

Promotion of cooperation on networks? The myopic best response case

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Abstract. We study the effect of a network of contacts on the emergence of cooperation on social dilemmas under myopic best response dynamics. We begin by summarizing the main features observed under less intellectually demanding dynamics, pointing out their most relevant general characteristics. Subsequently we focus on the new framework of best response. By means of an extensive numerical simulation program we show that, contrary to the rest of dynamics considered so far, best response is largely unaffected by the underlying network, which implies that, in most cases, no promotion of cooperation is found with this dynamics. We do find, however, nontrivial results differing from the well-mixed population in the case of coordination games on lattices, which we explain in terms of the formation of spatial clusters and the conditions for their advancement, subsequently discussing their relevance to other networks.

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1 Introduction

The origin and sustainability of cooperation in animal and human societies is a long-standing puzzle whose importance cannot be overstated [1]. Since the pioneering works by Hamilton [2,3], and the introduction of the theoretical setup of evolutionary game theory [4], a number of reasons have been advanced as possible explanations for the ubiquity and robustness of cooperative behavior [5] (see also [6]). Among these proposals, network reciprocity, or the existence of a (possibly social) network of contacts that determines the individuals a particular one interacts with, has received much attention in the last two decades. This specific line of research started with a seminal work by Nowak and May [7], who studied the Prisoner's Dilemma (PD) [8] game on a square lattice, finding evidence for substantial amounts of cooperation in parameter regions where defection was the only possible outcome in a well-mixed population (i.e., when every individual interacts with every other one). Subsequent work has explored many other choices for the network as well as other dynamical rules for the update of strategies in the game, giving rise to a considerable amount of work [9] which, however, yielded quite a few contradictory results

and no global picture of the observed phenomenology. In fact, only recently [10] such a general conclusion was presented for homogeneous spatial networks, the case of heterogeneous ones being well understood from other recent works [11,12].

In this paper we aim at extending our work on games on spatial structures and homogeneous degree networks to the case when the updating of the strategies follows the myopic best response rule [13,14]. There are a number of reasons that support the relevance of such a study. First, previous works on games on networks considered in general only imitative rules, i.e., a specific individual updates her strategy by imitating the strategy of one of her neighbors selected through different protocols (see e.g. [9] for a review). Such updating procedure makes only modest requirements on the cognitive capabilities and/or information or memory of the players: in these contexts, best response schemes are the next step of sophistication, positing that individuals revise their strategies by choosing the best reply to the strategies used by their neighbors in the previous time step. This choice of updating based only on the previous action of the neighbors is the reason why this dynamics is usually referred to as myopic [15], although for brevity we will just use the term “best response” in what follows. Second, best response not only endows the individuals of the model with more complete intellectual

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capabilities but also is an innovative rule, as it allows extinct strategies to be reintroduced in the system whereas imitative dynamics cannot do that. Third, best response is the rule of choice in most studies from the economical viewpoint, the reason it is not so often considered among physicists being that it gives rise to a differential inclusion rather than a differential equation [16] and, subsequently, it is less amenable to analytical approaches. Finally, earlier works on best response dynamics on lattices [14] left interesting, hitherto unanswered questions such as the strong dependence on the initial conditions of the outcome of the evolution in coordination games, a point that we will specifically address here.

We discuss our results according to the following scheme: in Section 2 we summarize what is known about evolutionary games on networks of homogeneous degree, which includes spatially structured populations and random networks. We will consider in Section 3 the specific case of the Snowdrift game [21] as a paradigmatic example of the difficulties arising in these studies. We will then proceed in Section 4 to present our simulation results on best response dynamics on different types of networks and to subsequently discuss in detail the behavior observed in lattices, providing an explanation as to why there may be more or less cooperation on lattices than on the well-mixed case, depending on the initial conditions. Finally, Section 5 concludes the paper.

2 Evolutionary games on networks of homogeneous degree

2.1 2×2 evolutionary games

The basic ingredient in the models we are going to discuss is evolutionary game theory and, in particular, 2×2 games. Let us now briefly introduce the main concepts. A symmetric 2×2 game is a game with 2 players who choose between 2 strategies and with no difference in role. Each player obtains a payoff given by the following matrix

$$\begin{array}{cc} & \begin{array}{c} C \\ D \end{array} \\ \begin{array}{c} C \\ D \end{array} & \begin{pmatrix} 1 & S \\ T & 0 \end{pmatrix} \end{array} \quad (1)$$

The rows represent the strategy of the player who obtains the payoff and the columns that of her opponent.

The strategies are labeled as C and D for cooperate and defect, because we interpret the game as a social dilemma. Indeed, certain values of S and T undermine a hypothetical situation of mutual cooperation. If $S < 0$ a cooperator faces the risk of losing if the other player defects, performing worse than with mutual defection. If $T > 1$ a cooperator has the temptation to defect and obtain a payoff larger than that of mutual cooperation. Both tensions determine the social dilemmas represented by symmetric 2×2 games [17]. Restricting the values of the coefficients within the intervals $-1 < S < 1$ and $0 < T < 2$, we have the Harmony game [18] (HG, $0 < S, T < 1$) and three classic social dilemmas: the Prisoner's Dilemma

(PD, $-1 < S < 0, 1 < T < 2$), the Stag-Hunt game [19] (SH, $-1 < S < 0 < T < 1$), and the Hawk-Dove [20] or Snowdrift game [21] (SD, $0 < S < 1 < T < 2$). Each game corresponds, thus, to a quadrant in the ST -plane.

To study the competition between cooperation and defection from an evolutionary perspective, the payoffs obtained by playing the game are considered as fitness and a darwinian dynamics is introduced to promote the fittest strategy. The classic framework to do so is the replicator dynamics [16,22], which assumes an infinite and well-mixed population, i.e. a population with no structure, where each individual plays with every other. Let x be the density of cooperators, and f_c and f_d the fitness of a cooperator and a defector, respectively. The replicator dynamics states that x evolves as [16]

$$\dot{x} = x(1-x)(f_c - f_d). \quad (2)$$

Then, if cooperators are doing better than defectors their density rises accordingly, and the opposite occurs if they are doing worse. Provided that the initial density of cooperators x^0 is different from 0 and 1, the asymptotic state of this dynamical system is, for each game (x^* represents the asymptotic density of cooperators) [16]: HG, full cooperation, $x^* = 1$; PD, full defection, $x^* = 0$; SH, full cooperation if $x^0 > x_e$, or full defection if $x^0 < x_e$; SD, mixed population with $x^* = x_e$, regardless of the initial density x^0 . Both in SH and SD the coexistence equilibrium has a cooperation density $x_e = S/(S+T-1)$. It is important to note that the outcome of these four games encompasses all the possible cases for any symmetric 2×2 game [23] (see also [6]).

2.2 2×2 evolutionary games on networks

As we stated in Section 1, in 1992 Nowak and May [7] introduced spatial structure in the context of evolutionary games by considering the players located at the nodes of a square lattice, playing the game only with their neighbors (and playing the same action vs every one of them) and not with the whole population. They introduced evolution in this setup by using the unconditional imitation rule (also known as "imitate-the-best" [9]), where each player chooses the strategy of the neighbor with largest payoff, provided this payoff is greater than the player's. With this rule, they found that cooperators survived by self-organizing in clusters, where the interactions within the clusters yielded larger payoffs to cooperators than those obtained by defectors at the boundaries of cooperators' clusters.

Nowak and May's pioneering work opened the way to a large number of studies focused on different games, different evolutionary rules, and different lattices or networks. As a loose conclusion of those works, it was generally believed that the existence of structure in the population, whether spatial or of another kind, enhanced the emergence of cooperation in games where defection was the norm. However, such conclusion did not agree with all the available research, contradictions arose at several points

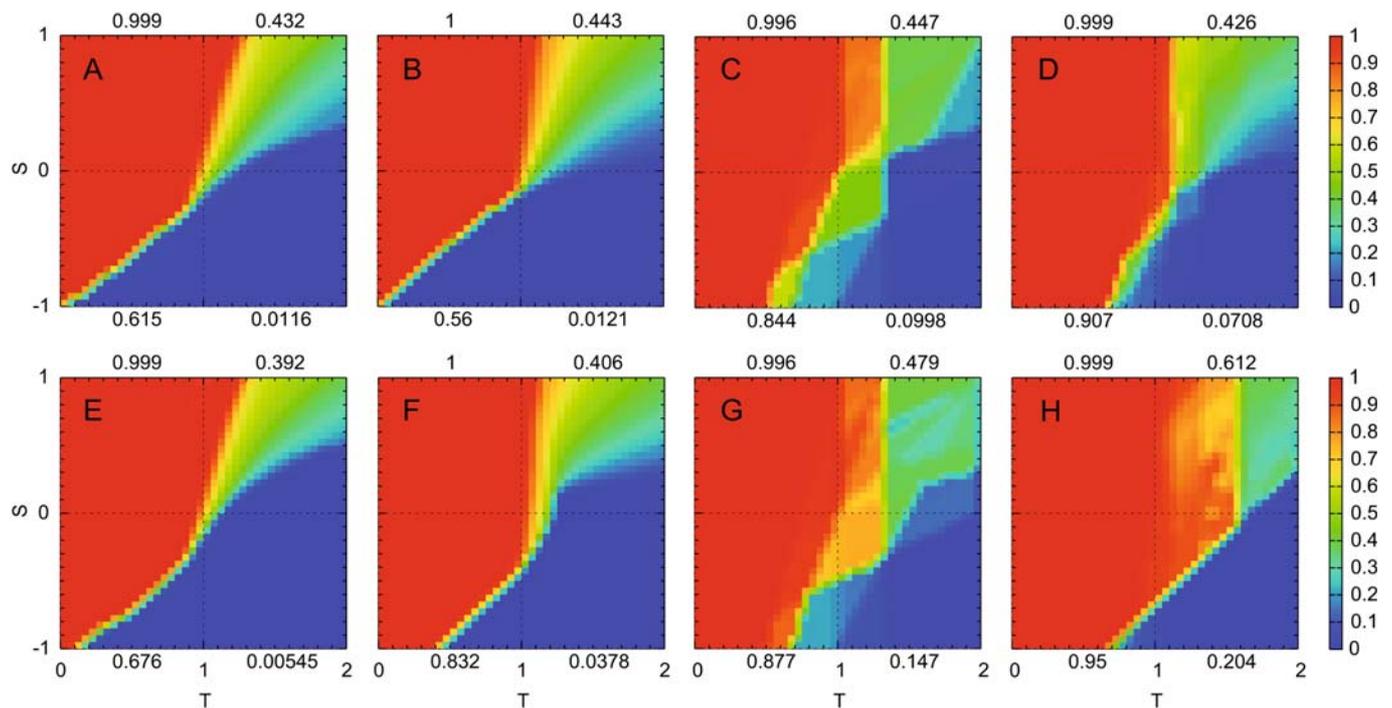


Fig. 1. (Color online) Asymptotic density of cooperators x^* in homogeneous random networks (upper row, A to D) compared to regular lattices (lower row, E to H), with degree $k = 4$ (first and third columns, A, E, C, G) and $k = 8$ (second and fourth columns, B, F, D, H). Left plots (A, B, E, F) correspond to the replicator update rule, whereas right ones (C, D, G, H) use unconditional imitation as update rule. The initial density of cooperators is $x^0 = 0.5$ in all cases. The graphs display the key role of both the clustering of the network and the update rule (see main text). The promotion of cooperation is, in general, restricted to SH. The influence on PD is only significant when the update rule is unconditional imitation.

(an example of which will be discussed in Sect. 3 below) and there were no studies that identified parameter regions where one could firmly establish it. Therefore, in [10] we carried out a very ambitious simulation program, as a result of which we were able to reach some unambiguous conclusions. While a full report of our results is available in [10], we find it convenient to briefly recall here a few of the ideas presented there, both as background information for the reader as well as to introduce the way we will present our new results.

One of the most relevant findings reported in [10] was that the spatial structure of a population, when modeled by a regular lattice, only has a significant effect on cooperation when the clustering coefficient is high, as seen by comparing results with those of homogeneous random networks of the same degree. This is illustrated by Figure 1, where results for random homogeneous networks (upper row) are compared to results on regular lattices (lower row). Only when there is high transitivity or clustering in the network [24], as occurs for regular lattices of degree $k = 8$, significant differences appear.

Figure 1 also highlights the crucial influence of the update rule. Right columns show the results with unconditional imitation (introduced above), while the left ones present those obtained with the so-called replicator rule [16,22]. This rule is defined as follows: let $i = 1 \dots N$ label the individuals in the population. Let s_i be the strategy of player i , π_i her payoff and N_i her

neighborhood, of size k . With the replicator update rule one neighbor j of player i is chosen at random, $j \in N_i$. The probability of player i adopting the strategy of player j is given by

$$p_{ij}^t \equiv \mathcal{P}\{s_j^t \rightarrow s_i^{t+1}\} = \begin{cases} (\pi_j^t - \pi_i^t)/\Phi & : \pi_j^t > \pi_i^t \\ 0 & : \pi_j^t \leq \pi_i^t \end{cases}, \quad (3)$$

with $\Phi = k(\max(1, T) - \min(0, S))$ to ensure $\mathcal{P}(\cdot) \in [0, 1]$.

Figure 1 presents the results for the space of 2×2 games as a whole, using a color code which will be the same hereafter. Furthermore, we have introduced a quantitative measure \mathcal{C}_G for the overall asymptotic cooperation in game G ($=$ HG, PD, SH, SD), given by the mean value of x^* over the corresponding region in the ST -plane. This global index of cooperation has a range $\mathcal{C}_G \in [0, 1]$ and appears on the quadrant of each game. Both the qualitative assessment of the plots and the comparison of the values of this quantitative index yield an important regularity in the effect of the social structure modeled by this kind of networks: cooperation is generally enforced in coordination games (SH), specially when the clustering coefficient is high. The influence on anti-coordination games (SD) depends on the update rule and, remarkably, the positive effect on PD requires a particular update rule, namely unconditional imitation. We refer the interested reader to [10] for a complete discussion on these and related issues.

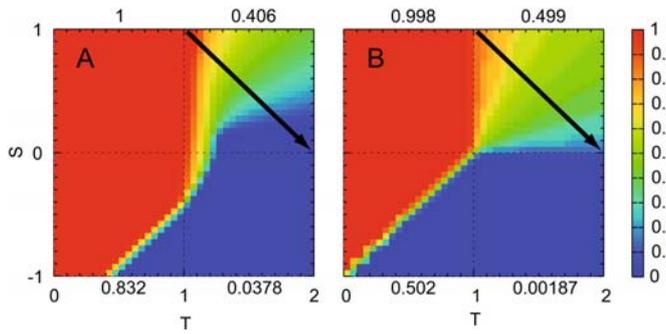


Fig. 2. (Color online) Simulations of the models discussed in [25] (A) and [26] (B). The lines mark the range of parameters studied in both works. See main text for a discussion.

3 The case of SD

Previous section introduced an important property of these evolutionary models, the crucial dependence that the result has on the update rule. We will now discuss a specific example which, on the one hand, allows us to make this point and, on the other hand, will introduce us to our main subject, namely the effects of best response dynamics.

In 2004, Hauert and Doebeli [25] presented a study of the Snowdrift game in which they concluded that, contrary to the general belief, the existence of a spatial structure may inhibit cooperation. To that end, they studied the dependence of the level of cooperation on the parameter r , the cost-benefit ratio (understood as cost of cooperating and corresponding benefit accrued), which in our parameterization is $r = (T - S)/2$. They did not consider the whole ST -plane but only the line given by $T = 1 + r$, $S = 1 - r$ (see Fig. 2) and obtained asymptotic levels of cooperation below those found in well mixed populations.

One year later, Sysi-Aho and coworkers [26] repeated the same study changing only the dynamical rule: where Hauert and Doebeli had used the replicator rule mentioned above, Sysi-Aho and coworkers used the best response rule, introducing a probability $p < 1$ to update the strategy, i.e., at every time step every player chose her strategy as a best response to their neighbors with probability p or left it unchanged with probability $1 - p$. The reason to do that is to prevent the systems from falling onto a sequence of alternate states of full defection and full cooperation, which is an artifact of the rule (and which are never reached as soon as $p < 1$, see [26]). They studied the same range of parameters than [25], finding that cooperation subsisted even for r close to 1, being larger (resp. smaller) than in a well mixed population for large (resp. small) r .

In order to understand better this issue, we have reproduced the simulations in [25,26], using the same square lattices, neighborhoods ($k = 8$) and initial conditions (cooperation and defection equally likely) they used, obtaining the results we summarize in Figure 2. As we may see, the two different dynamics lead to rather different results when looked at in the framework of the whole ST -plane. We stress that, along the line indicated in the plots, we

exactly reproduce the results reported in those previous studies. There is indeed a decrease of cooperation when using the replicator dynamics as in [25], along the line of interest but also in the SD quadrant as a whole. However, when the dynamics is best response we do observe a stepped profile in the SD quadrant as Sysi-Aho et al. did, while the mean value of the cooperation over the quadrant is the same as in a well mixed population. The reader is referred to the middle panel of Figure 3, where the results on a well mixed population with the same initial condition are depicted.

These results open up a series of questions, beginning with the following: what is the effect of best response dynamics on other networks, given that on a square lattice its effect is not very noticeable? Let us recall that best response is a step further towards “intelligence” of the agents as compared to the replicator rule, an imitative, non-innovative dynamics, and therefore we might expect that players could exploit better the existence of a network. This issue is what we discuss in detail in what follows. However, there is a more general problem, namely what does it mean “promotion of cooperation by the structure of a population”? Do we refer to a specific set of parameters, such as the line studied in [25,26]? Do we refer to a global measure of the cooperation level in a region, such as the values we compute for each quadrant? Or do we refer to the ST -plane as a whole? Note that in this last case the replicator rule used by Hauert and Doebeli, while indeed leading to less cooperation in the SD game, yields a very large increase of the cooperative region in the SH quadrant, favoring players to coordinate in the Pareto-dominant equilibrium [22]. Had we been looking at that quadrant only, we would have certainly concluded that cooperation is promoted by the square lattice. Or, looking at the SD game, had we considered unconditional imitation as the update rule, we would have found evidence of spatial structure fostering cooperation in SD. While we will not dwell any further in this issue here (but see [10] for a discussion in depth of these problems) we want to stress that statements about promotion of cooperation should be made in a much more specific manner without trying to attach to them unchecked general implications beyond the scope of the case under study.

4 Best response dynamics

4.1 Different types of networks

Motivated by the reasons discussed in the introduction and by the issues raised by our study in Section 3, we undertook the study of the effects of the best response dynamics on a large family of networks. We considered lattices, which may represent spatial structure, homogeneous random networks (random networks where all nodes have exactly the same degree), Erdős-Renyi random networks [27], small-world networks, Barabási-Albert [28] scale free networks and Klemm-Eguíluz [29] scale free networks with different mean degrees, thus exploring all possible combinations of small-world phenomena, scale

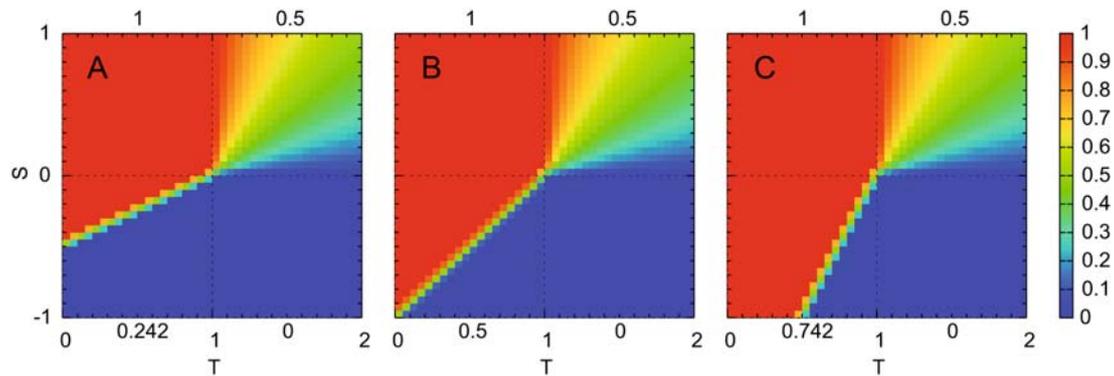


Fig. 3. (Color online) Asymptotic density of cooperators x^* in well mixed populations under best response dynamics, with $p = 0.1$ (see main text), when the initial density of cooperators is $x^0 = 1/3$ (left, A), $x^0 = 1/2$ (middle, B) and $x^0 = 2/3$ (right, C).

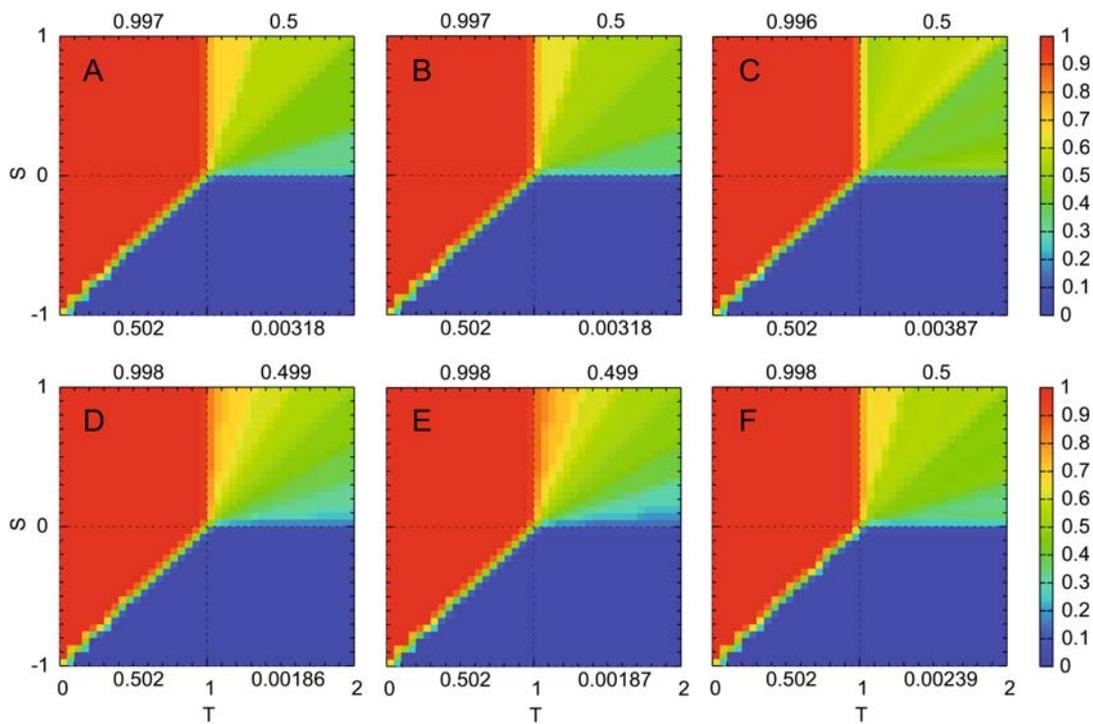


Fig. 4. (Color online) Asymptotic density of cooperators x^* in random (left, A and D), regular (middle, B and E), and scale-free networks (right, C and F) with degrees $k = 4$ (upper row, A to C) and 8 (lower row, D to F). The update rule is best response with $p = 0.1$ and the initial density of cooperators is $x^0 = 0.5$. Differences are negligible in all cases; note, however, that the steps appearing in the SD quadrant are slightly different.

free behavior and large or small clustering. We have also considered the complete graph as the reference for a well mixed population, results for which are presented in Figure 3.

We will begin by focusing on the case in which the initial conditions are a 50% of cooperators and a 50% of defectors. Our reference for comparison will then be Figure 3B, corresponding to a well-mixed (complete network) population with that initial condition. A small subset of our results for the other networks is presented in Figure 4, showing that the asymptotic behavior does not depend at all on the type of network considered, with the only and unimportant exception of slight differences in the stepping

in SD. We want to stress that we have tested many other networks aside from those presented here with exactly the same results. Not only the mean cooperation levels per quadrant are practically the same, but also the dependence on S and T for each quadrant. We note that this is a very remarkable result, in so far as for most other (imitative) dynamics studied there is always a largely noticeable effect of the type of network on which the games are played, as we have seen above (cf. Figs. 1 and 2; see also [10,11] for more details on homogeneous and scale free networks, respectively). We have made every effort to ensure that our results are robust and independent of the technicalities of the simulations. To begin with, we have

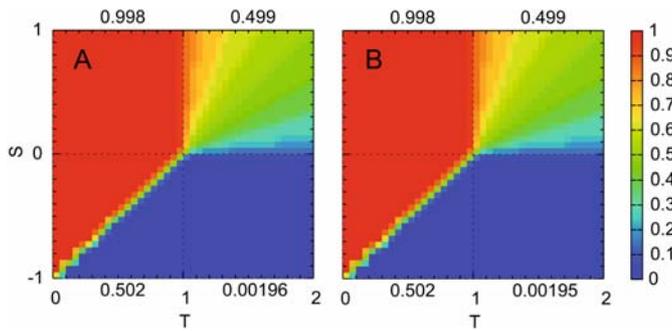


Fig. 5. (Color online) Asymptotic density of cooperators x^* in regular lattices with degree $k = 8$. The update rule is best response and the initial density of cooperators is $x^0 = 0.5$. Left (A): asynchronous updating with $p = 0.1$; right (B): synchronous updating with $p = 0.01$. Both results are virtually identical to that of Figure 4E, which was obtained with synchronous updating and $p = 0.1$.

checked a correct convergence, verifying that simulation times larger up to a factor of 10 lead to the same results. On the other hand, the probability parameter p , which governs how often agents update their strategy, is not relevant, and neither it is the use of asynchronous updating (see Fig. 5), in which an agent is chosen at random to update its strategy after a round of the game has been played by her and her neighbors (we note that asynchronicity was reported to have a small but noticeable effect with imitative update rules in [10]). With all these checks, we can finally state our first conclusion, namely that best response dynamics is insensitive to the type of network on which the game is played, meaning that when the networks are initiated from the same proportion of cooperators and defectors the asymptotic results are the same as well. This unexpected result may have interesting implications in applied contexts, implications which we will discuss in the concluding section.

4.2 The effect of initial conditions

Having found that best response is largely unaffected by playing the different 2×2 games on a network, we extended our analysis of the problem to consider different initial conditions. We initiated our simulations with an initial density of cooperators $x^0 = 1/3$ or $x^0 = 2/3$ and repeated our simulation program for our family of networks. In order to understand the observed phenomenology, we find it convenient to begin the discussion by the case of lattices. Figure 6 shows the results for lattices with different number of neighbors and $x^0 = 1/3$ (the outcome of the simulations for $x^0 = 2/3$ is similar, with the green region in the SH quadrant of Fig. 5 being symmetrically located below the $S = T - 1$ line). As can be seen from Figure 6, for 4 and 6 neighbors the results for $x^0 = 1/3$ are indistinguishable from the results for $x^0 = 1/2$. This must be compared with Figure 3A, which shows for a well-mixed population a cooperative region in the SH quadrant that is roughly a 50% of the one we have obtained on

lattices. As the asymptotics is the same for lattices with 4 and 6 neighbors, we also conclude that the clustering of the network does not play any role. However the lattice with 8 neighbors shows a striking result, namely the appearance on the SH quadrant of a region with intermediate values of cooperation. Let us recall that, as a stand-alone game, SH has two equilibria, full cooperation or full defection, and that depending on the initial condition the system ends up in one or the other. This is the behavior observed in all the lattices and networks studied so far with imitative rules and also was what we observed for best response with $x^0 = 0.5$.

In order to understand this surprising feature, the appearance of regions with seemingly mixed behavior in SH and its dependence on the number of neighbors, appearing for 8 neighbors but not for smaller numbers, we have looked in detail at the time evolution of specific realizations. Figure 7 presents a representative example of this dynamics with 8 neighbors and for values of S and T in the intermediate region we are interested in. We clearly observe that there is an initial stage in which isolated cooperators (in red) die out and only clusters with a significantly larger presence of cooperators avoid extinction and eventually begin to grow. A careful look at the pictures and at the whole time evolution allows one to notice that clusters grow due to the advancement of some of its sides, in particular those which are basically flat but have just one kink, whereas some others are stopped unless hit by some other advancing front. Very clearly observable among the latter are diagonal fronts (see, in particular, the last three frames in the sequence). This can be easily understood analytically by just considering the different possible types of front and the conditions for their advancement in terms of best response dynamics. Thus, for a planar front such as

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C C C D D D
C C C D D D
C C C D D D
C C C D D D
C C C D D D
C C C D D D

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the borderline defectors will become cooperators if $S > 3(T - 1)/5$, whereas cooperators are transformed into defectors if $S < 5(T - 1)/3$. Therefore, in the intermediate region $5(T - 1)/3 < S < 3(T - 1)/5$ planar fronts are stable in the lattice with 8 neighbors. It is easy to check that diagonal fronts are stable in this region as well. This must be compared with the situation when there is a kink in the front, as in

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C C C D D D
C C C D D D
C C C D D D
C C C C D D
C C C C D D
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in this case the defector at the kink becomes a cooperator if $S > T - 1$, and the cooperator at the kink becomes a defector in the opposite case, $S < T - 1$. Therefore,

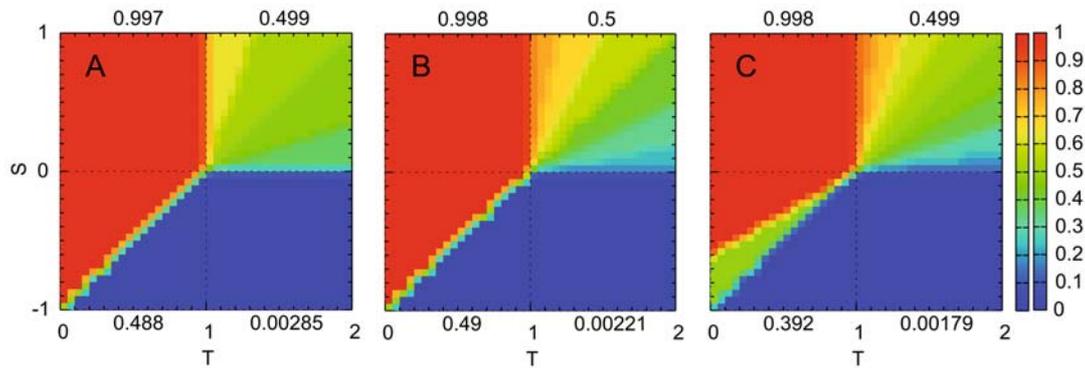


Fig. 6. (Color online) Asymptotic density of cooperators x^* in regular lattices with initial density of cooperators $x^0 = 1/3$. The degrees are $k = 4$ (left, A), $k = 6$ (middle, B) and $k = 8$ (right, C). The update rule is best response with $p = 0.1$. Equivalent (symmetrical) results are obtained for $x^0 = 2/3$ (see main text).

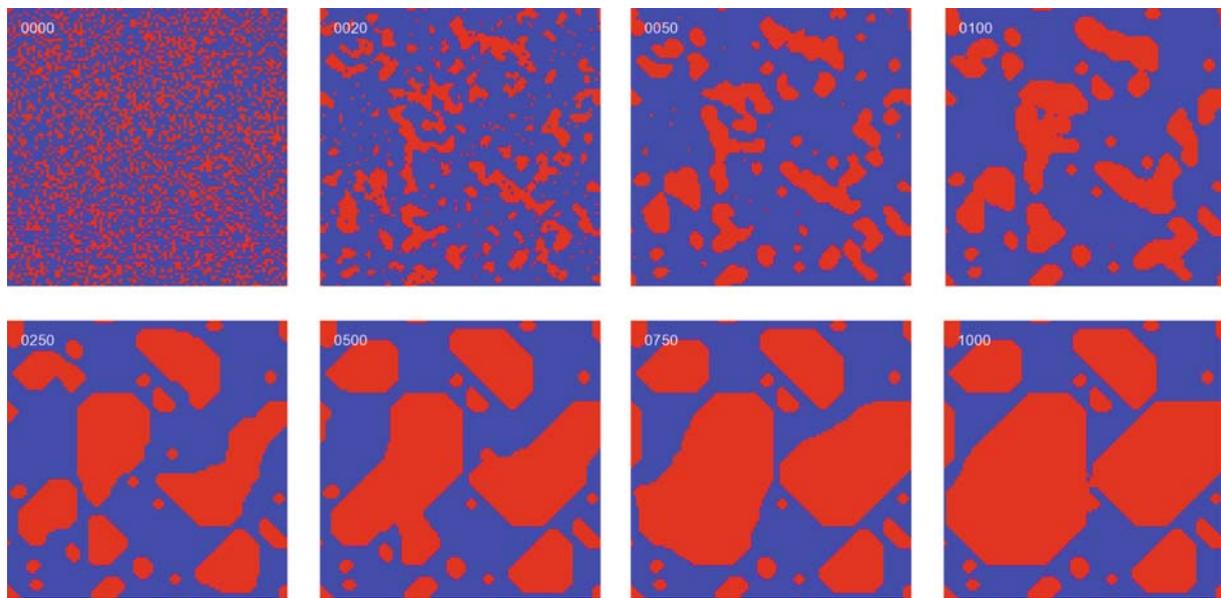


Fig. 7. (Color online) Snapshots of the evolution on a regular lattice of degree $k = 8$, under best response dynamics with $p = 0.1$. Sites depicted in red are cooperators, blue ones are defectors. The initial density of cooperators is $x^0 = 1/3$. The game parameters are $S = -0.6$, $T = 0.2$, in the SH quadrant. Times are as indicated in the snapshots. System size is 100×100 with periodic boundary conditions. In this particular realization, cooperation eventually dominates entirely the lattice.

for this type of front there is no stability region and it always advances in one direction or the other. This analysis shows that the fate of a specific realization depends on the geometry of the clusters arising at its first stage. What we see in the special region in Figure 6C is the result of some simulations that cannot reach full cooperation because the growing clusters arrest at some point in the simulations, whereas in other cases their geometry is such (as in Fig. 7) that the system ends up dominated entirely by cooperators. We note that the discussion is symmetrical with respect to the initial condition and it can be applied to explain the results for $x^0 = 2/3$ (not shown).

It remains to be explained why the other lattices do not show this region of intermediate behavior. The reason can again be traced back to the geometry of the clusters: consider, for instance, the case with 4 neighbors. One can then show that diagonal fronts are never stable and, in

fact, are subject to the same conditions as planar fronts with a kink, making it much more difficult for a particular realization to have all its clusters arrested. Hence, in the simulations on the lattice with 4 neighbors situations with a mixed population of cooperators and defectors are never found. A similar reasoning explains the results for the 6 neighbor lattice.

The mechanism behind the appearance of the mixed populations suggests that the coexistence of cooperators and defectors in the SH may be due to finite size effects [32,33]. To check this possibility, we have simulated the lattice with 8 neighbors on a large variety of sizes, plotting a histogram of the number of realizations that stop on a mixed population. The result is depicted in Figure 8, showing clearly that as the system size grows the number of asymptotically mixed realizations becomes negligible. This is a strong indication that in the infinite

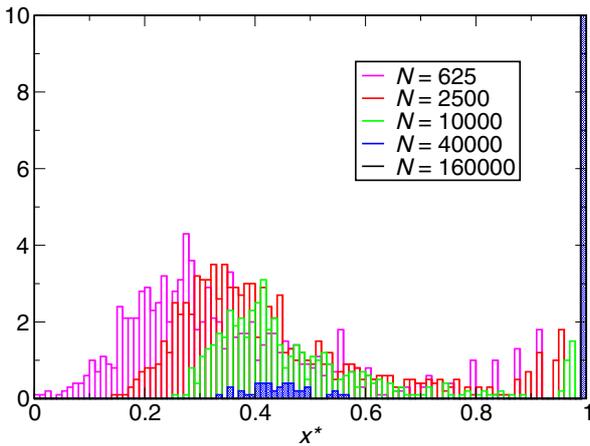


Fig. 8. (Color online) Normalized histogram of the asymptotic densities of cooperators x^* reached in the simulations on regular lattices of degree $k = 8$ with initial density of cooperators $x^0 = 1/3$. The update rule is best response with $p = 0.1$. Sizes are as indicated in the plot. Note that the value at $x^* = 1$ goes out of range for the two largest sizes. S and T are as in Figure 7.

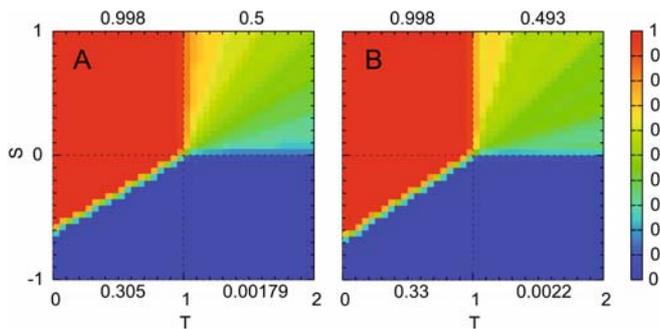


Fig. 9. (Color online) Asymptotic density of cooperators x^* in homogeneous random (left, A) and scale free (right, B) networks with initial density of cooperators $x^0 = 1/3$. The update rule is best response with $p = 0.1$. Equivalent (symmetrical) results are obtained for $x^0 = 2/3$.

size limit the 8 neighbor lattice will behave as those with less neighbors, arriving always at the same asymptotics as for the $x^0 = 0.5$ initial condition, at least for initial data not too close to $x^0 = 0$ or $x^0 = 1$. We note, however, that for applications of these ideas to real life problems, where populations (e.g., in a social context) are finite, it may be possible to observe mixed populations in SH dilemmas under best response dynamics, if the population is structured spatially.

Finally we have also looked at the effect of initial conditions on other network models besides regular lattices. Two examples of our simulations are presented in Figure 9. By comparing with the well mixed result in Figure 3A, we see that initial conditions different from $x^0 = 0.5$ also give rise to non trivial effects on random and scale free lattices. No regions of mixed populations are observed, and the mean cooperation level increases slightly when going from well mixed to homogeneous random and from there to scale free networks, but always to a lesser extent than

on lattices. Unfortunately, the topology of these networks renders the analysis in terms of clusters unapplicable. We do not have a clear understanding of this small promotion of cooperation (inhibition in the case $x^0 = 2/3$) on these networks, but we can tentatively hypothesize the development of weak, long range correlations, which would be responsible for such a weaker effect, in contrast to those of lattices, of a short and stronger nature.

5 Discussion and conclusions

In this paper we have presented a thorough analysis of myopic best response dynamics on 2×2 social dilemmas with different types of networks. The original motivation of our results was the clarification of the discrepancies among the work of Hauert and Doebeli [25] and Sysi-Aho et al. [26], who reported an inhibition (resp. promotion) of cooperation in the SD game on square lattices. We have shown that the problem is that the comparison is basically meaningless in so far as they use different update rules (proportional update or replicator dynamics vs. best response) and also because restricting the comparison to SD games leads to missing the larger picture of the set of possible dilemmas. Indeed, with the proportional update rule there is a large increase of cooperation in the SH quadrant, which contrasts with the phenomenology of the SD dilemma (as is generically the case, see [10]).

Our analysis of myopic best response dynamics has yielded two main results: first, there is practically no effect of the type of network considered on the asymptotic behavior for any value of the game parameters; and, second, there is a noticeable effect of the initial conditions with particular importance in the case of lattices. The first result is most interesting, more so if viewed in the context of previous work on imitative rules [10,11]. In principle, given the definition of best response dynamics, one would not expect important effects on the Harmony and PD quadrants, where there is only one global best response. Note that this is also the outcome obtained on well-mixed populations and on regular lattices with stochastic imitative rules [10]. However, the fact that SD has only one stable equilibrium of a mixed character, that may be difficult to fulfill in the presence of a network, or the bistable nature of SH allow to expect some influence from the interaction networks. The fact that there is none (at least at a global level, for the whole network; local peculiarities are not being considered here) is therefore quite remarkable and casts a shadow of doubt on the applicability of the studies of imitative rules to specific socio-economic contexts. Indeed, if using a rule with a higher degree of intelligence renders the network effect unnoticeable, one can presume that the behavior of actual economic agents will be closer to best response and hence network-independent to some extent. Interestingly, this idea may be related to the fact that models of network formation in economics often lead to very simple networks (see, e.g., [30,31], and references in the latter), which might be another hint of the secondary role played by the network structure in economical applications.

On the other hand, we have also seen that there are network effects even under best response dynamics, that arise in the dependence of the asymptotic behavior on the initial conditions. Lattices are prominent examples of these phenomena and, using the regularity of their structure, we have been able to trace the differences with well-mixed populations to the formation of clusters, whose stability depends on the degree of the lattice (at least for finite size lattices). We have thus seen that cooperation may be promoted (resp. inhibited) in the SH quadrant in situations when the initial amount of cooperators would be insufficient (resp. sufficient) to take the system to full cooperation. The effect is also observed on other networks, always with the same symmetrical character, but the reason for this behavior remains still an open issue. Again, in an economical context this result is interesting because we see that best response, when used on a network, exhibits a dependence on the initial conditions different from that of the well-mixed setting.

Finally, from a more general viewpoint, we have presented yet another example that the phenomenology of social dilemmas on networks is largely non universal. While best response turns out to be peculiar in the sense that it gives rise to network-independent behavior, it is important to realize that once again changing the dynamics of the strategy update leads to large, non-trivial changes in the results of evolution as compared to other rules. Furthermore, as we have already mentioned, only a full study of the whole space of dilemmas under consideration (in our case, 2×2 games) can shed some light on the mechanisms governing the evolution on different networks and under different rules. In this sense, we have presented here a detailed analysis of best response games which highlights the fact that, as compared to what is observed on well-mixed populations, the dilemma that is most affected is SH, i.e., risky situations, rather than contexts in which the important tension is the temptation to defect. This is clearly related to the bistable character of SH, in which the best response tends to be equal to the opponent's actions, and suggests that similar mechanisms may be at work in games and dilemmas with more players and/or strategies. Further research along these lines is needed to confirm these intuitions.

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