

1 **A game theory model to explore the role of cooperation and diversity**
2 **in community food security: the case of Southern Malawi.**

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18
19 **Abstract**

20 The Sustainable Development Goals aim at ending food insecurity by 2030. Therefore,
21 civil society needs to understand the inherent complexities of both socio-economic and
22 ecological dynamics and their interdependencies. In particular, the behavioural
23 dynamics that underpin human agents are crucial in driving the final outcomes in terms
24 of community food security and require further attention. Using household behaviour
25 within a rural village of Southern Malawi as an example, we describe a game theory
26 model representing cropping strategies: (1) cooperation, as driven by other-regarding
27 preferences, and (2) conformation, the tendency to converge to similar crop planting
28 choices as opposed to differentiation (and thus crop diversity). We find that the latter
29 plays a crucial role in driving the system towards successful strategies: how individuals
30 relate to social norms has greater effect. Cooperation is only necessary for community
31 success when the community converges on crop planting choices. On the contrary,
32 differentiation, the affirmation of the individual unique identity, can succeed with or
33 without cooperation. We further elaborate on the idea that community level sustainability
34 can be reached through different pathways, which might require food exchange
35 mechanisms within and beyond the system boundaries.

36
37 **Keywords**

38 human behaviour; game theory, social networks, diversity, social-ecological systems,
39 food security.

40 **Length of Manuscript:** 6100 words and 5 figures and 1 table.

11. Introduction

2

3 Food provisioning is a key challenge of coupled human-natural systems. The
4 Sustainable Development Goals (SDGs) adopted by the United Nations have set a clear
5 target to end food insecurity by 2030 (UN 2015a). In order to achieve this target
6 understanding the inherent complexities of both socio-economic and ecological
7 dynamics, and their mutual feedbacks across scales is crucial (Ostrom 2008). While
8 agro-ecological dynamics are well explored in the literature (e.g. Chappell and LaValle
9 2011), the behavioural dynamics that underpin human agents require more research to
10 understand how they drive community food security, especially in rural regions of the
11 Global South.

12

13 Using household crop planting choices within a rural village as an example, we
14 introduce a behavioural compass based on two dimensions: (1) cooperation, the degree
15 to which individual success depends on neighbours' success due to other-regarding
16 preferences, which in turn captures a community scale objective function, and (2)
17 conformation, the tendency to converge towards similar crop planting choices as
18 opposed to differentiation, and as a consequence, crop diversity. We argue that by
19 acknowledging the interplay among these two dimensions in shaping household
20 cropping strategy, a rural agricultural system can better reorganize, spontaneously or by
21 means of policy interventions, to improve food security at the community level.

22

23 In the following paragraphs we explain the compass axes, used as a reference system
24 to characterize household behaviour and show how the compass applies to the selected
25 case study: a rural village of Southern Malawi. In Methods, we describe the context and
26 the data used for the application and the mathematical formalization of the model. We
27 also describe its behaviour as we change the main household parameters (other-
28 regarding preferences and homophily) within their interval of reference. In Results, we
29 assess the model performance when the game parameters (number of households,
30 number of crops and network topology) are set to match the case study characteristics,
31 deriving the household behaviour that delivers the best outcomes. In Discussion, we
32 elaborate on the implications of the model findings for the community food security.
33 Finally, we conclude with the key messages of this study.

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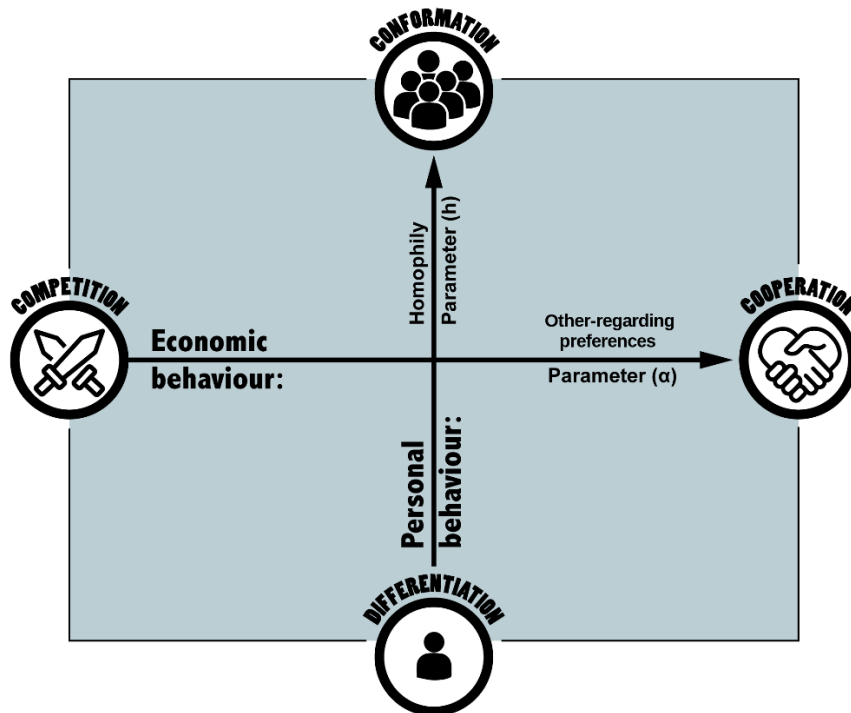
35 Inspired by the political compass of Maddox and Lilie (1984) and Lester (1994), we refer
36 to a behavioural compass (Fig. 1), a four by four grid where the horizontal axis
37 represents economic behaviour and displays competition at its left side and full
38 cooperation at its right side, while the vertical axis represents personal behaviour and
39 displays differentiation at the bottom end and conformation at its top end.

1 Economic behaviour describes how the individual relates to the others for the purpose
2 of achieving resources, including consumption and utilisation of natural resources.
3 Cooperation can be understood as the process of acting in coordination with others to
4 maximise individual or mutual benefit (Perc et al. 2017). Competition arises when at
5 least two parties strive for a goal that cannot be shared, or is desired individually but not
6 in sharing.

7 Personal behaviour describes how individual beliefs and values relate to social norms.
8 Conformation is the process of matching individual behaviour with the dominant social
9 norms, while differentiation—as in psychological differentiation (Witkin et al. 1974)—is
10 the process of affirming an individual unique identity against social norms. Thus, the
11 higher differentiation, the higher the diversity of behaviours within the system.

12
13 In an early example of a similar approach, Deutsch (1949) identifies three types of “goal
14 structures on achievement”: cooperative, competitive and individualistic. Cooperation
15 and competition have been extensively researched in ecology and economics (game
16 theory, in particular: see Fehr and Fischbacher 2002) as individual operating modes
17 (from the seminal work of Axelrod and Hamilton 1981 to more recent reviews, such as
18 Perc et al. 2017). Differentiation and conformation belong to the domain of behavioural
19 and experimental methods in psychology and economics and, in the context of this
20 study, are particularly close to the research on how social norms and the framing of
21 options influence individual behaviour, with significant repercussions for human
22 cooperation and natural resource exploitation (Kahneman 2011).

23



24
25

Fig. 1 The behavioural compass

12. Methods

2

3 In this article, we refer to community food security as an emergent property of
4 household food security, which is a consequence of how households interact, acquire
5 and utilise assets, including natural resources (Chambers and Conway 1991). In most
6 cases, the livelihoods of rural households remain largely dependent on agriculture. This
7 article explores data of a rural region located in the southern part of Malawi, a small,
8 landlocked African country home to approximately 15 million people. More than 90% of
9 the rural population are smallholder farmers, responsible for cultivating plots with an
10 average size of just 0.8 hectares (NSO 2012). Maize is the socially preferred crop
11 among smallholders and the main staple diet of the population (NSO 2012). According
12 to Chinsinga and Chasukwa (2012), although over 97% of smallholder farmers grow
13 maize, only 10% are net sellers and up to 60% are net buyers.

14 Typically, most of the households within a rural village are at least acquainted with each
15 other and it is common practice to exchange on-farm labour and food donations in
16 periods of need (Dobbie et al. 2018). Similar dynamics have been witnessed in rural
17 settings around the globe (Patel 2009) including: sharing of unused ingredients that
18 would otherwise be trashed, sharing of knowledge about farming practices, good
19 nutrition and recipes, invitations to join meals, exchange of seeds and food commodities
20 to increase diversity of diet, exchange of portions of meals in different periods in time to
21 cushion temporary scarcity (i.e. food banking). In Malawi, this system of informal social
22 exchanges coexists with a market-oriented structure that heavily relies on local and
23 regional food markets for the integration and redistribution of food commodities, as a
24 consequence of the liberalisation reforms promoted by the International Monetary Fund
25 and the World Bank (Dorward and Kydd 2004).

26

27 **2.1 The case of food security in Southern Malawi**

28 In Malawi over 50% of the population live on less than one US dollar a day and the
29 proportion of ultra-poor people (defined as the proportion of population below the
30 minimum level of dietary energy requirement) is highest within Southern Malawi at
31 approximately 34.2% (Gondwe 2014). Food security is particularly problematic in rural
32 areas where agriculture is primarily rain-fed, leaving smallholders vulnerable to climatic
33 shocks (Sahley et al. 2005). A single rainy (growing) season, between the months of
34 November and March, is followed by a dry season from April to October. Only a limited
35 number of farming households with access to *dimba* fields of the valleys located at the
36 source of streams, creeks, or rivers may take advantage of residual moisture and
37 extend cultivation beyond the end of the rains (Orr et al. 2009). High population
38 densities, small average plot size and poor soil quality further increase food insecurity.
39 The rural population is anticipated to grow from approximately 8.4 million in 1990 to
40 almost 20 million by 2030 (UN 2015b) and this growth will have wide ranging impacts

1 upon land and labour availability as well as market prices and productivity. Additional
2 exogenous trends such as soil degradation and climate change further compound food
3 insecurity (Schmidhuber and Tubiello 2007).

4
5 Indeed, food security is multidimensional issue (Connolly-Boutin and Smit 2016). A total
6 of four dimensions are recognized under the “four pillars” framework—created by FAO
7 and operationalised for modelling purposes in Dobbie and Balbi (2017)—including food
8 availability, access, utilisation and stability. The production of food is related primarily to
9 food *availability* (Headey and Ecker 2013). *Access* refers to the amount of food a
10 household can produce, purchase from the market and/or derive from other means
11 (Burchi and De Muro 2016). Households might draw upon social safety nets such as
12 food for work programmes or adopt coping strategies like selling livestock or borrowing
13 food (Devereux 2016). A third dimension, *utilisation*, refers to the ability of households
14 to process accessible food. This is dependent upon the household ability to obtain
15 sufficient quantities of fuel and clean water. Finally, *stability* dictates how robust
16 availability, access and utilisation dimensions are to shocks and stresses over time
17 (Burchi and De Muro 2016), such as those related to climatic and demographic change.
18 This article focuses primarily on the first two pillars: availability and access.

19
20 In the following paragraph we present the alternative principles of the behavioural
21 compass applied to the cropping strategies of farming households, and illustrate how
22 the combination of these principles is relevant for understanding community food
23 security (Fig. 1). The four storylines shown in Table 1, which describe observed
24 household cropping behaviours in rural regions of the Global South, are synthesized
25 from the Participatory Rural Appraisals (PRA: Schreckenberg et al. 2016) that took
26 place in multiple villages of the Zomba Region, in Southern Malawi.

27
28 According to the *competition and conformation* storyline (Table 1), a household would
29 maximise the yield of maize using all the available inputs subsidized by the government
30 (e.g. hybrid maize seeds and chemical fertilizer), and sell the produce at market. This is
31 currently the most common strategy within the villages of Southern Malawi, where
32 liberalisation reforms have failed to the produce the intended improvements and the
33 social norms regarding what to grow are widely and voluntarily accepted. Food
34 preferences are key in this regard. Maize has become the main food asset in the local
35 markets and is considered a strategic commodity. This social norm is not only
36 detrimental in terms of diet variety and healthy nutrition, but also makes the region more
37 at risk in the face of climate change by imposing a climate-vulnerable crop to the
38 majority. With *cooperation and conformation* a household would plant maize and share
39 the produce with the neighbours that have provided the labour to farm the land. This
40 used to happen more often in the past when the farm clubs were popular, before the

1 reforms, although mainly for cash crops. With *cooperation and differentiation* a
 2 household would maximise crop diversity choosing a mixed cropping pattern with
 3 different proportions of maize, pigeon pea, sweet potato, cassava, fruit trees, and
 4 vegetables. With *differentiation and competition* a household would maximise profit by
 5 planting cash crops, such as tobacco, peppers, or cotton. The emphasis on profit rather
 6 than on nutrients makes it a risky paradigm from a food security perspective, especially
 7 if we assume that self-sufficiency is relevant at the village level, as opportunity cost may
 8 drive agricultural entrepreneurs to invest on non-food crops beyond the social optimal
 9 level (e.g. Anderman et al. 2014).

10
 11

Table 1. Behavioural compass: village food security storylines.

<p>2nd <u>Quadrant: Competition & Conformation</u> The household produces the socially preferred crop, maximizing possible yield in order to have more commodity to be exchanged at the market. Seed and fertilizer subsidies influence this behaviour.</p>	<p>1st <u>Quadrant: Cooperation & Conformation</u> The household produces the socially preferred crop, after which the produce is aggregated to that of other producers belonging to a farm club, and shared equitably.</p>
<p>3rd <u>Quadrant: Competition & Differentiation</u> The household chooses a cropping pattern that maximizes income, with the objective of obtaining a more valuable commodity (diversified quality) to be exchanged at the market (e.g. cash crop).</p>	<p>4th <u>Quadrant: Cooperation & Differentiation</u> The household plants crops in order to maximize crop diversity, differentiating cropping pattern vs. its neighbours. Later food exchanges increase diet diversity.</p>

12
 13

2.2 Data and model formalization

15 Household-level data for a village in Southern Malawi was collected over a period of
 16 four days in July 2015 as part of the larger research project (Schreckenber et al.
 17 2016). Four trained Malawian enumerators used a household questionnaire to collect
 18 information on farming practices, crops planted, harvested and sold, other income
 19 generating activities, perceived food security, and socio-demographic characteristics of
 20 the households. After a village mapping exercise, in which three village representatives
 21 listed the household heads and mapped their locations, all households (N = 46) were
 22 selected (census) to participate. This same dataset was also used to calibrate an agent-
 23 based model to simulate the community food security into the future (Dobbie et al.
 24 2018). In this study we utilize the information on the crops planted, harvested and sold,
 25 which include: maize, groundnuts, tubers, pigeon peas, peas, sorghum, beans, soy,
 26 vegetables, and cotton.

27 We propose a game theory model for describing the household cropping strategies. We
 28 assume that each household is able to change its main crop type, at each time step,

1 and that there is only one planting season per year. The crop selection depends on two
 2 main factors, which describe personal and economic behaviour, as per behavioural
 3 compass. The first one is homophily, which determines the household tendency to
 4 imitate its neighbours and is therefore the parameter corresponding to the
 5 differentiation-conformation axis (i.e. personal behaviour). The second one describes
 6 other-regarding preferences, the influence of community level satisfaction in the
 7 household decision, and corresponds to the competition-cooperation axis (i.e. economic
 8 behaviour). The main rules of the game are given in the following bullet points:

- 9
- 10 • The topology of interactions is given by an initial network of N nodes, each node
 11 representing one household, whose connections, represented in a matrix A with
 12 elements a_{ij} , remain constant in time.
- 13 • Each node has a state, σ_i , which represents the household chosen main crop,
 14 and is updated asynchronously at each time step.
- 15 • The state, $\sigma_i \in \{1, S\}$, with S being the total number of crops, changes at each
 16 time step according to the best response mechanism. This, implies that each
 17 household tests all the possible crops and selects the one that maximizes its
 18 payoff. Crops are initialized according to a uniform distribution.
- 19 • The other-regarding preferences parameter $\alpha \in [0, 1]$, weights the importance of
 20 the community in the strategy decision for each household. Its value is fixed in
 21 time.
- 22 • The homophily parameter, $h \in [0, 1]$ represents the desired proportion of
 23 neighbours matching the same crop planting choices. The homophily parameter
 24 is also fixed in time.
- 25 • At each time step all households update their state σ_i to maximize their payoff
 26 (last bullet point). The total number of time steps or iterations is given by T .
- 27 • The satisfaction function $s_i \in [0, k_i]$, where k_i is the degree of node i , depends on
 28 the distance Δ_i , which measures how close are the links of a node from their
 29 optimal configuration. In other words, it retrieves the distance with respect to the
 30 number of connected households with same crop, according to h .

31

$$32 \quad (1) \quad \Delta_i = |h - \frac{1}{k_i} \sum_j \delta_{\sigma_i \sigma_j} a_{ij}|$$

33

$$34 \quad (2) \quad s_i = k_i e^{-\Delta_i}$$

35

- 36 • The payoff function $\Pi_i \in [0, \max k_i]$ depends on the individual and the community
 37 satisfactions, weighted by the individual other-regarding preferences parameter.

1 Therefore the payoff is not a purely economic measurement, but it depends on
2 both economic and personal behaviour.

3
4 (3) $\Pi_i = (1 - \alpha)s_i + \frac{\alpha}{N} \sum s_i$

5
6 The homophily parameter can capture both coordination ($h \approx 1$) and anti-coordination
7 dynamics ($h \approx 0$) and the situations in between. The last two points indicate that
8 individual satisfaction is subjective to each household and driven by own level of
9 homophily: own satisfaction is assessed in comparative terms. Other households are
10 considered in terms of their state (i.e. the planted crop) and we assume full knowledge
11 of nodes state within the network.

12 Payoff combines individual and community satisfaction; here the other-regarding
13 preferences parameter is key: the higher the more important will be the satisfaction of
14 others. This formalization allows the consideration of two scales of concern (individual
15 and communal) in the objective function driving household behaviour.

16 We consider cooperation as driven by the importance assigned to the community in
17 terms of neighbours' satisfaction. In this framework, high other-regarding preferences (α
18 ≈ 1) means to adapt crop selection so that neighbours' satisfaction can maximize
19 individual payoff in addition to individual satisfaction. Low other-regarding preferences
20 ($\alpha \approx 0$) means to not consider neighbours' satisfaction. Regardless of this parameter all
21 households maximize their own payoff and thus are modelled as self-interested
22 individuals.

23
24 In the following section, we describe the model behaviour by means of numerical
25 simulations on artificial networks of different kinds. Then, in Results, we apply the model
26 to the data of a village in Southern Malawi and summarize the main findings.

27 28 **2.3 Model behaviour**

29 The game theory model has three control parameters, namely the other-regarding
30 preferences parameter α , the homophily parameter h , and the number of possible node
31 states S , which represent the main crop types. The model can be implemented on
32 different network topologies (Latora, Nicosia and Russo 2018). Here we compare three
33 networks, namely Regular Lattices (RL), Random Networks (RN) and Scale Free
34 Networks (SF), with the same number $N = 100$ of nodes and the same average degree
35 $\langle k \rangle = 4$. Regular Lattices of $\langle k \rangle = 4$ correspond to a uniform distribution of the nodes in
36 the plane, where each node is connected to its two nearest neighbours in each
37 dimension. In order to have $\langle k \rangle = 4$ for all the nodes, we considered periodical
38 boundaries. Random Networks (Erdos and Rényi 1960) of N nodes and $\langle k \rangle = K$ are
39 achieved by fixing the link probability to $K/(N-1)$. Their degree distribution can be well
40 approximated with a Poisson probability distribution. Scale Free Networks (Barabási

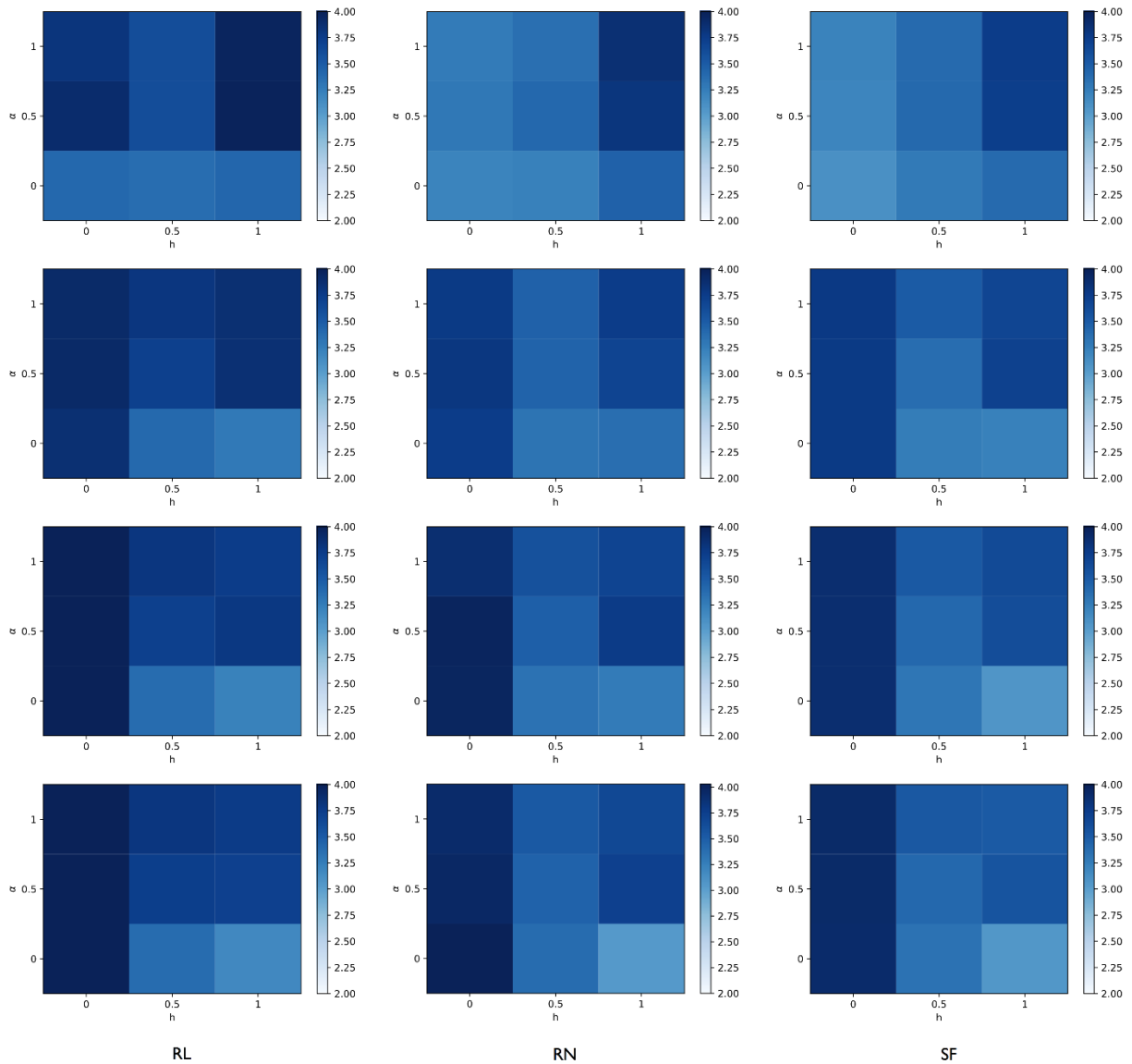
1 and Albert 1999) are characterized by a power law decay in the degree distribution,
2 allowing the existence of hubs, or nodes with a high degree.

3 Simulations have been performed for $\alpha \in \{0, 0.5, 1\}$, $h \in \{0, 0.5, 1\}$, and $S \in \{2, 3, 4, 5\}$,
4 and the results have been averaged over ensembles of 100 different network
5 realizations. In each simulation we have run the game for $T = 20$ iterations under a best
6 response update rule, which assumes that each of the nodes tests all the possible
7 states and selects the one with the highest payoff. The value of $T = 20$ is enough to
8 guarantee the convergence of the system to the equilibrium state. The combination of
9 the best response update mechanism and the small size of network explains why most
10 of the information of the equilibrium solution is provided for small values of T (e.g. $T =$
11 5).
12

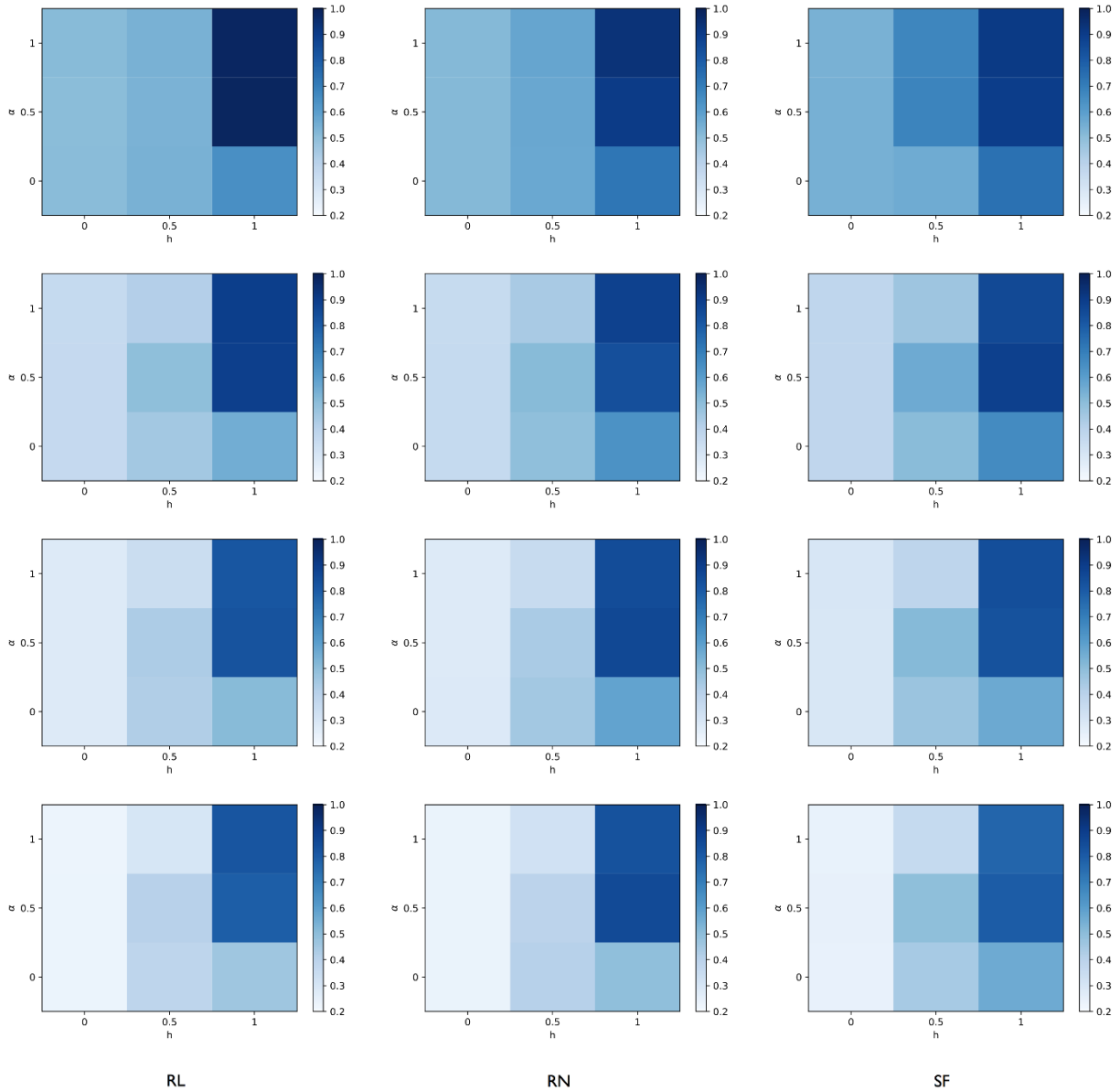
13 The asymptotic values of payoff and dominating state fraction are shown in Fig. 2 and
14 Fig. 3. The first is the value of payoff when the system approaches equilibrium, the latter
15 represents the proportion of households within the village matching the same crop
16 planting choices. Both figures show the average values of output as a function of other-
17 regarding preferences and homophily for different topologies and values of S . The
18 columns correspond to the topology, RL, RN and SF, while the rows correspond to the
19 total number of states. In each subplot the colour in the white-blue scale represents the
20 value of the payoff for the specific selection of α and h .
21

22 From the simulations we can infer the most relevant features of the model:

- 23 • The relation between S (number of states/crops) and $\langle k \rangle$ (average degree of the
24 network) determines the strategy with the highest payoff. For low $S/\langle k \rangle$,
25 conformation is more favourable than differentiation. The opposite is true for
26 higher values of $S/\langle k \rangle$.
- 27 • The system is polarized between consensus (high dominating state fraction (df)),
28 for $h = 1$, and dissension (low df), for $h = 0$. The intermediate situations imply
29 lower payoffs. This indicator suggests the existence of a phase transition
30 between these two regimes.
- 31 • Cooperation is especially relevant when the community converges on crop
32 planting choices: it is a necessary ingredient for the success of households that
33 conform. The opposite is not true. Differentiation can survive without cooperation.
- 34 • The dynamics are stable except for the combination of $h = 0.5$ and $\alpha \in \{0.5, 1\}$.
35 We expand on this finding when studying the empirical case study. The higher S ,
36 the lower the fluctuations. The tendency towards stability depends on the
37 topology. RL achieve it faster than SF, and these are faster than RN (See Online
38 Resource 1: Fig. 6 and Fig. 7. Here we plot the absolute value of the discrete
39 derivative at $T = 20$, as a mechanism for evaluating if the result corresponds to
40 an equilibrium situation).



1
2 **Fig. 2 Payoff when the system approaches equilibrium (asymptotic payoff) for different**
3 **values of other-regarding preferences and homophily in different network topologies.**
4 Number of nodes $N = 100$, iterations $T = 20$, average degree $\langle k \rangle = 4$. Each row of figures
5 corresponds to the simulations for a number of node states $S = \{2, 3, 4, 5\}$. The columns refer to
6 the type of network: Regular Lattice (RL), Random Network (RN) and Scale Free Network (SF).
7 Each subfigure represents a combination of type of network and total number of node states.
8 The squares in the subfigures describe the payoff in a scale from 2 (lighter) to 4 (darker) for
9 different combinations of other-regarding preferences and homophily.



1
2 **Fig. 3 Proportion of households converging to the same crop selection (dominating state**
3 **fraction) for different values of other-regarding preferences and homophily in different**
4 **network topologies.** Number of nodes $N = 100$, iterations $T = 20$, average degree $\langle k \rangle = 4$.
5 Each row of figures corresponds to the simulations for a number of node states $S = \{2, 3, 4, 5\}$.
6 The columns refer to the type of Network: Regular Lattice (RL), Random Network (RN) and
7 Scale Free Network (SF). Each subfigure represents a combination of type of network and total
8 number of node states. The squares in the subfigures describe the dominating state fraction in a
9 scale from 0.2 (lighter) to 1 (darker) for different combinations of other-regarding preferences
10 and homophily.

11

13. Results

2

3 For the case under study, a small village in Southern Malawi, we have $S = 4$ main crop
4 types, and precise information on the type of crop adopted by each one of the 46
5 households in the village. The total number of crops is 8, but our model only takes into
6 consideration the crop with the maximum harvesting quantity for each household. Even
7 after this simplification, the use of the different crops is quite heterogeneous, with
8 adoption frequencies equal to (35, 9, 1, 1). These values account for the number of
9 households that have selected each of the crops as their preferred option and
10 correspond to 76% of the households adopting the same crop, making the dominating
11 state fraction (df) equal to 0.76. (See Online Resource 1, Fig. 8).

12

13 Our purpose here is to map the village in the bi-dimensional space described by the
14 behavioural compass (Fig. 1), by matching the model outcomes with the crop statistics
15 observed in the data. At the same time, this study shows the combination of other-
16 regarding preferences and homophily parameters valid for the description of real
17 scenarios.

18 In order to make use of our game theory model in a social network setting, we need to
19 infer the topology of the real network of interactions among households, which defines
20 the neighbours considered when measuring own satisfaction. This data was not
21 collected and is typically difficult to access.

22 A reasonable assumption is to model the social network in the village as a small world.
23 Small world networks (Watts and Strogatz 1998) are characterized by short average
24 path lengths (i.e. shortest path between all pairs of households) and high clustering (i.e.
25 the ratio of household connections that are connected among themselves). Such
26 assumption is also supported by evidence from the literature (e.g. Spielman et al. 2011,
27 Ligon and Schechter 2012).

28

29 The small world topology can be achieved with the Watts-Strogatz mechanism based
30 on the random rewiring of regular ring lattices (Watts and Strogatz 1998). Based on
31 Dobbie et al. (2018), who modelled the same village setting with a coupled agent-based
32 and network approach, we simulated the households as a small-world network with $N =$
33 46 average degree $\langle k \rangle = 9$ and rewiring probability $p = 0.25$. For this topology we run
34 the game trying to determine the value of h that corresponds to $df = 0.76$. In order to do
35 so, we discretize the $[0, 1]$ interval in 20 parts and obtain the value of the payoff and df
36 of an ensemble of 100 networks in each of them (See Fig. 4).

37

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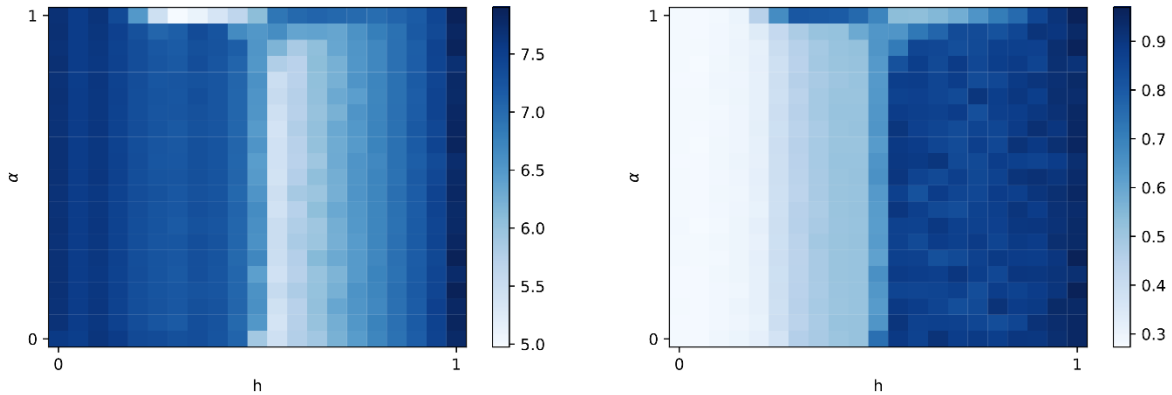


Fig. 4 Case analysis: a Southern Malawi village. Simulations for an ensemble of 100 small world networks of $N = 46$ nodes, average degree $\langle k \rangle = 9$, and rewiring probability $p = 0.25$, for different values of homophily (h) and with number of node states $S = 4$. Payoff is displayed on the left side and dominating state fraction on the right side. Both display three different regimes of the system, with higher values in darker shades.

In the numerical simulations we find three different regimes of the system:

1. The first one, for $h < 0.5$, is characterized by a high payoff. The dominating state fraction (df) increases with h , which enables a large average payoff, because households tend to experience in their neighbours the diversity of crops desired. In other words, the households succeed in adapting to a relatively high level of diversity, and achieve an overall high payoff for the village.
2. In the second regime, $h > 0.5$, the system is driven to consensus ($df > 0.8$), which is rewarded for high values of h , but not for average ones. For $h < 1$ households aim at a small amount of diversity among crops, but the system does not provide it. Thus a higher payoff is granted to the households that conform. Differently from the first region, households are driven to consensus for all the values of h , which is detrimental in terms of payoff.
3. The last regime is found in the area between the previous two. It is constrained to the values close to $h = 0.5$ and $\alpha = 1$. In the frontier with the second region, $h > 0.5$, we can see how cooperation allows to adapt the df to the value of h . The opposite behaviour is achieved for $h < 0.5$: the fluctuations break the successful dynamics of the best response mechanism and lead to a consensus state. Therefore, within the third region, cooperation is beneficial when the system is characterized by $h > 0.5$, and detrimental for $h < 0.5$.

It is important to notice that, in the first two regions, the payoff is largely determined by the homophily parameter (h) and the other-regarding preferences parameter (α) is almost irrelevant. Both differentiation and conformation can be successful strategies although they imply very different outcomes at the system level: dissension in the first case and consensus (on which crop to plant) in the latter.

1 However, the third region, as we show in the next paragraphs, is the one where our
2 case study is located: maize is the widely preferred staple food and most of the
3 residents wouldn't consider a meal without maize as satisfactory (Dorward and Kydd
4 2004). The main property of the third region is that it switches the behaviour of the two
5 other regions at their margins. With high other-regarding preferences, when everybody's
6 satisfaction is considered in the individual payoff, conformation can lead to higher
7 payoffs.

8

9 In order to fit our model to the village dataset, we considered values in an interval of
10 0.01 with respect to the observed df . However, these results suggest the existence of a
11 continuous function of α and h fulfilling the desired condition, $df = 0.76 \pm 0.01$, which
12 may be achieved by increasing the resolution in the simulations. To this end we have
13 further inspected the system around the third region, zooming into values of h
14 constrained to the $[0.45, 0.55]$ interval (Online Resource 1, Fig. 9), and separately, to
15 values of α constrained to the $[0.9, 1]$ interval (Online Resource 1, Fig. 10).

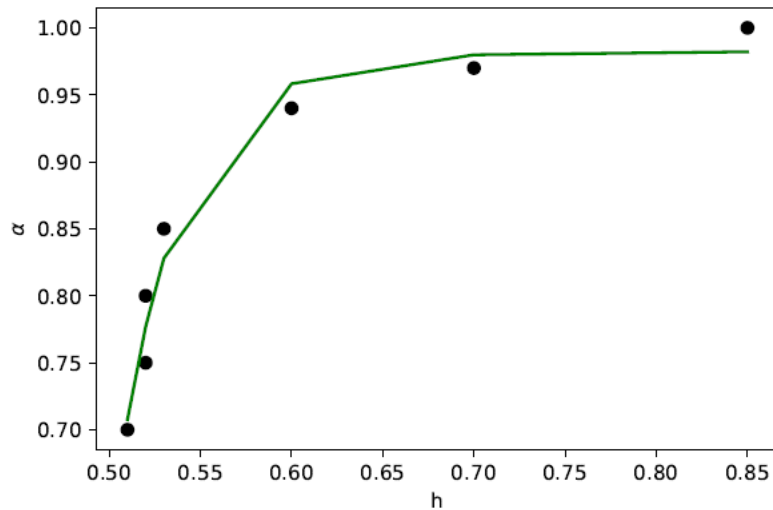
16

17 The first zoom sheds light on the fact that cooperation with intermediate values of h
18 prevents the system from achieving an equilibrium situation. However, the highest
19 fluctuations of df have an upper bound of 0.12, which indicates an overall predictable
20 behaviour. The second zoom explains the success of cooperation for $h > 0.5$ and its
21 redundancy for $h < 0.5$. The simulation shows that the fluctuations are exclusive of the h
22 < 0.55 region, and prevent the system from achieving a higher payoff. Thus we cannot
23 associate this outcome to a stationary state of the system.

24

25 Finally, we identify the values of α and h that correspond to the dataset provided by our
26 case study (Online Resource 1, Table 2). We can distinguish two sets of points. A first
27 one of $h < 0.5$ and $\alpha = 1$, and a second one with $h > 0.5$. As we anticipated, the first set
28 lies in a high fluctuations region, we therefore focus on the second one, for which we fit
29 a curve that provides the part of the phase space that corresponds to our dataset (Fig.
30 5). Fig. 5 reinforces the insight provided by the analysis of the generic model in Section
31 3.1: conformation is a winning strategy when cooperation is in place. In particular, the
32 last two points display above average payoffs. In the next section we elaborate on what
33 are the implications of these dynamics in terms of food security.

34



1
2 **Fig. 5 Curve fitting the combinations of homophily (h) and other-regarding preferences**
3 **(α) with highest payoff.** The dots show the points with homophily h > 0.5 in Table 2 (Online
4 Resource 1), which are the combinations of h and α for which the value of dominating fraction in
5 the numerical simulations matches the one of the village. The relation between the parameters
6 of the behavioural compass, which describe the village, is captured by the non-linear function
7 $\alpha = a h^{-15} + b$ with $a = -1.12795 \cdot 10^{-5}$ and $b = 9.8225 \cdot 10^{-1}$

8
9 **94. Discussion**

10
11 This article has presented the behavioural compass—a way of framing human decisions
12 considering the predisposition of human beings to compete/cooperate with and
13 imitate/differ from own peers—and applied it to the case of community food security in a
14 rural region of the Global South. Although we argue that this representation of human
15 behaviour could be applied to a heterogeneous set of social phenomena, including off-
16 farm livelihood strategies, this is outside the scope of the current article.

17
18 The Sustainable Development Goals (SDGs) adopted by the United Nations have set a
19 clear target to end food insecurity by 2030. The aim of Goal 2 is to “end hunger, achieve
20 food security and adequate nutrition for all, and promote sustainable agriculture.” How
21 agri-food systems can self-organize to promote practices capable of meeting this goal is
22 still an open question. A review by Rausser et al. (2015) describes two main global
23 paradigms: the Industrial food and agricultural industry (IFA) and the naturalization food
24 and agricultural paradigm (NFA). IFA relates to industrial food production and large-
25 scale food distribution system; NFA captures a multi-faceted movement concerned with
26 key principles such as: local, small and organic; slow food (Petrini 2003); agroecology
27 (Gliessman 2009); food sovereignty (Patel 2009) and diversified farming systems
28 (Kremen and Miles 2012).

1 It is reasonable to think that that agri-food systems can be mapped within a continuum
2 between these two paradigms and dynamically move in one direction or the other
3 according to social and ecological conditions. For example, a country of the Global
4 South like Malawi has been witnessing a shift towards IFA under the liberalisation
5 reforms that happened since the eighties (Dorward and Kydd 2004). At the same time
6 NFA is gaining momentum in the cities of the Global North, where the masses have
7 been relying on the IFA paradigm, almost since the industrial revolution (Ilieva 2016).

8
9 In this study we adopted a village-centric perspective and a system approach, whereby
10 the system performance is stemming from the behaviour of and interactions among the
11 individual components, to explore how household behaviour can affect food security at
12 the community level. We implicitly considered that self-sufficiency in food production is a
13 positive feature for a rural agri-food system of the Global South, but here we discuss its
14 consequences.

15
16 Because satisfaction, and thus payoff, is not measured in terms of nutritional
17 requirements we can only suggest how human behaviour is expected to influence food
18 security. Indeed, our model is primarily concerned with behavioural dynamics in crop
19 adoption and the level of dissension/consensus within the community. To draw
20 conclusions on the nutritional level of community food security, we would have to couple
21 this behavioural model with agricultural and environmental modules (see Balbi et al.
22 2015) and a module of exchanges, including market interactions and bartering
23 dynamics, possibly considering food imports and exports. In the work presented here
24 we assume that all the households have the same nutritional requirements, and that the
25 production costs and economic profit are equal for all crop types. We also don't
26 consider any difference in the crops that are consumed by the producers and the ones
27 that are sold in market. Accessing food via production or via market doesn't affect the
28 benefits and food price is exactly the production cost, thus the community is assumed to
29 operate through direct food exchanges.

30
31 Although past studies have emphasized the role of cooperation in self-organized
32 systems (see Perc et al. 2017 for a review), our game theory model suggests that
33 homophily—the continuum between conformation and differentiation—can play a
34 relevant role. Both conformation and differentiation can be successful strategies leading
35 to high payoffs at the individual level, but according to which prevails, they drive the
36 system to very different community outcomes. In the first case, we have a community
37 converging on crop planting choices, in the second case the community maintains crop
38 diversity.

1 Both cases can happen regardless of the importance given to others satisfaction, which
2 is what drives cooperation in our model. This can have significant consequences in
3 terms of food security at the village level. The first case is a specialization case that
4 could succeed only if the village is connected to the outside via food exchange
5 mechanisms (e.g. regional food markets) through which households can sell their own
6 produce and buy the lacking nutrients. The alternative is a community that can perhaps
7 self-sufficiently meet its own caloric needs, but not a healthy diet diversity. The second
8 case is a diversification case that can succeed under a self-sufficiency scenario, but
9 requires internal exchanges to happen (via local market or bartering) within the village
10 to meet the necessary diet diversity at the household level.

11
12 However, the case study we analysed in this article, using empirical data from a rural
13 village in Southern Malawi, has provided more insights about the role of cooperation.
14 Our analysis suggests that the village is located in a very particular region of the
15 parameters space where the influence of other-regarding preferences is relatively high
16 for any point in the curve presented in Figure 5. This curve acts as the boundary
17 between the regions representing the two cases mentioned above. Thus cooperation
18 has a key role to play: the higher other-regarding preferences the more the system is
19 driven to consensus, and the analysis of expected payoffs suggests that the village is
20 better off with consensus. Partial dissension can survive for other-regarding preferences
21 below 0.9. It thus seems that sacrificing some degree of cooperation in favour of more
22 crop diversity would reduce the individual expected payoff, although it would make
23 sense from a nutritional diversity perspective at the community level. Indeed, the current
24 village situation is confirming these insights: in this part of Malawi food exchanges are
25 largely market-driven with regional markets playing an important role and maize is by far
26 the socially preferred crop (Schreckenberget al. 2016). Moreover, the results reinforce
27 the idea of close to perfect knowledge among households at the village scale, and the
28 convenience of considering others satisfaction in own payoff, thus being more
29 cooperative.

30
31 One additional limitation of our work is the assumption about the real topology of the
32 village network. Although we expect an error of approximation of the actual social
33 network, our assumption is based on previous studies and the sensitivity analysis in
34 Section 2.3 explains the similarity of model behaviour when the topology is altered.
35 Accordingly, we expect the relation between the other-regarding preferences and the
36 homophily parameters to be robust against changes in the topology. For what concerns
37 village-level results, we expect them to hold for small communities (e.g. 30 to 100
38 households corresponding to 100-500 individuals) with a relatively high degree of
39 mutual knowledge among community members and parcelled access to the resource—
40 in this case farming land.

1 Further analysis will explore the explanatory power of the behavioural compass at
2 multiple spatial scales. We argue that the behavioural compass could be extended on a
3 third axis, thus representing a three dimensional space, to capture multiple levels of
4 governance (Lebel et al. 2006) and their institutions (from the micro- to macro-scale), in
5 line with the thinking of Waring et al. (2015).
6

75. Conclusions

8
9 The main finding of our simulations suggests that individual and community success is
10 more strongly related to personal behaviour than to economic behaviour. In other
11 words, how individuals relate to social norms has greater effect than how much they
12 care about the success of their neighbours (i.e. our working definition of cooperation).
13 While we expected the latter to be the main driving force in determining community
14 success, we found that cooperation is only necessary for community success when the
15 community converges on crop planting choices. On the contrary, differentiation, the
16 affirmation of the individual unique identity, can succeed with or without cooperation.
17

18 Under a food security perspective, when there is a diversity of options about crop
19 adoption, differentiation is likely to deliver more positive outcomes. Only when the
20 options are few, then conformation is a winning strategy. In a situation such as a rural
21 region of the Global South—in this article we used the example of a village in Southern
22 Malawi—this translates into clear policy implications about the role of food sovereignty,
23 and in particular food knowledge sharing. Food security, among other things, depends
24 on a variety of nutrients for a healthy and balanced diet, thus enabling biological
25 diversity in agriculture can be beneficial.

26 Instead, it seems that the policy reforms of the past have pushed countries like Malawi,
27 similarly to other rural African regions, to specialize their agricultural production towards
28 one or two socially preferred crops, via crop-specific subsidies and other incentives.
29 This made sense under a scenario of market liberalisation that was going to deliver
30 efficient redistribution of food commodities among different regions (for food security
31 purposes), thus allowing beyond-system food transactions. Evidence from the ground
32 suggests that this paradigm has only partially delivered its promises (see Dorward and
33 Kydd 2004) and that stimulating more diverse crop adoption at the household level is a
34 reasonable strategy to improve community food security.

35 We argue that this process is not likely to happen endogenously, because, in the
36 current setting, households regard conformation as a safer strategy, due to the
37 dominant social norms on crops.
38

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14

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