color online) Order parameter O versus p for typical values of the density d for the case where the inflexibility is independent of the agent opinion. For comparison, we also exhibit the result for the model in the absence of inflexible agents [11] ($d = 0.0$). The pairwise interactions $\{\mu_{ij}\}$ and the update scheme used are annealed, (a) synchronous and quenched and (b) asynchronous, respectively. The population size is $N = 1000$ and data are averaged over 100 simulations.

exponents β , γ , and ν , by means of the usual FSS equations

$$O(d, N) \sim N^{-\beta/\nu}, \quad (6)$$

$$\chi(d, N) \sim N^{\gamma/\nu}, \quad (7)$$

$$U(d, N) \sim \text{constant}, \quad (8)$$

$$p_c(d, N) - p_c(d) \sim N^{-1/\nu}, \quad (9)$$

which are valid in the vicinity of the transition.

As an illustration, we exhibit in Fig. 2 the behavior of the quantities of interest as well as the scaling plots for $d = 0.2$ quenched random couplings and asynchronous updates. Our estimates for the critical exponents coincide with those for the original model ($d = 0$), i.e., we obtained $\beta \approx 0.5$, $\gamma \approx 1.0$, and $1/\nu \approx 0.5$. These exponents are robust: They are the same, within error bars estimated from the FSS analysis, for all values of d , independently of the update scheme considered and of the kind of random couplings $\{\mu_{ij}\}$ (quenched or annealed).

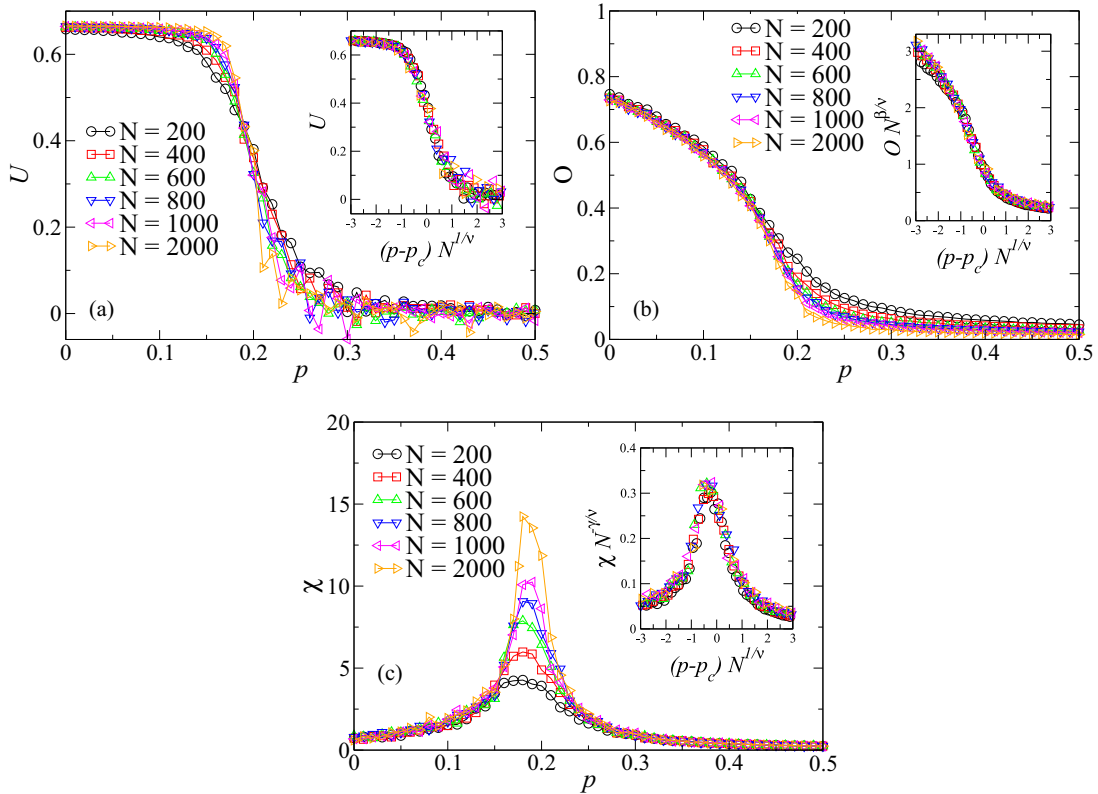


FIG. 2. (Color online) (a) Binder cumulant, (b) order parameter, and (c) susceptibility for the case where the inflexibility is independent of the agent opinion for $d = 0.2$ and different population sizes N (main plots). The corresponding scaling plots are shown in the respective insets. Data are for quenched random variables $\{\mu_{ij}\}$ and asynchronous update scheme. The best data collapse was obtained for $p_c = 0.196$, $\beta = 0.5$, $\gamma = 1.0$, and $1/\nu = 0.5$.

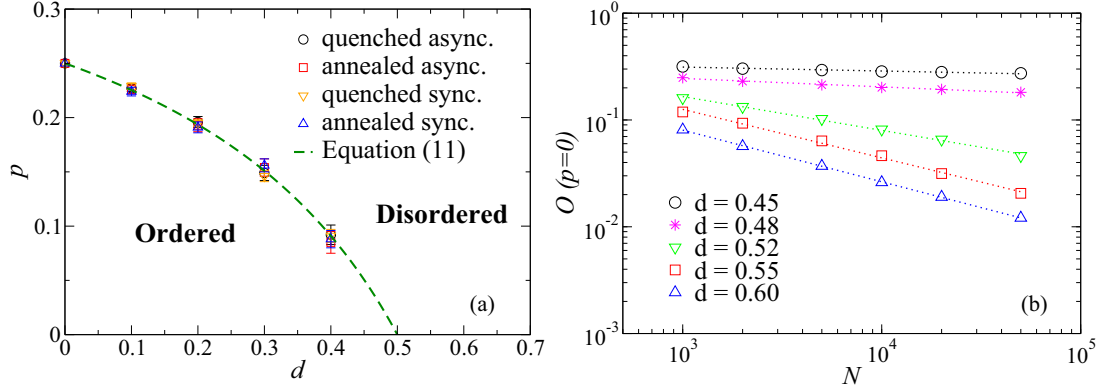


FIG. 3. (Color online) (a) Phase diagram of the model in the plane p versus d for the case where the inflexibility is independent of the agent opinion. We exhibit results for quenched and annealed couplings $\{\mu_{ij}\}$ and synchronous and asynchronous updates. The dashed line is a qualitative description of the phase boundary given by Eq. (11), as discussed in the text. (b) Maximal values of the order parameter as a function of the population size N , in the log-log scale, for values of d near the critical density $d_c \approx 0.5$ are shown.

Taking into account the FSS analysis for typical values of d , we exhibit in Fig. 3(a) the phase diagram of the model in the plane p versus d . The symbols are the numerical estimates for the critical points $p_c(d)$. Based on the analytical results of the annealed version of a similar model [12], where the critical points are given by a ratio of two first-order polynomials, we propose the following qualitative form for the critical frontier:

$$p_c(d) = \frac{x d + 1}{y d + 4}, \quad (10)$$

where x and y are real numbers, and we have taken into account the analytical result of the model in the absence of inflexible agents $p_c(d=0) = 1/4$ [11]. Fitting the numerical values of $p_c(d)$ with Eq. (10), we obtained

$$p_c(d) = \frac{2d - 1}{4.5d - 4}. \quad (11)$$

Equation (11) is plotted in Fig. 3(a) together with the numerical results. One can see that the curve describes qualitatively well the phase boundary between the ordered and the disordered phases, and the simulation data agree within error bars with Eq. (11). Based on Eq. (11) one can estimate the critical density d_c above which the system cannot order. This critical value is $d_c \approx 0.5$, and above it the three opinions $+1$, -1 , and 0 coexist in the population ($1/3$ on average for each one), which is a characteristic of the disordered phase of this kind of model [11,12]. To test the validity of Eq. (11) and the estimated value of d_c , we simulated the system for $p = 0$ and different population sizes N and we measured the order parameter $O(p=0)$. One can see in Fig. 3(b) that for $d < 0.5$ the values of $O(p=0)$ remain almost constant for increasing sizes. Nonetheless, for $d > 0.5$ the values of $O(p=0)$ decrease as a power law of N , which indicates that we have $O(p=0) = 0$ in the thermodynamic limit. Thus, these results suggest that the system will be in a disordered paramagnetic phase for $d > 0.5$. As discussed above, the critical exponents are the same for all values of d , indicating a universality on the order-disorder frontier. Thus,

for sufficiently large values of the fraction of inflexible agents the order-disorder transition is eliminated.

B. Inflexible agents restricted to the extreme opinions

We consider a variant of the model considered in the previous section. Instead of selecting as intransigent agents a fraction d of the population totally at random, one can restrict the inflexibility to agents that initially have one of the extreme opinions ± 1 . In other words, with probability d an agent is set as inflexible only if this agent has an initial opinion either $o = +1$ or $o = -1$. This is also a realistic case since in some countries there are intransigents supporting two extreme opinions, while the remaining individuals are undecided or intend to nullify their votes. In this case, these “neutral” agents can be persuaded by the decided individuals and adopt one of the extreme opinions (e.g., left or right candidate).

As in the previous case, we first studied the behavior of the order parameter O as a function of the fraction p of negative interactions for typical values of the density d . In Fig. 4 we exhibit a representative example for the case where the random couplings $\{\mu_{ij}\}$ are quenched variables and the states are updated in an asynchronous way. One can see that the qualitative behavior is similar to the one presented in Sec. IIIA, i.e., we observe order-disorder transitions at different values p_c that depend on d . However, there are qualitative differences and a comparison to Fig. 1(b) shows that the increase of d affects the order parameter less in the case where the intransigents are restricted to agents with the extreme opinions ± 1 . As a consequence, the values of $p_c(d)$ are different from those observed in the previous subsection. This fact can be easily understood. The agents with extreme opinions (be they inflexible or not) can provoke a change of the 0 opinions to $+1$ or -1 , which does not happen if the undecided agents are allowed to be intransigents. This fact favors the ordering in the system, and so the magnetization per spin in the actual case presents greater values than in the case where the inflexibility is not restricted and can be a characteristic of the individuals independently of their opinions. Once again, the nature of $\{\mu_{ij}\}$ and the update scheme used do not affect the results.

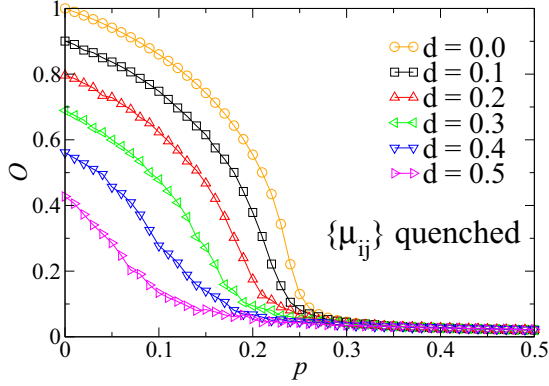


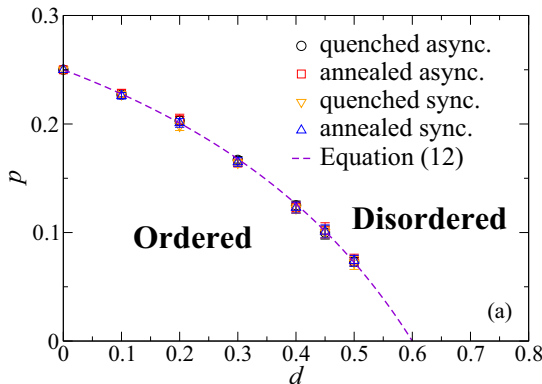
FIG. 4. (Color online) Order parameter O versus p for typical values of d for the case where the inflexibility is associated with agents carrying the extreme opinions ($o = \pm 1$). For comparison, we also exhibit the result for the model in the absence of inflexible agents [11] ($d = 0.0$). The pairwise interactions $\{\mu_{ij}\}$ and the update scheme used are quenched and asynchronous, respectively. The population size is $N = 1000$ and data are averaged over 100 simulations.

As in the previous section, we performed an FSS analysis to obtain the critical points $p_c(d)$ and the critical exponents β , γ , and ν . In Fig. 5(a) we exhibit the phase diagram of the model in the plane p versus d . The symbols are the numerical estimates for the critical points $p_c(d)$. Again, we propose the qualitative form of Eq. (10) for the order-disorder frontier. Fitting the data, we obtained

$$p_c(d) = \frac{1.67d - 1}{3.4d - 4}. \quad (12)$$

In other words, we have a similar frontier as in the previous case, but with different parameters. Equation (12) is plotted in Fig. 5(a) together with the numerical results. One can see that the curve describes qualitatively well the phase boundary between the ordered and the disordered phases, and the simulation data agree within error bars with Eq. (12).

Based on Eq. (12) one can estimate the critical density d_c above which the system cannot order. This critical value is



$d_c \approx 0.6$. Thus, for sufficiently large values of the fraction of inflexible agents, the order-disorder transition is eliminated. Notice that the critical density in this case ($d_c \approx 0.6$) is greater than the critical density of the previous case ($d_c \approx 0.5$), where the intransigent agents may be chosen independently of their opinions. The origin of this difference is again related to the agents with $o = 0$ opinions, as discussed in the beginning of this section. In fact, as the presence of noninflexible agents with $o = 0$ opinions favors the ordering in the system, as discussed above, the critical fraction p_c becomes larger in the present case than in the case where the inflexibility is not restricted and can be a characteristic of a given individual independently of his opinion. As in the previous section, we performed simulations for the system with $p = 0.0$ and different population sizes N to test the validity of the estimated value of d_c . One can see in Fig. 5(b) that for $d < 0.6$ the values of $O(p = 0)$ remain almost constant for increasing sizes. Nonetheless, for $d > 0.6$ the values of $O(p = 0)$ decreases as a power law of N , which indicates that we have $O(p = 0) = 0$ in the thermodynamic limit. Thus, these results reinforce the idea that the system will be in a disordered paramagnetic phase for $d > 0.6$. It is important to say that we obtained the usual exponents $\beta \approx 0.5$, $\gamma \approx 1.0$, and $1/\nu \approx 0.5$ for all values of d considered in Fig. 5(a), which indicates that the universality class of the model is not affected when we consider inflexible agents only among the individuals with the extreme opinions $o = \pm 1$.

C. Inflexible agents restricted to a given opinion

Finally, we also study another variant of the model considered in Sec. IIIA. Instead of selecting the dN inflexible agents at random, one can restrict the selection to a given opinion group. In other words, with probability d an agent becomes inflexible, but now only if this agent has initial opinion either $o = +1$ or $o = -1$ or $o = 0$. This can also be a realistic situation since in some countries there is only a certain point of view (or opinion) that is shared by an intransigent group.

We can first consider the case where the intransigents are chosen among the agents with opinion $o = 0$. We exhibit in

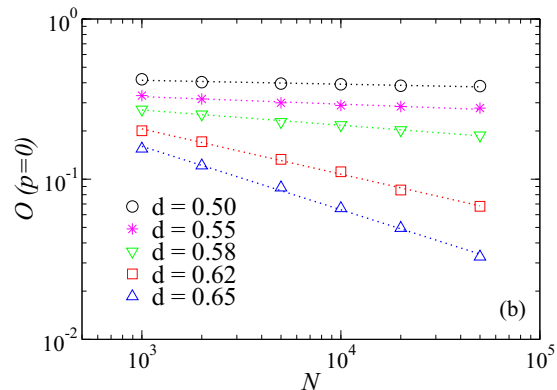


FIG. 5. (Color online) (a) Phase diagram of the model in the plane p versus d for the case where the inflexibility is associated with the extreme opinions ($o = \pm 1$). We exhibit results for quenched and annealed couplings $\{\mu_{ij}\}$ and synchronous and asynchronous updates. The dashed line is a qualitative description of the phase boundary given by Eq. (12), as discussed in the text. (b) Maximal values of the order parameter as a function of the population size N , in the log-log scale, for values of d near the critical density $d_c \approx 0.6$.

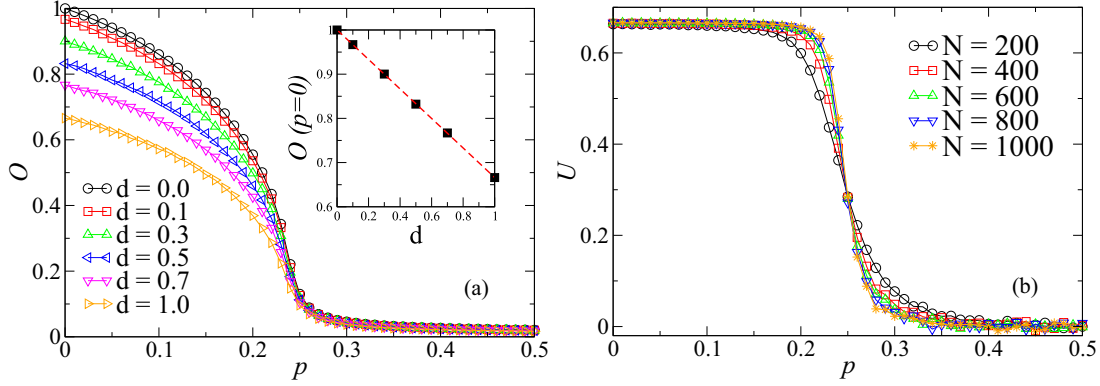


FIG. 6. (Color online) Results for the case where the intransigents are chosen among the agents with opinion $o = 0$. (a) Order parameter as a function of p for typical values of d for population size $N = 1000$. The inset shows the maximal value of the order parameter for a given value of d that occurs for $p = 0$. Fitting the data, we obtained $O(p = 0) = 1 - d/3$. (b) Binder cumulant for $d = 1.0$ and different sizes N , showing a crossing at $p_c \approx 0.25$. In both cases the interactions $\{\mu_{ij}\}$ are quenched random variables and we have considered asynchronous updates.

Fig. 6(a) the results for the order parameter as a function of p for typical values of d . One can see that the maximal value of the order parameter (for $p = 0$) decreases for increasing values of d . This result is expected since the initial condition is fully disordered (1/3 of each opinion) and a fraction d of the agents with opinion $o = 0$ are selected as intransigents at $t = 0$. In this case, the maximum of the order parameter should be $O_{\max} = O(p = 0) = 1 - d/3$, which is confirmed by the simulations [see the inset of Fig. 6(a)]. Although the values of the order parameter for $p < p_c$ are different for distinct values of d , the order-disorder transition occurs at the same point. An example is given in Fig. 6(b), where we exhibit the Binder cumulant as a function of p for the maximum of the density of intransigents $d = 1.0$. One can observe a crossing of the curves at $p_c \approx 0.25$. We also performed an FSS analysis (not shown), which confirms the same exponents observed in the previous sections, i.e., we have $\beta \approx 0.5$, $\gamma \approx 1.0$, and $1/\nu \approx 0.5$. These results are independent of d , which indicates that the universality class of the model is not affected when we consider inflexible agents only among the individuals with opinion $o = 0$.

In the case where the intransigents are restricted to agents with opinion $o = +1$,¹ the results are different from the previous case. We have observed that the order parameter decays with increasing values of p , as usual, but the lower values of O are not so small as usual [see Fig. 7(a)].

In addition, the order parameter curves, as well as the susceptibility ones, do not depend on the system size (see Fig. 7), as usually occurs in phase transitions [28,29]. These results suggest that there is no phase transition when we consider inflexible agents only among agents with one of the extreme opinions $o = +1$ or $o = -1$. To confirm this picture, we plot in Fig. 8 the Binder cumulant for two different values of d , namely $d = 0.3$ [Fig. 8(a)] and $d = 0.6$ [Fig. 8(b)], and different sizes N . We can observe that in both cases the Binder cumulant curves do not cross, indicating that there is no phase transition [26]. Notice also from Fig. 8 that the absence of the phase transition is more pronounced for higher values of d . Thus, one can conclude that there is a crossover in the

¹The behavior of the model in the case where they are restricted to agents with opinion $o = -1$ is identical to the case $o = +1$.

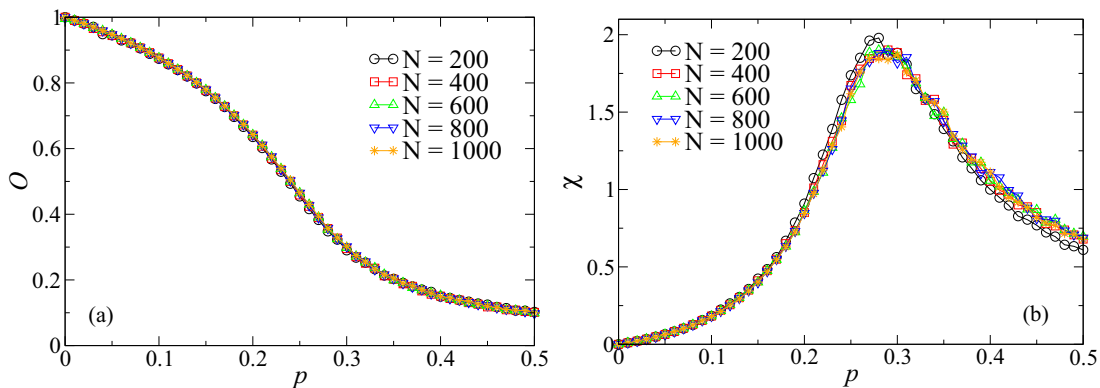


FIG. 7. (Color online) Results for the case where the inflexible agents are chosen among the agents with opinion $o = +1$ for $d = 0.3$. (a) Order parameter and (b) susceptibility as functions of p for different population sizes N . Notice that there is no dependence of the results on the system size. In both cases the interactions $\{\mu_{ij}\}$ are quenched random variables and we have considered asynchronous updates.

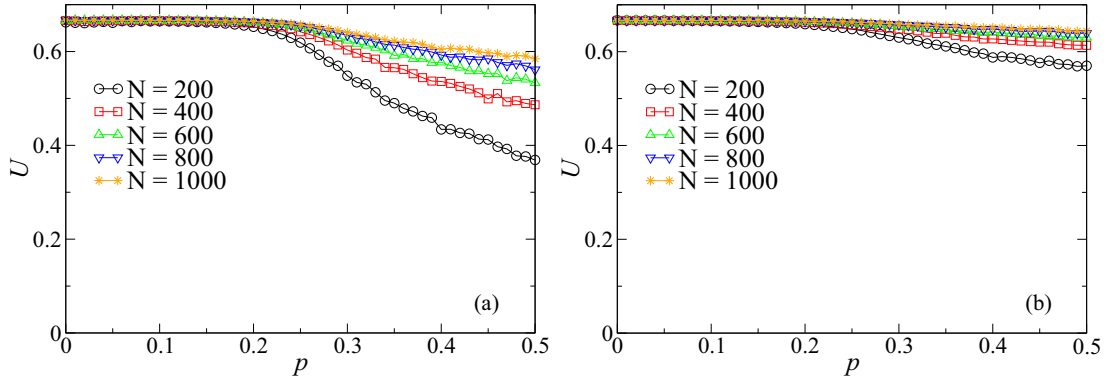


FIG. 8. (Color online) Binder cumulant for the case where the intransigents are chosen among the agents with opinion $o = +1$ for different population sizes N . (a) $d = 0.3$ and (b) $d = 0.6$. Notice the absence of the crossing of the curves. In both cases the interactions $\{\mu_{ij}\}$ are quenched random variables and we have considered asynchronous updates.

population, i.e., the order parameter decreases when we rise the fraction of negative interactions p , but there are no divergences associated with this crossover, suggesting the absence of a phase transition.

Summarizing this section, our results show that when we consider the inflexible agents distributed only among the agents carrying a given opinion, the critical behavior is identical to that of the model in the absence of intransigents ($d = 0$) [11] when the considered opinion is $o = 0$, i.e., we have a transition at $p_c = 1/4$. On the other hand, if the inflexibility is related to the extreme opinions $o = +1$ or $o = -1$, the phase transition is suppressed.

IV. FINAL REMARKS

In this work, we have studied a discrete-state opinion model where each agent carries one of three possible opinions or attitudes, represented by the variables $+1$, -1 , and 0 . The microscopic rules define that agents with the extreme opinions ± 1 should pass by the intermediate (undecided) state 0 before adopting the opposite extreme opinion. We have considered a population of N agents in the mean-field limit, where each individual can interact with all others. The competitive interactions, ruled by negative (with probability p) and positive (with probability $1 - p$) couplings, produce an effect similar to Galam's contrarians. Moreover, a fraction d of the population is constituted by intransigents, averse to changing their opinions. In this sense, our model takes into account both contrarians and intransigents in the process of opinion formation.

The subset of inflexible agents (a fraction d of the population) is randomly selected at the beginning of the simulation, keeping the inflexible character throughout the dynamics. This is a realistic social feature. Indeed, intransigent individuals usually do not change their attitude with time.

We have analyzed cases where the inflexibility is not restricted to a given opinion group, as well as cases where inflexibility is associated with the extreme (± 1) opinions or with a given group supporting one of the three possible attitudes. Moreover, we have also considered that the agents' states (opinions) are updated by means of either sequential (asynchronous) or parallel (synchronous) schemes.

In the first formulation of the model, the inflexible agents are chosen independently of their opinions. By analyzing the quantities of interest (magnetization per spin, susceptibility, and Binder cumulant), we have found that the system exhibits continuous nonequilibrium phase transitions between an ordered phase and a disordered one. The transition points depend on the density d of intransigents, similarly to what happens in other models [19,21]. The simulations show that the values of $p_c(d)$ decrease for increasing values of d , hence the disordered phase broadens with increasing d . Numerical outcomes suggest that there is a critical density $d_c \approx 0.5$ above which the system cannot order, i.e., the system is in a fully disordered (paramagnetic) state for all values of p . The critical exponents on the order-disorder frontier are the same, $\beta = 1/2$, $\gamma = 1$, and $\nu = 2$, independently of d , which means a universality in the model. These results are not affected by the update scheme used (synchronous or asynchronous) nor by the nature of the random couplings (quenched or annealed).

In the second formulation of the model, the inflexible agents are chosen only among the agents with (initially) extreme opinions. In this case, the model behavior is qualitatively similar to the previous one. However, the critical density in this case is greater, $d_c \approx 0.6$. Thus, the ordered phase is larger when the agents with $o = 0$ opinions are free to interact, which is the main fact responsible for the observed differences. However, the critical behavior of the model is robust with respect to the selection of the inflexible agents. In fact, the critical exponents on the order-disorder frontier are the same as in the previous case, $\beta = 1/2$, $\gamma = 1$, and $\nu = 2$, independently of d . This confirms the universal behavior of that phase transition. Again, this result is not affected by the update scheme used (synchronous or asynchronous) nor by the nature of the random interactions (quenched or annealed).

We have also considered the case where the inflexible agents are chosen among the individuals with a given initial opinion. For the case where this opinion is $o = 0$, the critical behavior of the system is not affected by the presence of the intransigent agents, i.e., the phase transition occurs at $p_c = 1/4$ for all values of d . On the other hand, when the intransigents are chosen among the agents with opinion $o = +1$ (or alternatively, $o = -1$), the phase transition does not occur anymore. This conclusion was supported by the behavior

of the quantities of interest. In fact, the order parameter and the susceptibility curves do not depend on the system size, and the Binder cumulant curves for different population sizes do not cross. All these features suggest the absence of the order-disorder transition [28,29].

Notice that in the cases where the phase transition occurs, the critical exponents are always the same, $\beta = 1/2$, $\gamma = 1$, and $\nu = 2$. This is an expected result since we are dealing with a mean-field formulation of the model, where each agent can interact with all others. Observe that the values of β and γ are the same as the mean-field exponents of the Ising model, but the exponent ν presents a different value. As discussed in Ref. [11], interpreting ν as $\nu' D$ where D is the effective dimension in this long-ranged model and considering this effective dimension as $D = 4$, then the value of the effective correlation length exponent becomes $\nu' = 1/2$, which coincides with the mean-field value.

Our results also show that the particular nature of the random couplings μ , as well as of the update scheme, does not affect the results. This may seem an obvious result, however, in other variants of the kinetic exchange opinion model, we observed that the results can be affected by numerical considerations like the fluctuation or not of the pairwise interactions μ (annealed and quenched versions, respectively), or the synchronous or asynchronous update schemes, as was recently shown in Ref. [12].

Despite the simplicity of our model, it can be relevant for the description of real social systems. In our model, $o = 1$ represents a favorable opinion and $o = -1$ an unfavorable

one, while $o = 0$ means indecision. The order parameter considered corresponds to the overall rating and an ordered state means there is a clear-cut decision made. A disordered state means the absence of a decision. Thus, the contrarian effect, quantified by the parameter p , induces a disordered phase for sufficiently large p . In addition, the inclusion of inflexible agents, quantified by the parameter d , makes this effect more pronounced since the critical points p_c decrease for increasing values of d . Thus, the presence of such two effects, contrarians and intransigents, favors the disordered state, indicating that in the presence of extremists it is more difficult to reach a final decision in a public debate, which is a realistic feature of the model. In addition, the results show that the consensus states are never obtained when inflexible agents are present. This is also realistic in elections or public debates in general. In fact, the occurrence of consensus states with the order parameter $O = 1$ occurs in very particular situations, whereas the states with partial order ($O < 1$) are more common [20,27].

We hope that theoretical opinion models considering realistic individuals like contrarians and intransigents may also guide proper new experiments (such as inquiries or surveys) to be conducted for improving the construction of agent-based models, as well as for the validation of such models.

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