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OPEN Reciprocity in spatial evolutionary public goods game on doublelayered network

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Spatial evolutionary games have mainly been studied on a single, isolated network. However, in real world systems, many interaction topologies are not isolated but many different types of networks are inter-connected to each other. In this study, we investigate the spatial evolutionary public goods game (SEPGG) on double-layered random networks (DRN). Based on the mean-field type arguments and numerical simulations, we find that SEPGG on DRN shows very rich interesting phenomena, especially, depending on the size of each layer, intra-connectivity, and inter-connected couplings, the network reciprocity of SEPGG on DRN can be drastically enhanced through the inter-connected coupling. Furthermore, SEPGG on DRN can provide a more general framework which includes the evolutionary dynamics on multiplex networks and inter-connected networks at the same time.

Cooperation is a ubiquitous phenomenon in nature from micro-organisms to human society. The emergence of cooperation among the selfish individuals has been a long lasting conundrum in various scientific disciplines¹⁻¹². There have been many attempts to explain how the cooperation emerges through the interaction among selfish individuals. Among those studies game theory has provided an important theoretical framework to understand the emergence of cooperation through the strategic interactions among individuals. And it has been successfully applied to diverse fields such as evolutionary biology and psychology¹, computer science and operations research^{2,3}, political science and military strategy^{4,5}, cultural anthropology⁶, ethics and moral philosophy⁷, economics^{8,9}, traffic flow^{10,11} and public health¹². The central aim of game theory is to determine conditions needed for cooperation to emerge between egoistic individuals¹³⁻¹⁵. Recently, many studies have focused on spatial evolutionary games to understand how steady-state strategies emerge in various structures and to identify the characteristic features of steady-state strategies¹⁵⁻¹⁸. Interestingly, in structured population, individuals only interact with their nearest neighbors and it becomes possible for cooperators to survive by forming clusters in which they defend themselves against defectors' exploitation. This is known as network reciprocity¹⁶⁻²

Spatial evolutionary games have mainly been studied on a single, isolated network. However, empirical evidences show that many, if not all, real world systems are not isolated but many different types of networks are interlinked²⁶. For example, family, friendship and work-related networks are interlinked by each individual in society. Metabolic synthesis, protein-protein interaction, signaling and regulatory networks altogether constitute an inter-cellular network in a cell. Various financial, trade and political networks are also interlinked to form a global economic system. The ecological system is also composed of different level of hierarchical networks. Thus, the game theory on a single network cannot provide a complete explanation on how the cooperation emerges in nature. Only recently, spatial evolutionary games on the interdependent networks, multiplex networks, and interconnected networks have been studied to understand how imitation and interaction between the networks influence the final cooperation levels²⁶⁻³².

In the interconnected network, there are actual physical links between different networks rather than the dependency links in interdependent networks. Propagation of microcredit across the countries, possibly across the interconnected networks is an example of the propagation of cooperations from one network to another. In Bangladesh, microcredit has grown in popularity in the 1970s. Group-lending is a key part of microcredit. The loan to one participant in group-lending depends upon the successful repayment from another member, thus cooperation among participants is very important in microcredit. After few years, microcredit is widely used in developing countries and is presented as having "enormous potential as a tool for poverty alleviation"³³. In 2007, there are more than 500 organizations in the United States that provide to microcredit owners³⁴.

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To understand how the coupling between networks develops the cooperation or the network reciprocity, in this report, we investigate the spatial evolutionary public goods game (SEPGG) on double-layered random networks (DRN's). Especially, the spatial evolutionary public goods game (SEPGG) has attracted considerable attention, because it offers valuable insights into prevailing socioeconomic problems such as pollution, deforestation, mining, fishing, climate control and environmental protection¹⁸. SEPGG has been intensively studied to find interesting results such as the cyclic dominance^{35,36}, transition nature³⁷, payoff distribution³⁸, and the effects of underlying topologies^{19,26–31,38–44}. The effects of loner^{39,45}, punishment and reputation⁴⁶, and noise⁴⁷ on the emergence of cooperation have been also investigated. As a prototype of interconnected network, we consider a double-layered random networks (DRNs), but the generalization to interconnected network with more than two layers is straightforward. A DRN consists of two random networks with any size and average degree. If the size of each layer is the same then the DRN can be regarded as a multiplex network, in which each layer has the same set of nodes. On the other hand, if the size of each layer is different, then each node of one random network with the smaller size is linked to a randomly chosen node of the other network. In this case the DRN can be regarded as an interconnected network in which two different networks are interlinked through the interconnected links. Therefore, the DRN considered in this study can provide a more general framework to investigate the emergence of cooperation in various types of interlinked networks. We use the biased imitation process²⁴, in which a randomly chosen agent imitates the strategy of the interlinked neighbor on the opposite layer with probability p, or that of a randomly chosen node among intralinked neighbors on the same layer with the probability 1 - p. By simulations, we obtain the steady-states depending on *p*. Especially, for p > 0, we find that the anomalous cooperator-enhanced states on the layer which have no cooperators for p = 0. This anomalous cooperator-enhance states resembles the propagation of microcredit. We also explain theoretically how this network reciprocity occurs.

Previous Study

To understand "Tragedy of the commons"⁴⁸ problem with large participants has been studied through the SEPGG on the complete graph (CG) and dense random networks¹⁷. Depending on the multiplication factor *r* and the size of graph *N*, either **Loner-only state (L-state)**, which is the anomalous state with no active participants, or **Defector-only state (D-state)**, which means the state of "tragedy of the commons", has been shown to appear on CG¹⁷. Furthermore, we have shown the following crossover behaviors as the mean-degree $\langle k \rangle$ of underlying random networks changes¹⁷. For small *r*, the L-state crosses over to the D-state and the D-state successively crosses over to the **Cooperator-only state (C-state)** as $\langle k \rangle$ decreases. For large *r*, the direct crossover from the D-state to the C-state occurs as $\langle k \rangle$ decreases. We have been found that cooperation gradually increases as the number of participants or $\langle k \rangle$ decreases, which is the origin of these crossovers. Hence, the crossovers describe how the enhanced cooperation on sparse networks overcomes "tragedy of the commons" on dense networks.

Results

SEPGG on double-layered random network. Now we want to explain how the **double-layered random networks (DRNs)** are composed. The first random network (layer) with the size N_1 and the average intradegree $\langle k_{intra} \rangle_1$ and the second random layer with $N_2(\geq N_1)$ and $\langle k_{intra} \rangle_2$ are separately constructed. Then, to make DRN with $n_{inter}(\leq N_1)$ interlinks, n_{inter} different nodes both on the first layer and the second layer are chosen randomly. Each chosen node on the layer 1 is made to be randomly linked to a chosen node of the second layer without making multiple interlinks to a certain node. (See Fig. 1). We call a DRN with $N_1 = N_2(=N)$ and $\langle k_{intra} \rangle_1 = \langle k_{intra} \rangle_2$ a **symmetric DRN** and a DRN with $N_1 \leq N_2$ or $\langle k_{intra} \rangle_1 \neq \langle k_{intra} \rangle_2$ an **asymmetric DRN**.

SEPGG model on a constructed DRN is defined as what follows. Each agent is assigned to a node on DRN. The strategy s_i of the agent on a node *i* can be Cooperator (C), Defector (D) or Loner (L). In each update an agent *i* is randomly chosen. First, we calculate the payoff P_i of *i* using the following rule. Let $n_{i,C}$ be the number of agents with C, $n_{i,D}$ be that with D and $n_{i,L}$ be that with L among the $k_{i,intra} + k_{i,inter} + 1$ agents. Here $k_{i,intera}$ ($k_{i,inter}$) is the intradegree (interdegree) of node *i* and $n_{i,C} + n_{i,D} + n_{i,L} = k_{i,intra} + k_{i,inter} + 1$. P_i is given by,

$$P_{i} = \begin{cases} \frac{rcn_{i,C}}{n_{i,C} + n_{i,D}} - c & \text{if } s_{i} = C \\ \frac{rcn_{i,C}}{n_{i,C} + n_{i,D}} & \text{if } s_{i} = D \\ \sigma & \text{if } s_{i} = L \end{cases}$$
(1)

Here, *c* is the cost contributed to the common pool by a C, r(>1) is the multiplication factor and σ is the fixed payoff of an L. We imposed the condition $0 < \sigma < c(r-1)^{35}$. Then, *i* changes its strategy through a biased imitation process as what follows²⁴. With the probability *p*, the interlinked neighbor *j* on the opposite layer is selected. With the probability 1 - p, an intralinked neighbor *j* on the same layer is randomly selected. If *i* has no interlink, we choose a neighbor *j* from the same layer regardless of *p*. The strategy of *i* is changed into the strategy of *j* with the transition probability f_{ij} , where

$$f_{ij} = \frac{\exp(\beta P_j)}{\exp(\beta P_j) + \exp(\beta P_i)}.$$
(2)

Here, $\beta(\geq 0)$ controls the amount of noise. When $\beta \rightarrow 0$ *i* randomly adopts the strategy of *j*. However, for $\beta > 0$ we have shown that there exist distinctive states on random network¹⁷. As summarized below the steady-state



Figure 1. Schematic diagram for the construction of a DRN. The first random layer with the size N_1 and the average intradegree $\langle k_{intra} \rangle_1$ and the second random layer with $N_2 (\geq N_1)$ and $\langle k_{intra} \rangle_2$ are separately constructed. To construct random layers we use the Erdös-Rényi (ER) network model⁵⁷ whose degree distribution is known to satisfy the Poisson distribution. Then, for DRN with $n_{inter} (\leq N_1)$ interlinks, $n_{inter} (\leq N_1)$ different nodes are randomly chosen on both layers. Then, n_{inter} links are made, so that one-to-one correspondence between the chosen nodes on the first layer and those on the second layer occurs.

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depends only on $r_0 \equiv \left(rc/\sigma N^{\tanh\left(\frac{\beta c}{2}\right)}\right)$ and $\langle k_{intra} \rangle^{17}$. Therefore, in this paper, we mainly use $\beta = c = \sigma = 1$ for the numerical analyses without loss of generality.

Results on the DRN with $N_1 = N_2 = N$ and $k_{inter} = 1$. Let's first study the SEPGG model on the DRN with $N_1 = N_2 = N$ on which any agent *i* has one interlink or $k_{i,inter} = 1$. We focus the steady-state densities $\{\rho_{\alpha,C}^S(\equiv(\rho_{\alpha,C}(t\to\infty)), \rho_{\alpha,D}^S(\equiv\rho_{\alpha,D}(t\to\infty)), \rho_{\alpha,L}^S(\equiv\rho_{\alpha,L}(t\to\infty))\}$ under the initial condition $\{\rho_{\alpha,C}^I(\equiv\{\rho_{\alpha,C}(t=0)\}=1/3, \rho_{\alpha,D}^I(\equiv\rho_{\alpha,D}(t=0))=1/3, \rho_{\alpha,L}^I(\equiv\rho_{\alpha,L}(t=0))=1/3\}$. Here, $\rho_{\alpha,C}(t)$ means the density of C on the layer $\alpha(=1, 2)$ at the time *t* and etc. All quantities are averaged over 2000 realization of networks. For each network realization strategies are randomly assigned to each agent with given $\{\rho_{\alpha,C}^I, \rho_{\alpha,D}^I, \rho_{\alpha,L}^I\}$.

When p = 0, there exists no coupling between the layers and the steady-states of each layers are the same as those on a single random network, which we have already studied in ref. 17. The followings are brief summary of previous results. Depending on r and $\langle k \rangle_{\alpha} = \langle k_{intra} \rangle_{\alpha} + \langle k_{inter} \rangle_{\alpha}$, the steady-state on each layer for p = 0 becomes one of the following 5 states. For $r_0 = 0.3(<1)$

I) **C-state with** $\{\rho_{\alpha,C}^{S} = 1, \rho_{\alpha,D}^{S} = 0, \rho_{\alpha,L}^{S} = 0\}$ for $\langle k \rangle_{\alpha} < 0.204 \times N^{\nu_{1}}$ with $\nu_{1} \simeq 0.520(2)$.

II) **D-state with**
$$\{\rho_{\alpha,C}^{S} = 0, \rho_{\alpha,D}^{S} = 1, \rho_{\alpha,L}^{S} = 0\}$$
 for $0.204 \times N^{\nu_{1}} < \langle k \rangle_{\alpha} < 0.0247 \times N^{\nu_{2}}$ with $\nu_{2} \simeq 0.898(2)$.

$$(p_{\alpha,C} - 0, p_{\alpha,D} - 0, p_{\alpha,L} - 1) \text{ for } (n_{\alpha} > 0, n_{\alpha})$$

For
$$\beta = c = \sigma = 1$$
 and $r_0 \left| \equiv rc/\sigma N^{\tanh\left[\frac{j\pi}{2}\right]} \right| = 10.0(>1)$

IV) C-state with { $\rho_{\alpha,C}^{S} = 1$, $\rho_{\alpha,D}^{S} = 0$, $\rho_{\alpha,L}^{S} = 0$ } for $\langle k \rangle_{\alpha} < 7.836 \times N^{\nu_{3}}$ with $\nu_{3} \simeq 0.51(1)$. V) D-state with { $\rho_{\alpha,C}^{S} = 0$, $\rho_{\alpha,D}^{S} = 1$, $\rho_{\alpha,L}^{S} = 0$ } for $\langle k \rangle_{\alpha} > 7.836 \times N^{\nu_{3}}$.

Next we consider the case of p = 1 (complete coupling). In complete coupling, any pair of interlinked nodes with identical strategies, i.e., a C-C pair, a D-D pair or an L-L pair, cannot be changed. Any pair with different strategies should change into a pair with identical strategies by the first successful transition. Therefore, the final steady-state is the absorbing state in which any interlinked pair has a common strategy and $\rho_{1,C}^S = \rho_{2,C}^S$, $\rho_{1,D}^S = \rho_{2,D}^S$, $\rho_{1,L}^S = \rho_{2,L}^S$. Considering the final absorbing state under the initial condition, $\rho_{1,C}^I = \rho_{2,C}^I$, $\rho_{1,D}^I = \rho_{2,D}^I$, $\rho_{1,L}^I = \rho_{2,L}^S$, there cannot be anomalous effects to break the density symmetry between layers during evolution. So, we expect $\rho_{1,C}(t) = \rho_{2,C}(t)$, $\rho_{1,D}(t) = \rho_{2,L}(t)$ for p = 1, which is confirmed by simulations on various DRNs. To understand the steady-state behavior for p = 1 in a mean-field level, we study the PGG model on the symmetric DRN of two complete graphs (CGs). The payoff of a node on CG simply depends on ρ_{C} , ρ_D and ρ_L^{17} . When $N \gg 1$, the payoff of a node with D on both CG 1 and 2 is written as $P_D(t) = \frac{rc\rho_{1,C}(t)}{\rho_{1,C}(t) + \rho_{1,D}(t)}$ and the payoff of a node with C is $P_C(t) = P_D(t) - c$. Therefore, the transition probability $f_{(1,D)(2,C)}(t)$ that a D-node on the CG 1 accepts the C-strategy of the interlinked node on the CG 2 is equal to $f_{(2,D)(1,C)}$ and $f_{(1,D)(2,C)} = \frac{exp(\beta P_C(t))}{exp(\beta P_D(t)) + exp(\beta P_D(t))} = 1/[1 + exp(\beta c)](\equiv f_{DC})$. Likewise $f_{CD}(\equiv f_{(1,C)(2,D)} = f_{(2,C)(1,D)}) = \exp(\beta c)/[1 + \exp(\beta c)]$, $f_{LD}(\equiv f_{(1,L)(2,D)} = f_{(2,L)(1,D)}) = 1/[1 + exp(\beta (\sigma - [rc\rho_{1,C}(t)]/(D)])]$



Figure 2. $\rho_{\alpha,C}^{s}, \rho_{\alpha,D}^{s}$ and $\rho_{\alpha,L}^{s}$ on the symmetric DRNs with N = 16,000 for p = 1. (a) Plots of $\rho_{\alpha,C}^{s}, \rho_{\alpha,D}^{s}$ and $\rho_{\alpha,L}^{s}$ on the DRN of two complete graphs against r. The lines represent the mean-field results from equation (3). (b) Plots of $\rho_{\alpha,C}^{s}, \rho_{\alpha,D}^{s}$ and $\rho_{\alpha,L}^{s}$ against $\langle k_{intra} \rangle_{1} (= \langle k_{intra} \rangle_{2})$ of the symmetric DRN for r = 30.0. The horizontal lines represent the mean-field results (c) Plots of $\rho_{\alpha,C}^{s}, \rho_{\alpha,D}^{s}$ and $\rho_{\alpha,L}^{s}$ and mean-field results for r = 2.0. The symbols and lines standing for ρ^{s} s in this figure are used commonly in any plot in this paper.

 $[\rho_{1,C}(t) + \rho_{1,D}(t)])$ and $f_{LC}(\equiv f_{(1,L)(2,C)} = f_{(2,L)(1,C)}) = 1/[1 + \exp(\beta(\sigma - [rc\rho_{1,C}(t)]/[\rho_{1,C}(t) + \rho_{1,D}(t)] + c))]$. Similarly, $f_{DL}(\equiv f_{(1,D)(2,L)} = f_{(1,L)(2,C)} = 1 - f_{LD})$ and $f_{CL}(\equiv f_{(1,C)(2,L)} = f_{(1,L)(2,C)} = 1 - f_{LC})$ are obtained. Since the first successful update changes an interlinked pair of nodes with different strategies (**active pair**) into the pair with the same strategies (**dead pair**), the final absorbing state appears very rapidly. If the active pair is changed into the dead pair in accordance with the initial transition probabilities, then the steady-state for p = 1 on the DRN of two CGs is calculated by

$$\rho_{1,C}^{S}(=\rho_{2,C}^{S}) = \rho_{1,C}^{I}\rho_{2,C}^{I} + (\rho_{1,C}^{I}\rho_{2,D}^{I} + \rho_{1,D}^{I}\rho_{2,C}^{I})f_{DC}^{I} + (\rho_{1,C}^{I}\rho_{2,L}^{I} + \rho_{1,L}^{I}\rho_{2,C}^{I})f_{LC}^{I},
\rho_{1,D}^{S}(=\rho_{2,D}^{S}) = \rho_{1,D}^{I}\rho_{2,D}^{I} + (\rho_{1,D}^{I}\rho_{2,C}^{I} + \rho_{1,C}^{I}\rho_{2,D}^{I})f_{CD}^{I} + (\rho_{1,D}^{I}\rho_{2,L}^{I} + \rho_{1,L}^{I}\rho_{2,D}^{I})f_{LD}^{I},
\rho_{1,L}^{S}(=\rho_{2,L}^{S}) = \rho_{1,L}^{I}\rho_{2,L}^{I} + (\rho_{1,L}^{I}\rho_{2,C}^{I} + \rho_{1,C}^{I}\rho_{2,L}^{I})f_{CL}^{I} + (\rho_{1,L}^{I}\rho_{2,D}^{I} + \rho_{1,D}^{I}\rho_{2,L}^{I})f_{DL}^{I}.$$
(3)

As shown in Fig. 2(a), the mean-field equation (3) explains the simulation results on the DRN of two CGs for any *r* very well. When $r \gg 1$, $f_{LC}^I \simeq f_{LD}^I \simeq 1$ and $f_{CL}^I \simeq f_{DL}^I \simeq 0$ for $\sigma \ll r$. Using $f_{CD}^I = 1 - f_{DC}^I =$ $1/[1 + \exp(\beta c)] = 1/[1 + e]$ for $\beta = 1$ and c = 1 and the initial condition $\rho_{\alpha,C}^I = \rho_{\alpha,D}^I = \rho_{\alpha,L}^{I} = 1/3$, we get $\rho_{\alpha,C}^S = 0.393$ and $\rho_{\alpha,D}^S = 0.496$ and $\rho_{\alpha,L}^S = 1/9$ from equation (3), which explain the data for r < 20 in Fig. 2(a). When r < 20, f_{LC}^I , f_{LD}^I , f_{CL}^I and f_{DL}^I show nontrivial behavior. The equation (3) also predicts that $\rho_{\alpha,L}^S$ gets larger and $\rho_{\alpha,D}^S$ as well as $\rho_{\alpha,C}^S$ gets smaller as *r* decreases for r < 20.

and $\rho_{\alpha,D}^{S}$ as well as $\rho_{\alpha,C}^{S}$ gets smaller as *r* decreases for r < 20. Next, we study how the steady-state on the DRN with general $\langle k_{intra} \rangle_{\alpha}$'s for p = 1 varies from the mean-field result. In Fig. 2(b,c), the simulation results on symmetric DRNs are shown. The result for large *r* (or r = 30) in Fig. 2(b) deviates from the mean-field result. The deviation becomes much more enhanced as $\langle k_{intra} \rangle_{\alpha}$ gets smaller. Furthermore, $\rho_{\alpha,C}^{S}(\rho_{\alpha,D}^{S})$ also gets larger (smaller) compared to the mean-field result. $\rho_{\alpha,L}^{S}$ is nearly the same as the mean-field execution for the very small $\langle k_{intra} \rangle_{\alpha}$. The result for small *r* (or r = 2) in Fig. 2(c) is nearly the same as the mean-field result for small $\langle k_{intra} \rangle_{\alpha}$. For small $\langle k_{intra} \rangle_{\alpha}, \rho_{\alpha,C}^{S}$ is slightly larger than the mean-field result and both $\rho_{\alpha,D}^{S}$ are slightly smaller than the mean-field results.

In Fig. 3, the simulation results for p = 1 on asymmetric DRNs are shown. When p = 1 the interlinked pairs of the same strategies make dead pairs as in the symmetric DRN, the ρ_{α}^{Ss} on both layers are the same when $N_1 = N_2$. As shown in Fig. 3, for a given $\langle k_{intra} \rangle_1$, ρ_{α}^{Ss} are nearly constant regardless of $\langle k_{intra} \rangle_2$ except for $\langle k_{intra} \rangle_2 \lesssim \langle k_{intra} \rangle_1$.



Figure 3. $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ on the asymmetric DRNs with N = 16000 for p = 1. (a) Plots of $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ against $\langle k_{intra} \rangle_2$ and mean-field results for r = 30.0 and $\langle k_{intra} \rangle_1 = 4$. (b) Plots of $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ against $\langle k_{intra} \rangle_2$ and mean-field result for r = 30.0 and $\langle k_{intra} \rangle_1 = 20$. (c) Plots of $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ against $\langle k_{intra} \rangle_2$ and mean-field result for r = 2.0 and $\langle k_{intra} \rangle_1 = 4$. (d) Plots of $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ against $\langle k_{intra} \rangle_2$ and mean-field result for r = 2.0 and $\langle k_{intra} \rangle_1 = 4$. (d) Plots of $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ against $\langle k_{intra} \rangle_2$ and mean-field result for r = 2.0 and $\langle k_{intra} \rangle_1 = 4$. (d) Plots of $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ against $\langle k_{intra} \rangle_2$ and mean-field result for r = 2.0 and $\langle k_{intra} \rangle_1 = 4$. (d) Plots of $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ against $\langle k_{intra} \rangle_2$ and mean-field result for r = 2.0 and $\langle k_{intra} \rangle_1 = 20$.

Thus, the steady-state ρ_{α}^{S} 's on the asymmetric DRNs depends mainly on $\langle k_{intra} \rangle_1$ and independent of $\langle k_{intra} \rangle_2$ except for the symmetric range $\langle k_{intra} \rangle_2 \sim \langle k_{intra} \rangle_1$. For large r (r=30) both $\rho_{\alpha,C}^{S}$ and $\rho_{\alpha,L}^{S}$ on the asymmetric DRN with the small $\langle k_{intra} \rangle_1$ (Fig. 3(a)) is larger than that on the DRN with the large $\langle k_{intra} \rangle_1$ (Fig. 3(b)), while $\rho_{\alpha,D}^{S}$ for the small $\langle k_{intra} \rangle_1$ is smaller. For small r (r=2) as shown in Fig. 3(c,d), the dependences of $\rho_{\alpha,C}^{S}$ and $\rho_{\alpha,D}^{S}$ on the asymmetric DRN on $\langle k_{intra} \rangle_1$ are the same as those for large r. In contrast, $\rho_{\alpha,L}^{S}$ on the asymmetric DRN is nearly constant of the mean-field values regardless of $\langle k_{intra} \rangle_1$. When r gets larger in the interval $r \gtrsim 20$, $\rho_{\alpha,C}^{S}$ ($\rho_{\alpha,D}^{S}$) on the asymmetric DRN becomes slightly larger (smaller) than that for mean-field expectation, but $\rho_{\alpha,L}^{S}$ hardly varies.

Now we want to explain the theoretical origins of the results for p = 1 in Figs 2 and 3. As explained when deriving the mean-field result, we use the fact that the final absorbing state for p = 1 appears very rapidly. Thus, the initial configuration is very important to decide the final absorbing state. Topologically localized cluster of a certain strategy on CG is impossible to form if there exist Cs, Ds and Ls simultaneously. In contrast, on the layer with relatively small $\langle k_{intra} \rangle_{\alpha}$, it is quite easy to form localized clusters of the same strategies due to fluctuation of the distribution of Cs, Ds and Ls. Such localized clusters reinforces the network reciprocity^{16-23,25}. As $\langle k_{intra} \rangle_{\alpha}$ decreases, it becomes much easier to form localized clusters. If a node i of a C-cluster on one layer is interlinked to a node *j* of a D-cluster on the opposite layer, then the node *j* is changed into a C-node when $P_i > P_j$ or $r > (n_{i,C}+1)(n_{i,D}+1)/[n_{i,C}(n_{j,D}+1) - (n_{i,C}+1)]$ for p = 1. Since in RN $n_{i,C}$ and $n_{i,D}$ is roughly comparable with $\langle k_{intra} \rangle_{\alpha}$, small $\langle k_{intra} \rangle_{\alpha}$ enhances the fluctuation. Thus, large r and small $\langle k_{intra} \rangle_{\alpha}$ make more D-nodes in D-cluster be changed into C-nodes compared to the mean-field expectation. This enhances the network reciprocity as shown in Fig. 2(b,c). If a node of a C-cluster on one layer is interlinked to a node of an L-cluster on the opposite layer, the C-node wins over the L-node for large r, but the L-node wins over the C-node for small r. Therefore, the fluctuation effect makes more C and suppresses D compared to the mean-field result and this effect becomes more enhanced as $\langle k_{intra} \rangle_{\alpha}$ decreases. In addition, it enhances L for small r and suppresses L for large r. Thus, on the symmetric DRN, $\rho_{\alpha,C}^{S}$ increases and $\rho_{\alpha,D}^{S}$ decreases as $\langle k_{intra} \rangle_{\alpha}$ for large *r* as in Fig. 2(b). For small *r*, D-node is relatively hard to change its strategy into C. As a result the fluctuation effect is suppressed a little bit compared with large r and appears only on the DRN with small $\langle k_{intra} \rangle_{\alpha}$ as in Fig. 2(c). On the asymmetric DRN, the C-cluster is



Figure 4. $\rho_{\alpha,C}^{S}$, $\rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ on the asymmetric DRN with $\langle k_{intra} \rangle_1 = 10$ and $\langle k_{intra} \rangle_2 = 40$ $r_0 = 0.3$ and N = 16000 are used. (a) Plots of $\rho_{1,C}^{S}$, $\rho_{1,D}^{S}$ and $\rho_{1,L}^{S}$ against p (Left) and the same plots of $\rho_{2,C}^{S}$, $\rho_{2,D}^{S}$ and $\rho_{2,L}^{S}$ (Right) on the DRN with $\langle k_{intra} \rangle_1 = 10$ and $\langle k_{intra} \rangle_2 = 40$ for $r_0 = 0.3$. (b) Time dependences of $\rho_{\alpha,C}(t)$, $\rho_{\alpha,D}(t)$ and $\rho_{\alpha,L}(t)$ on t for p = 0.1. (c) The same plots as (b) for p = 0.6. (d) The same plots as (b) for p = 0.999. t is the Monte-Carlo time.

more easily formed on the layer with smaller $\langle k_{intra} \rangle_{\alpha}$ (or the layer 1), which makes the D-nodes in the opposite layer change their strategy into C. Thus the steady-state is decided by the layer 1. This effect becomes more enhanced as $\langle k_{intra} \rangle_1$ decreases. This fluctuation effect on the asymmetric DRN explains the behaviors of $\rho_{\alpha,C}^S$ and $\rho_{\alpha,C}^S$ in Fig. 3. The fluctuation effect on $\rho_{\alpha,C}^S$ is relatively weaker as in Fig. 3.

 $\rho_{\alpha,L}^{S}$ in Fig. 3. The fluctuation effect on $\rho_{\alpha,L}^{S}$ is relatively weaker as in Fig. 3. We now explain the results for the steady-state on the asymmetric DRN for $0 , which show novel network reciprocity. On the asymmetric DRN with <math>\langle k_{intra} \rangle_1$ and $\langle k_{intra} \rangle_2 (> \langle k_{intra} \rangle_1)$, the steady-state should mainly depend on $\langle k_{intra} \rangle_1$ as expected from the result for p = 1. To see this effect on the asymmetric DRN we first set $\langle k_{intra} \rangle_1 = 10$, which is small enough that the steady-state on the layer 1 is the C-state when p = 0 (i.e., the state I and IV). If $\langle k_{intra} \rangle_2$ is also small enough, the state II) appears on the layer 2 for p = 0.

We first carry out the simulations on the asymmetric DRN with $\langle k_{intra} \rangle_2 = 40$ which is moderately large intradegree for $r_0 = 0.3(<1)$. As shown in Fig. 4(a), { $\rho_{\alpha,C}^S, \rho_{\alpha,D}^S, \rho_{\alpha,L}^S$ } for p = 0 or p = 1 reproduce the previously explained corresponding results very well. As p increases, $\rho_{1,C}^S$ first decreases slowly from $\rho_{1,C}^S(p = 0) = 1$ to $\rho_{1,C}^S(p \simeq 0.6) = 0.9$ and increases slowly to $\rho_{1,C}^S(p = 0.97) = 0.93$. Furthermore, $\rho_{1,D}^S < 0.1$ and $\rho_{1,L}^S \simeq 0$ for $0 . In contrast, <math>\rho_{2,C}^S$ rapidly increases as p increases from 0 and reaches 0.8 for p = 0.1 and increases relatively and increases relatively of $\rho_{1,C}^S(p = 0.97) = 0.93$. tively slowly until $\rho_{2,C}^{s}(p=0.97) = 0.93$. The C-dominant state on the layer 1 exists for any p(<1) as expected from the C-state(state I) for p = 0. The steady-state on the layer 2 for p > 0 is rather surprising, because $\rho_{2,C}^{s}$ becomes very large for even very small p. This indicates that the inter-layer coupling drastically enhances the network reciprocity. To see the origin of this result, we study $\rho_{\alpha,C}(t)$, $\rho_{\alpha,D}(t)$, $\rho_{\alpha,L}(t)$. For small p(=0.1), $\rho_{1,C}(t)$ $(\rho_{1,D}(t))$ increases (decreases) rapidly at early time *t* as shown in Fig. 4(b). Due to the weak inter-layer coupling, $\rho_{1,C}(t)$ increases to $\rho_{1,C}^{S} \simeq 1$ when $20 \lesssim t$ as the case p = 0. Since the layer 2 would be in D-state when p = 0 as shown in Fig. 4(a), $\rho_{2,D}^{1,C}(t)$ increases and is slightly larger than $\rho_{2,C}(t)$ at small t. Through the inter-layer coupling, D's on layer 2 can easily change their strategy into C due to the larger payoff of C on the layer 1 when $\rho_{1,C}(t)$ becomes large enough. As a result $\rho_{2,C}(t)$ increases to $\rho_{2,C}^{S} \simeq 0.8$ when $20 \leq t \leq 30$ both $\rho_{2,C}(t)$ and $\rho_{2,D}(t)$ follow $\rho_{1,C}(t)$ and $\rho_{1,D}(t)$. Thus, for small p, the C-dominance on the layer 1, which is fully developed due to weak inter-layer coupling, induces the C-dominance on the layer 2 at later time. For moderate p(=0.6), the dependences of $\rho_{1,C}(t)$ and $\rho_{1,D}(t)$ on t are much more similar to those of $\rho_{2,C}(t)$ and $\rho_{2,D}(t)$ as shown in Fig. 4(c). For moderate *p*, the effects from the inter-layer coupling are nearly equal to the effect of the intra-layer interactions. Due to the increased inter-layer coupling, $\rho_{2,C}(t)$ increases nearly synchronously with $\rho_{1,C}(t)$. Such rapid increase of $\rho_{2,C}(t)$ makes remnant D's on the layer 2. Those remnant D's on the layer 2 get relatively high payoff from the intra-layer interactions with relatively dense C's induced by inter-layer coupling. Then, through the inter-layer coupling, $\rho_{1,C}^{S}$ slightly decreases by remnant D's on the layer 2 for $0.2 \le p \le 0.6$. For large p(>0.6), $\rho_{1,C}^{S}$ increases as p increases. In this case, $\rho_{\alpha}(t)$'s first reach ρ_{α}^{S} for p = 1 very rapidly as shown in Fig. 4(d). Then, through the intra-layer interactions, $\rho_{1,C}(t)$ increases very slowly from $\rho_{1,C}^{S}(p=1)$, so that $\rho_{1,C}^{S}$ increases as p increases for p > 0.6. In contrast to the subtle dependence of $\rho_{1,C}^{S}$ on p, $\rho_{2,C}^{C}$ monotonically increases as p increases and $\rho_{2,C}^{S}(p=0.97)(=\rho_{1,C}^{S}(p=0.97)) = 0.93$. The behavior of $\rho_{\alpha,D}^{S}$ is easily understood from the behavior of $\rho_{\alpha,C}^{S}$ in Fig. 4(a-d). $\rho_{\alpha,L}^{S}$ is merely equal to 0, which can be understood from the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,C}^{S}$ in the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. These mechanisms explain the dependence of $\rho_{\alpha,L}^{S}$ is the result p = 0. dependence of steady-state on p in Fig. 4(a) rather well.



Figure 5. $\rho_{\alpha,C}^{s}$, $\rho_{\alpha,D}^{s}$ and $\rho_{\alpha,L}^{s}$ on the asymmetric DRN. N = 16000 are used. (a) Plots of $\rho_{1,C}^{s}$, $\rho_{1,D}^{s}$ and $\rho_{1,L}^{s}$ against p (Left) and the same plots of $\rho_{2,C}^{s}$, $\rho_{2,D}^{s}$ and $\rho_{2,L}^{s}$ (Right) on the DRN with $\langle k_{intra} \rangle_{1} = 10$ and $\langle k_{intra} \rangle_{2} = 500$ for $r_{0} = 0.3$. (b) The same plots as (a) on the asymmetric DRN against p with $\langle k_{intra} \rangle_{1} = 100$ and with $\langle k_{intra} \rangle_{2} = 2000$ for $r_{0} = 10.0$.

Next, we study the inter-layer coupling effects on the asymmetric DRN with $\langle k_{intra} \rangle_1 = 10$ and $\langle k_{intra} \rangle_2 = 500$ for $r_0 = 0.3(<1)$. For p = 0, the steady-state on the layer 1 is the C-state (the state I) and the steady-state on the layer 2 is the L-state (the state III) as shown in Fig. 5(a). As p increase, $\rho_{1,C}^S = 1(\rho_{1,D}^S = 0)$ sustains for p < 0.2 and $\rho_{1,C}^S$ ($\rho_{1,D}^S$) simply decreases (increases) to $\rho_{1,C}^S(p = 1)(\rho_{1,D}^S(p = 1))$. $\rho_{1,L}^S = 0$ except for p = 1. In contrast, $\rho_{2,C}^S$ increases very rapidly for small p and reaches the maximum at p = 0.34, $\rho_{2,C}^S(p = 0.34) \simeq 0.76$. Then, $\rho_{2,D}^S$ simply decreases for p > 0.34 to $\rho_{2,D}^S(p = 1)$. $\rho_{2,D}^S$ shows rather complex behavior. For very small p(<0.05), $\rho_{2,D}^S$ rapidly increases to the maximum, $\rho_{2,D}^S(p = 0.05) = 0.9$, and decreases for $0.05 to the minimum, <math>\rho_{2,D}^S(p = 0.34) = 0.23$. Then, $\rho_{2,D}^S$ monotonically increases for p > 0.34 to $\rho_{2,D}^S(p = 1)$. $\rho_{2,D}$ show nearly the same behavior as those in Fig. 4(b). For larger $\langle k_{intra} \rangle_2$ it is more difficult to make C's through intra-layer interaction and more inter-layer coupling or large p is needed to increase $\rho_{2,C}^S$. The time-dependences as in Fig. 4(b) do not occur for the larger p(>0.34) on the DRN with $\langle k_{intra} \rangle_1 = 10$ and $\langle k_{intra} \rangle_2 = 500$. As p increases further (p > 0.34), $\rho_{\alpha}(t)$'s behave nearly the same as in Fig. 4(c), which decreases (increases) ρ_{C}^S 's for larger p approach to $\rho_{\alpha}^S(p = 1)$'s smoothly. These mechanisms explain the dependence of steady-state on p in Fig. 5(a) rather well.

We also study the SEPGG for $r_0 = 10.0(>1)$ on the asymmetric DRN with $\langle k_{intra} \rangle_1 = 100$ and $\langle k_{intra} \rangle_2 = 2000$. As shown in Fig. 5(b), the steady-state for p = 0 on the layer 1 is the C-state (state IV) and that on the layer 2 is the D-state (state V). For p > 0, $\rho_{1,C}^S = \rho_{2,C}^S \simeq 1$ except for very small p. Thus, on this DRN, $\rho_{\alpha}(t)$'s show nearly the same behavior as those in Fig. 4(b) for $p \neq 0$ due to the very large r. Furthermore, the payoff of D on layer 2 is less than that of C's in layer 1 in general. Thus, the D's in the layer 2 easily change into C's through the inter-layer coupling as p increases, which makes $\rho_{2,D}^S(p)$ approaches to zero for p > 0.1.

In Fig. 6 we display the behavior of $\rho_{\alpha}^{S}(p)$'s for $r_{0} = 0.3(<1)$ on the asymmetric DRN with $\langle k_{intra} \rangle_{1} = 40$ and $\langle k_{intra} \rangle_{2} = 500$. As shown in Fig. 6, the steady-state for p = 0 on the layer 1 is the D-state (state II) and that on the layer 2 is the L-state (state III). Except for $p \simeq 0.5$, $\rho_{1,D}^{S} = 1$ and $\rho_{2,D}^{S} = 1$. But for $p \simeq 0.5$, both $\rho_{1,C}^{S}$ and $\rho_{2,C}^{S}$ show anomalous behavior. $\rho_{1,C}^{S}$ and $\rho_{2,C}^{S}$ become maxima at $p \simeq 0.43$ as $\rho_{1,C}^{S}(p = 0.43) = 0.53$ and $\rho_{2,C}^{S}(p = 0.43) = 0.39$. For smaller p(<0.4) and larger p(>0.6), the intra-layer interactions make $\rho_{1,D}^{S} = 1$ and the inter-layer couplings make $\rho_{2,D}^{S} = 1$. In contrast, the delicate anomalous behavior around $p \simeq 0.5$ is confirmed to



Figure 6. $\rho_{\alpha,C}^{s}$, $\rho_{\alpha,D}^{s}$ and $\rho_{\alpha,L}^{s}$ on the asymmetric DRN with $\langle k_{intra} \rangle_1 = 40$ and $\langle k_{intra} \rangle_2 = 500$. $r_0 = 0.3$ and N = 16000 are used. Plots of $\rho_{1,C}^{s}$, $\rho_{1,D}^{s}$ and $\rho_{1,L}^{s}$ against p (Left) and the same plots of $\rho_{2,C}^{s}$, $\rho_{2,D}^{s}$ and $\rho_{2,L}^{s}$ (Right) on the DRN with $\langle k_{intra} \rangle_1 = 40$ and $\langle k_{intra} \rangle_2 = 500$.

be originated from the balance between the inter-layer processes and the intra-layer processes, because for $p \simeq 0.5$, the intra-layer processes happen at nearly the same rate as the inter-layer processes. We confirm the following cyclic process I)-II)-III)-IV)-I) by simulations for $p \simeq 0.5$. I) The intra-layer interactions on the layer 1 make C-dense regions. II) This C-dense regions induce C's on the layer 2 through the inter-layer coupling. III) The C's on the layer 2 make D's by the intra-layer interactions. IV) The D's on the layer 2 shrink the C-dense regions through the inter-layer coupling. This cyclic process makes non-zero $\rho_{\alpha,C}^S$ and $\rho_{\alpha,D}^S$ around $p \simeq 0.5$. This anomalous effect also makes to appear the network reciprocity that $\rho_{\alpha,C}^S > 0$ from the inter-layer coupling, without which there cannot exist any C on both layers.

We also study the model for 0 on the symmetric DRN. We check for various*p*and*r* $and find that the steady-state densities on the symmetric DRN show exactly the same behavior as those on the single random network. We found that the steady-state on one layer inevitably is the same as that on the other layer. From the comparison of <math>\rho_{\alpha}^{S_s}$ for various *p* and *r* to confirm that the steady-state on the symmetric DRN for 0 is exactly the same as on the corresponding single network, and find that*p*only makes the time-delay (see Supplementary Information).

Results on the DRN with $\langle k_{intra} \rangle < 1$ or $N_1 \neq N_2$. We also study the model on the DRN with $N_1 = N_2 = N$ and $\langle k_{intra} \rangle < 1$. In Fig. 7(a), the results on the asymmetric DRN with $\langle k_{intra} \rangle_1 = 10$, $\langle k_{intra} \rangle_2 = 40$ and $\langle k_{intra} \rangle = 0.5$ for $r_0 = 0.3(<1)$ are shown. Comparing Fig. 7(a) to Fig. 4(a), the dependences of $\rho_{\alpha}^{S_3}$ so p(<1) for $\langle k_{intra} \rangle = 0.5$ are nearly the same as those up to p = 0.5 for $k_{inter} = 1$. In contrast $\rho_{\alpha}^{S_3}$ sit p = 1 for $\langle k_{intra} \rangle = 0.5$ show the nontrivial behaviors, because $N\langle k_{intra} \rangle = 0.5N$ interlinked pairs of nodes become dead pairs in the steady-state. On the DRN with $\langle k_{intra} \rangle < 1$ the dependences of $\rho_{\alpha}^{S_3}$ so p(<1) are generally confirmed to be nearly the same as those up to $p = \langle k_{intra} \rangle$ except at p = 1.

To know the effects of the difference between two layer sizes, we study the model on the DRN with $N_1 \neq N_2$. In Fig. 7(b), the results on the DRN with $N_1 = 16000$, $\langle k_{intra} \rangle_1 = 10$ and $N_2 = 32000$, $\langle k_{intra} \rangle_2 = 57$ for $r_0 = 0.3(<1)$ are shown. At p = 0, the steady-state of the layer 1 is the C-state (State I) and the steady-state of the layer 2 is the D-state (State II). As p increase, $\rho_{1,C}^S = 1(\rho_{1,D}^S = 0)$ sustains for p < 0.05 and $\rho_{1,C}^S(\rho_{1,D}^S)$ simply decreases (increases) to $\rho_{1,C}^S(p = 0.97) = 0.52$ ($\rho_{1,D}^S(p = 0.97) = 0.48$). $\rho_{1,L}^S = 0$ except for p = 1. In contrast, $\rho_{2,C}^S$ increases very rapidly for small $p(\leq 0.1)$ and reaches the maximum at p = 0.19, $\rho_{2,C}^S(p = 0.97) \simeq 0.51$. $\rho_{2,D}^S$ decreases very rapidly for small $p(\leq 0.1)$ and reaches the maximum at p = 0.19, $\rho_{2,C}^S(p = 0.97) \simeq 0.51$. $\rho_{2,D}^S$ decreases slowly for p > 0.19 to $\rho_{2,C}^S(p = 0.97) \simeq 0.51$. $\rho_{2,D}^S$ decreases very rapidly for small $p(\leq 0.1)$ and reaches the maximum at p = 0.19, $\rho_{2,D}^S(p = 0.97) \simeq 0.54$. Then, $\rho_{2,D}^S$ decreases slowly for p > 0.19 to $\rho_{2,D}^S(p = 0.97) \simeq 0.54$. Then, $\rho_{2,D}^S$ decreases slowly for p > 0.19 to $\rho_{2,D}^S(p = 0.97) \simeq 0.44$. Then, $\rho_{2,D}^S$ increases slowly for p > 0.19 to $\rho_{2,D}^S(p = 0.97) \simeq 0.44$. Then, $\rho_{2,D}^S$ increases slowly for $N_1 = N_2$. The difference between the results in Fig. 7(b) and those in Fig. 5(a) are originated from the enhanced intra-layer interactions on the layer 2, the there are more D's than to the case of $N_1 = N_2$ and ρ_D^S 's (ρ_S^S 's) on both layer increase (compared to those in Fig. 5(a). We also study the model on the DRN with $N_1 = 16000$, $\langle k_{intra} \rangle_1 = 40$ and $N_2 = 32000$, $\langle k_{intra} \rangle_2 = 14$ and find nearly identical results to those in Fig. 5(b). Because $N_1 < N_2$, the enhanced intra-layer interactions on the layer 2 induce the strong network reciprocity. As a result the C-state appears on both layers. We generally confirm that the intra

Discussion

In summary, we study the SEPGG on DRN. When two CG's of the same size interact through the inter-coupling with $k_{inter} = 1$ and p = 1, the steady-state density, ρ_{α}^{S} of each strategy on each layer α can be exactly described by the mean-field theory. If the $\langle k_{intra} \rangle_{\alpha}$ decreases then ρ_{α}^{S} on each layer slightly deviates from the mean-field expectation. Such deviation is relatively small when the multiplication factor r is small. While if $\langle k_{intra} \rangle_1 < \langle k_{intra} \rangle_2$



Figure 7. $\rho_{\alpha,C}^{S} \rho_{\alpha,D}^{S}$ and $\rho_{\alpha,L}^{S}$ on the asymmetric DRN with $\langle k_{intra} \rangle < 1$ or $N_1 \neq N_2$. $r_0 = 0.3$ is used. (a) Plots of $\rho_{1,C}^{S} \rho_{1,D}^{S}$ and $\rho_{1,L}^{S}$ against p (Left) and the same plots of $\rho_{2,C}^{S}, \rho_{2,D}^{S}$ and $\rho_{2,L}^{S}$ (Right) on the DRN with $\langle k_{intra} \rangle_{1} = 10$ and $\langle k_{intra} \rangle_{2} = 40$ for $\langle k_{intra} \rangle = 0.5$ and $N_1 = N_2 = N = 16000$. (b) The same plots as (a) on the asymmetric DRN against p with $N_1 = 16000$, $\langle k_{intra} \rangle_{1} = 10$ and with $N_2 = 32000$, $\langle k_{intra} \rangle_{2} = 57$ for $n_{inter} = N_1$.

(asymmetric DRN), then $\rho_{\alpha}^{S_3}$ are determined by $\rho_{\alpha}^{S_3}$ so the layer 1 and the network reciprocity can be reinforced through the inter-layer coupling for $0 . For <math>0 on the symmetric DRN, <math>\rho_{\alpha}^{S_3}$ show exactly the same behavior with those on the corresponding single network, and *p* only makes the time-delay (see Supplementary Information). The schematic diagrams of the non-vanishing $\rho_{\alpha}^{S_3}$ are also provided in Supplementary Information. Furthermore, we also investigate the behavior of $\rho_{\alpha}^{S_3}$ so the DRN with $\langle k_{intra} \rangle < 1$ or $N_1 \neq N_2$. On the DRN with $\langle k_{intra} \rangle < 1$, we find that $\rho_{\alpha}^{S_3}$ are nearly the same as those for $\langle k_{intra} \rangle = 1$ with $p = \langle k_{intra} \rangle$. Finally, if $N_1 < N_2$ then the steady-state density is determined by the state of layer 2, thus the network reciprocity of the entire network can be enhanced when $\langle k_{intra} \rangle_2$ is small enough.

Furthermore, since each individual in real world interacts to each other through several different channels of interactions, such interaction topology sometimes can be well described by the multiplex networks in which all the layers have the same set of nodes in general. The DRN with $N_1 = N_2$ and $k_{inter} = 1$ could be related to a kind of multiplex networks. In addition, DRN with $N_1 \neq N_2$ or $k_{inter} < 1$ corresponds to the interconnected networks in which each layer has different set of nodes. Therefore, our SEPGG on DRN model would provide more general framework to study the emergence of cooperation in more realistic systems.

Finally, we want to make some remarks on important open questions. Although we do not assume any detailed topological properties, many studies have revealed that some topological properties of a network such as degree heterogeneity^{22,49–52}, degree-degree correlation⁵³, and clustering coefficient⁵⁴ can significantly change the evolution of cooperation. The cost heterogeneity is also known to play a nontrivial role in the emergence of cooperation²². As β changes many interesting phenomena related to the phase transition has been reported^{55,56}. Thus it would be very important to study how such topological properties, cost heterogeneity, and noise level affect the evolution of cooperation in interlinked networks.

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Author Contributions

J.K., S.-H.Y. and Y.K. designed study. J.K. performed the simulations and analyzed data. Y.K. and S.-H.Y. wrote the manuscripts. All authors revised the manuscript.

Additional Information

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