

The evolution of human cooperation: homophily, non-additive benefits, and higher-order relatedness

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Acknowledgements

Country:

I acknowledge the Turrbal and Yugara people and as the owners of this land. I pay respect to their Elders, past and present, and recognise this land has always been a place of teaching, learning, and research.

Coauthors:



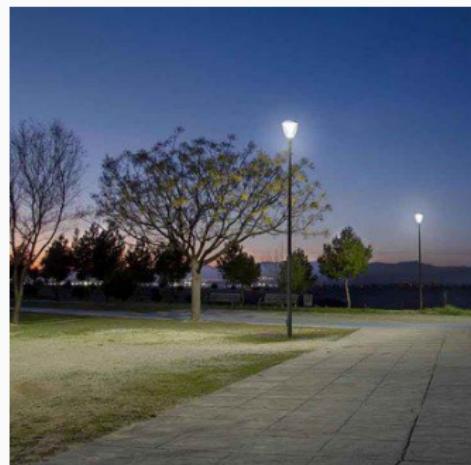
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Humans are cooperative

- Introspection — 'I am a moral being'
- Humans are a highly cooperative species
- Eusocial insects— relatives
- Humans— non-relatives, strangers



Why cooperate?

- Help others at a cost to yourself?
 - Violates evolutionary logic
 - Though tricky to think about costs and benefits
- Problem in its pure form – Game Theory

Prisoner's Dilemma

		player 2	
		cooperate	defect
		cooperate	3, 3
player 1	cooperate	4, 0	1, 1
	defect	0, 4	1, 1

- Implications:
 - You should never cooperate
 - A cooperative type will never evolve
- Yet people do
 - Anecdotally
 - Experimentally

Scope of this talk

		player 2	
		cooperate	defect
player 1	cooperate	3, 3	0, 4
	defect	4, 0	1, 1

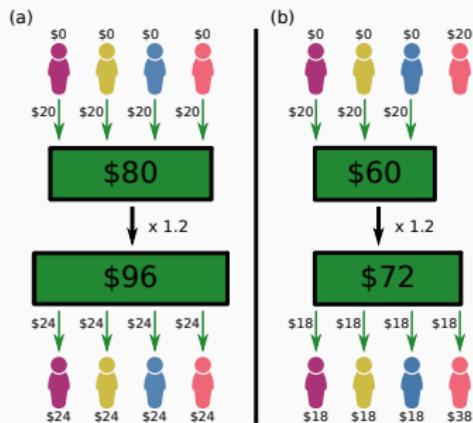
- Two big ideas I won't be covering today:
 1. Repeated interaction
 2. Reputation
- I will assume interactions are:
 1. One-shot
 2. Anonymous
- What I'll be talking about instead:
 1. Kin selection
 2. Nonlinear payoffs (or economies of scale)

Public goods game example

- Group-game version of Prisoner's Dilemma

- Example:

1. Public good that multiplies contributions by 1.2
2. Everyone contributes → maximise total payoffs
3. However, not contributing maximises individual payoff

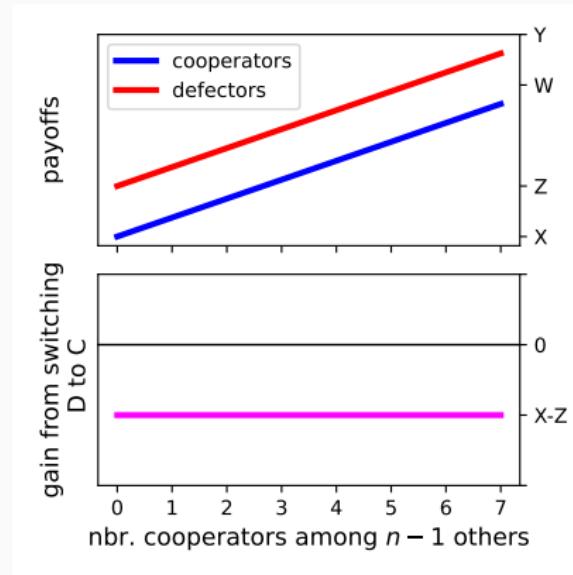


- Never makes sense to contribute
 - Returns are split equally
 - Marginal per-capita return = $1.2/4 = 0.3 < 1$
 - 30c return for every \$1 contributed

Linear public goods game

5

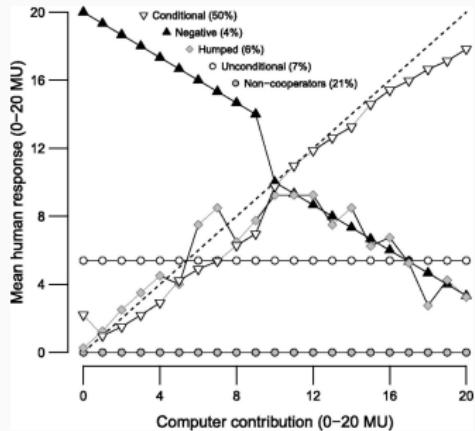
- Example was a linear public goods game
 - Benefit increases at constant rate with nbr. cooperators
 - No matter how many cooperators in the group, always lose by switching C to D
- n -player generalisation of PD



7

How do people really behave in linear PGGs?

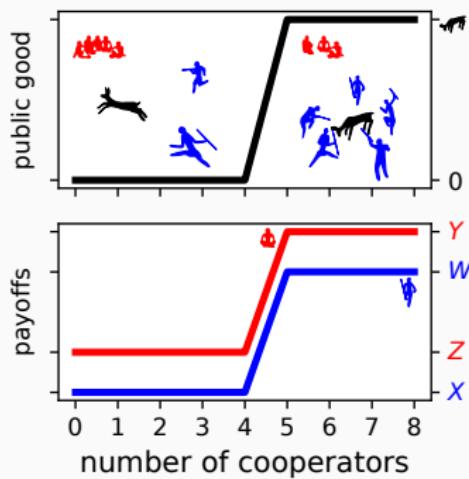
- Contribute + depends on contribution of others; why?
- Computer opponent, so not for fairness reasons
- Seem to sincerely believe this is payoff maximising
 - Iterated game → learn not to contribute
 - But it's interesting that's their default
- Deeply unnatural scenario: one-shot interaction with strangers
- Previous work focuses on two 'mistakes': iterated game, reputation concerns
- I want to focus on another 'mistake': mistaking the linear game for nonlinear



Burton-Chellew *et al.* (2016,
PNAS)

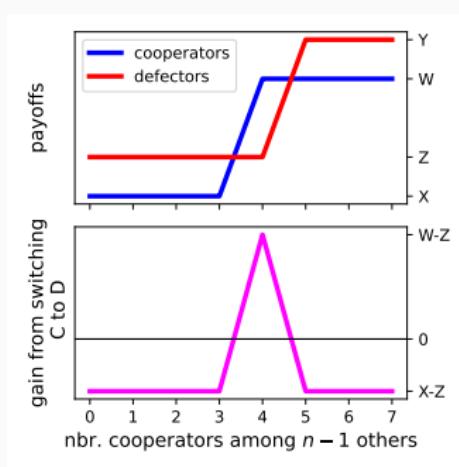
Nonlinear public goods game

- Claim: sigmoid-shaped benefit functions particularly relevant to our early history
- Defectors always get higher payoffs than Cooperators

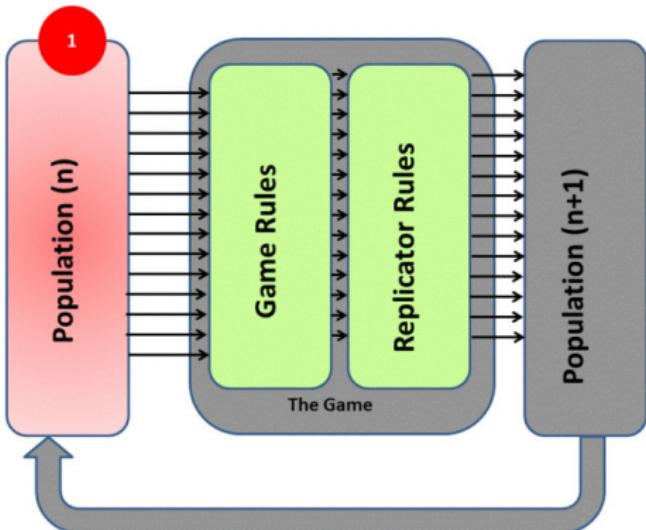


Nonlinear public goods game

- Claim: sigmoid-shaped benefit functions particularly relevant to our early history
- Defectors always get higher payoffs than Cooperators
- However, if you are the pivotal player, you should cooperate
- Evolutionary perspective:
 - if cooperates rare, don't cooperate
 - if cooperate common, might get higher payoffs if you also cooperate

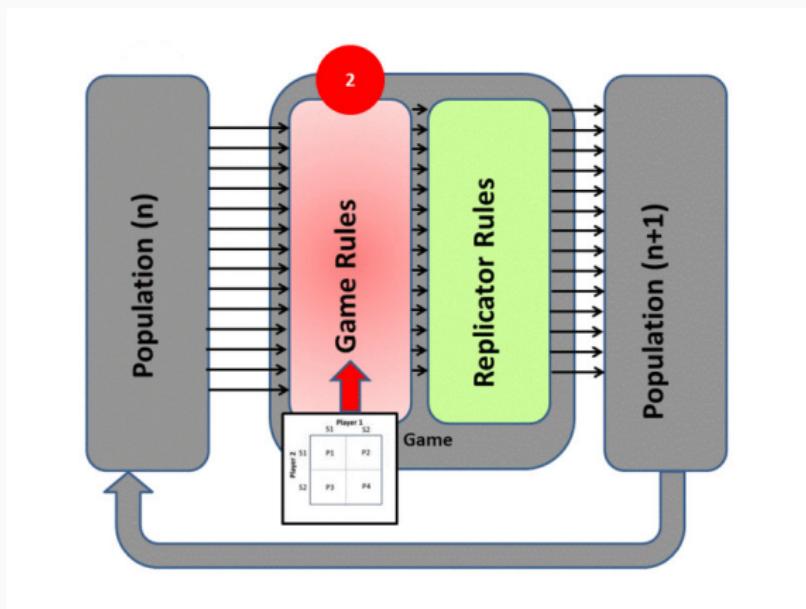


Replicator dynamics overview



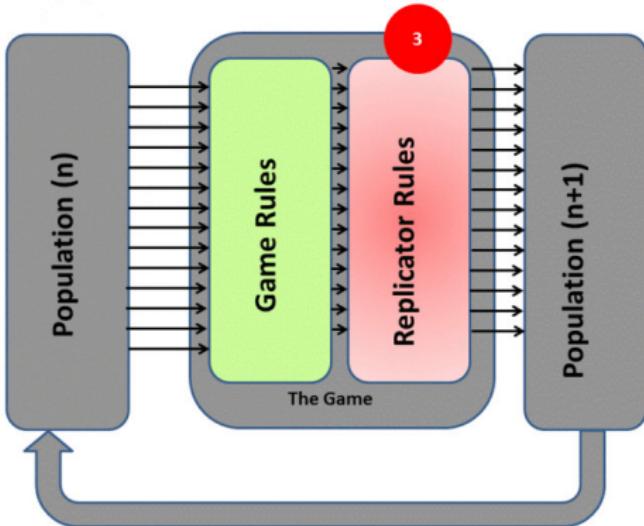
(HowieKor, Creative Commons)

Replicator dynamics overview



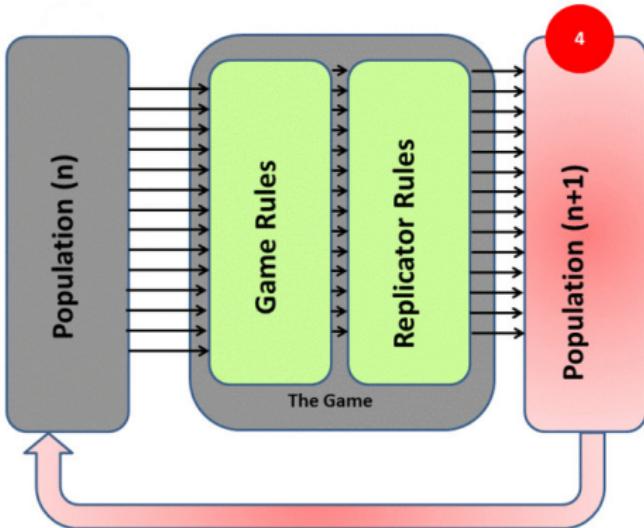
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Replicator dynamics overview



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Replicator dynamics overview



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Replicator dynamics maths (1)

Change in number of x -strategists in population:

$$\dot{N}_x = N_x (\beta + \bar{\pi}_x)$$

background effects on fitness

nbr. x -strategists

expected payoff to x -strategists

Change in proportion of x -strategists in population:

$$\dot{p}_x = p_x \left(\bar{\pi}_x - \sum_{i=1}^m p_i \bar{\pi}_i \right)$$

expected payoff to x -strategists

proportion of x -strategists

m is nbr. strategies

expected payoff in population

Replicator dynamics maths (2)

$$\dot{p}_x = p_x \left(\bar{\pi}_x - \sum_{i=1}^m p_i \bar{\pi}_i \right)$$

expected payoff to x -strategists
 m is nbr. strategies

proportion of x -strategists
expected payoff in population

New notation: strategy indicator vector $e_x = (0, \dots, 0, \underbrace{1}_{x\text{-th}}, 0, \dots, 0)$

For 2-strategy games:

$$\begin{aligned} \bar{\pi}_x &= \sum_{g=0}^{n-1} \pi(e_x, g_{\text{nf}}) \quad \mathbb{P}[G_{\text{nf}} = g_{\text{nf}}], \\ &= \sum_{g_{\text{nf}}=0}^{n-1} \pi(e_x, g_{\text{nf}}) \quad \binom{n-1}{g_{\text{nf}}} p_x^{g_{\text{nf}}} (1-p_x)^{n-1-g_{\text{nf}}} \end{aligned}$$

payoff
probability g_{nf} non-focals are cooperators

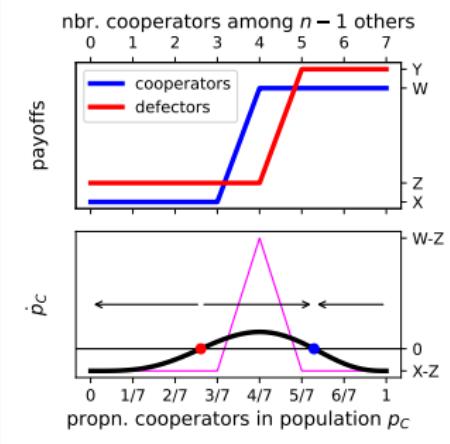
binomial

Cooperation can be sustained

10

Two main known results:

1. Cooperation can be sustained
 - Do people ‘mistake’ linear games for a nonlinear ones?
2. But cooperation cannot invade
 - Imagine a small nbr. of cooperators invading defectors...



note: \dot{p}_c is a Bezier curve
(Jorge Peña's papers)

But what if, instead of randomly formed groups, groups tend to form of family members?

Homophilic group formation



Change in proportion of x -strategists in population:

$$\dot{p}_x = p_x \left(\bar{\pi}_x - \sum_{i=1}^m p_i \bar{\pi}_i \right)$$

expected payoff to x -strategists

↑
proportion of x -strategists

↑
expected payoff in population

where

$$\bar{\pi}_x = \sum_{g_{\text{nf}}=0}^{n-1} \pi(e_x, g_{\text{nf}}) \mathbb{P}[G_{\text{nf}} = g_{\text{nf}} \mid G_0 = e_x]$$

nonfocal strategy distribution depends on focal's strategy

Some notation

Let ρ_ℓ be the probability that ℓ players sampled without replacement from the group have strategy 1.

prob. ℓ sampled have m common ancestors

$$\rho_\ell = \sum_{m=1}^{\ell} \theta_{\ell \rightarrow m} p_1^m$$

propn. strategy-1 in popultn

Examples:

- Sample 1 individual: $\rho_1 = p_1$
- Sample 2 individuals:

$$\rho_2 = \theta_{2 \rightarrow 1} p_1 + \theta_{2 \rightarrow 2} p_1^2$$

prob. same ancestor prob. strategy-1

prob. two ancestors prob. both strategy-1

Hisashi's equation

Ohtsuki (2014, Phil Trans R Soc):

$$\dot{p}_1 = \sum_{g_{\text{nf}}=0}^{n-1} \sum_{\ell=g_{\text{nf}}}^{n-1} (-1)^{\ell-g_{\text{nf}}} \binom{\ell}{g_{\text{nf}}} \binom{n-1}{\ell}$$

relatedness terms

$$[(1 - \rho_1) \rho_{\ell+1} \pi(e_1, g_{\text{nf}}) - \rho_1 (\rho_\ell - \rho_{\ell+1}) \pi(e_2, g_{\text{nf}})]$$

↑ payoff terms ↑

Linear PGG is a function of dyadic relatedness only

- If the PGG is linear, only need dyadic relatedness

$$\dot{p}_1 = f(\theta_{2 \rightarrow 1})$$

dyadic relatedness
↓
because:
↑ (Hamilton's r)

- Payoff function in n -player game can be written as a sum of payoffs in 2-player games

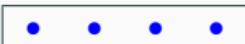
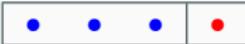
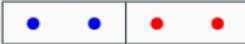
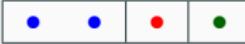
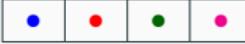
$$\pi^{(n)}(e_x, g_{\text{nf}}^{(n)}) \equiv \sum_{g_{\text{nf}}^{(2)}} \pi^{(2)}(e_x, g_{\text{nf}}^{(2)})$$

payoff in n -player game
↓
payoff in 2-player game
↓

- So the n -player linear game is equivalent to the sum of 2-player games
- So only dyadic relatedness is needed to calculate expected payoff
- But if the payoff function is nonlinear, higher-order relatedness coefficients are needed

How do we calculate the higher-order relatedness terms?

From group family-size distribution. For example:

	partition	$\theta_{2 \rightarrow 1}$	explanation
$F_{[4]}$		1	Any 2 will have a common ancestor.
$F_{[3,1]}$		$\frac{3}{4} \times \frac{2}{3} = \frac{1}{2}$	Both must be blue (family size 3).
$F_{[2,2]}$		$1 \times \frac{1}{3} = \frac{1}{3}$	Choose any, then its 1 family member.
$F_{[2,1,1]}$		$\frac{2}{4} \times \frac{1}{3} = \frac{1}{6}$	Only possible in the partition of 2.
$F_{[1,1,1,1]}$		0	Not possible.

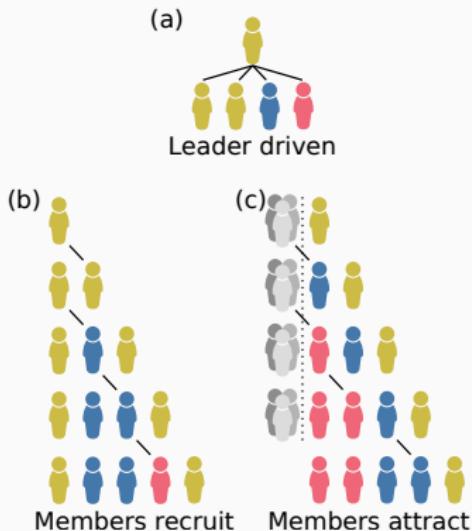
So if we can calculate the F_q , we can calculate the needed $\theta_{l \rightarrow m}$

(a) Leader driven:

- The leader is chosen at random from the population.
- Leader recruits/attracts kin with probability h and nonkin with probability $1 - h$.
- Group family size distribution

$$F_{[\ell, 1, \dots, 1]} = \binom{n-1}{\ell-1} h^{\ell-1} (1-h)^{n-\ell}.$$

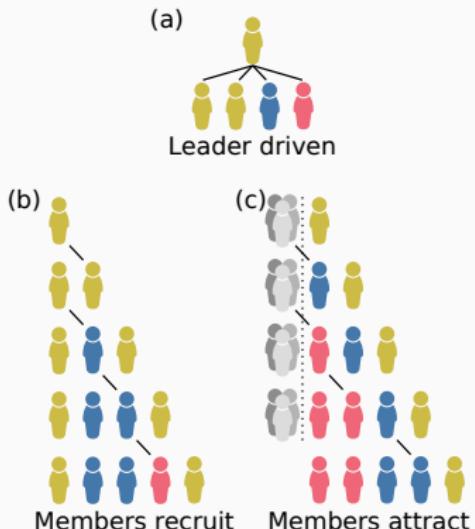
$h =:$ genetic homophily



$h =:$ genetic homophily

(b) Members recruit:

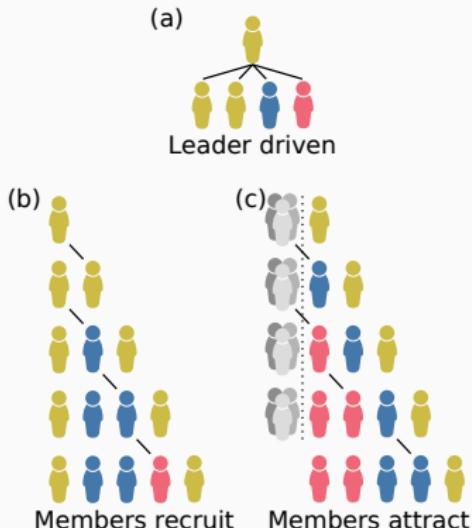
- First member is chosen at random.
- Current group members have an equal chance to recruit the next member.
- Member recruits kin with probability h and nonkin with probability $1 - h$.
- Equation in Kristensen *et al.* (2022)



(c) Members attract:

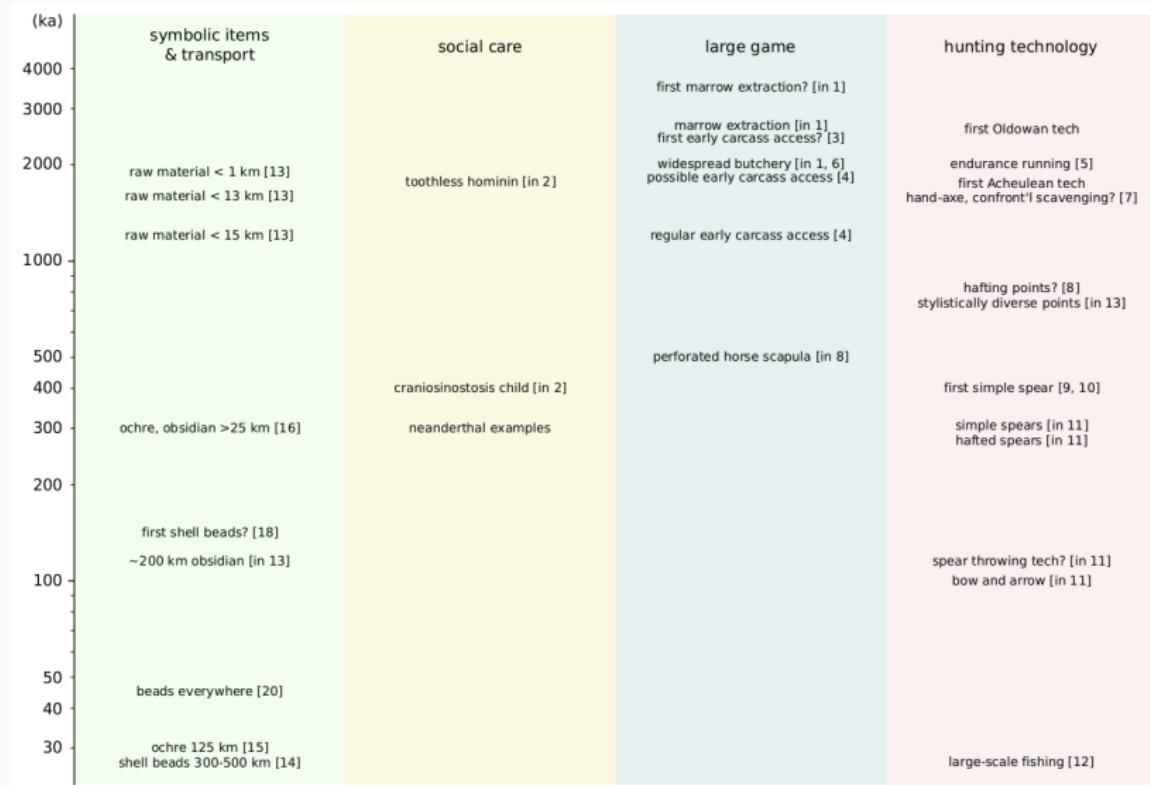
- First member is chosen at random.
- Current group members have equal weighting 1 of attracting a new member who is kin,
- but nonkin members are also attracted to the group itself with collective weighting $\alpha \in [0, \infty)$.
- See Ewens' formula (Ewen 1972) for equation.

$h =:$ genetic homophily

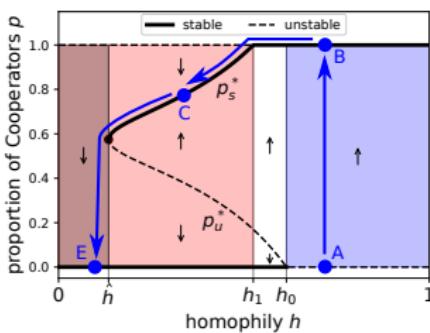
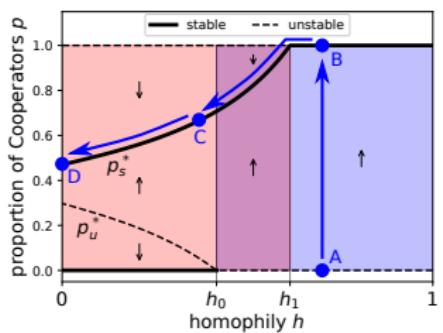
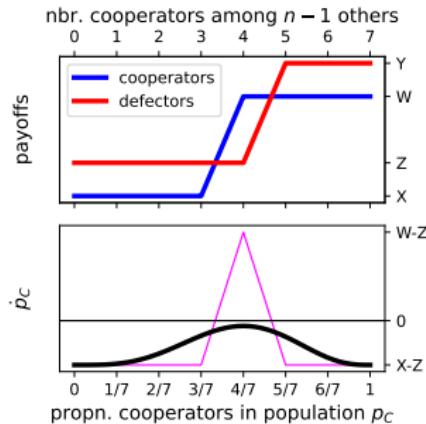
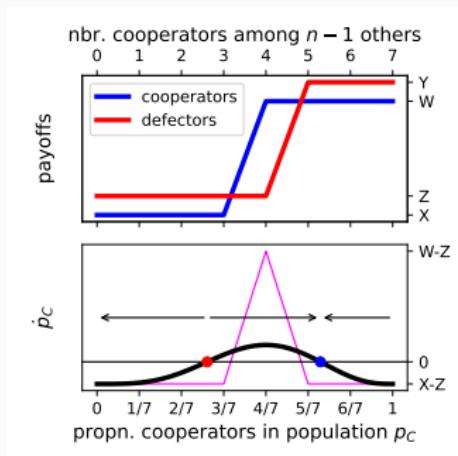


NOTE: can be interpreted in terms of 'matching rules', i.e., strategy homophily

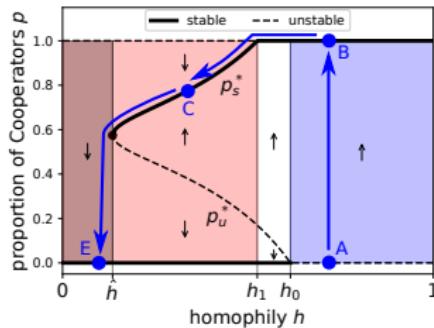
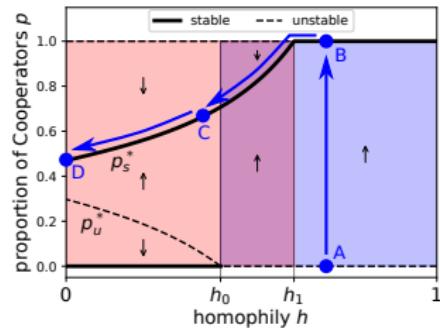
Homophily has decreased over time



Results



Mini summary



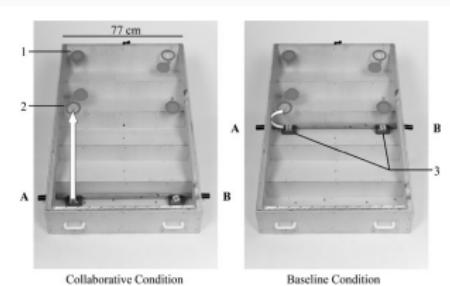
- Cooperation cannot invade a threshold game
 - Also true for sigmoid games (Peña *et al.* (2014) J Theor Biol)
- Can arise through historical homophily
- To persist, either:
 - Parameters such that it can be sustained in a well-mixed population
 - Some degree of homophily maintained

Many discrete strategies

- So far, 2 strategies; natural extension, m strategies
- Discrete strategies:
 - I could have modelled cooperate and defect as *degree of cooperation* — one continuous strategy
 - However, some strategies are naturally discrete
 - e.g., conditioning contribution on the existence of a punishment institution (Garcia and De Monte, 2013)
 - e.g., conditioning contribution on the contributions of others (Takezawa and Price, 2010)
 - typically involve some conceptual ‘leap’
 - Shared intentionality (Genty et al., 2020; Tomasello, 2020):
 - form a collective ‘we’ with a jointly optimised goal
 - make a joint commitment (!?) to the goal
 - coordinate our actions towards achieving it

Commitment

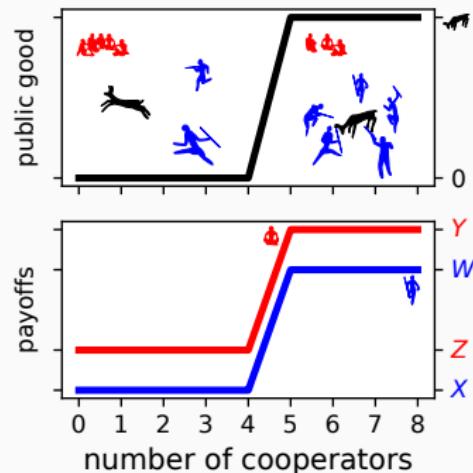
- Commitment is a norm:
 - One should do what one promised
 - Kerr and Kaufman-Gilliland (1994, J Pers Soc Psychol)
- Commitment distinguishes us from other apes
 - In a experimental situation where one individual receives their reward early, 3.5-year-old children will continue contributing until their partner also receives their reward (Hamann et al., 2012), whereas chimpanzees don't distinguish between continuing to help in an existing collaboration versus starting a new one (Greenberg et al., 2010).



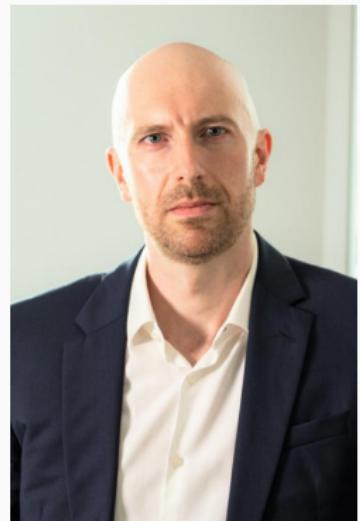
Hamann and Warneken (2012, Child Dev)

Commitment and coordination

- In the threshold game, hunters are a bit stupid
 - Cooperator will run off to do the hunt by themselves
- But people don't really behave this way— they coordinate
 - If we were in this situation, we'd have a conversation
 - That's also how people behave experimentally (e.g., Van de Kragt *et al.* (1983, Am Pol Sci Rev))
- Plus, coordination improves the evolutionary prospects for cooperation!



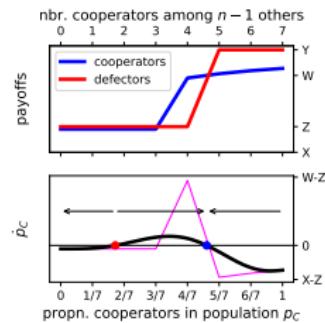
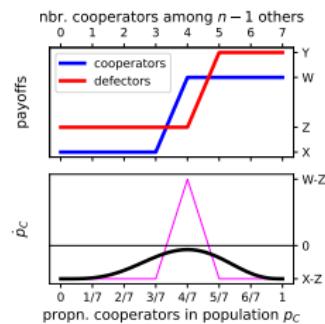
- Newton (2017 Games Econ Behav)
‘shared intentionality’ evolves under
fairly general conditions in a public
goods game



Jonathan Newton

Coordination in a threshold game example

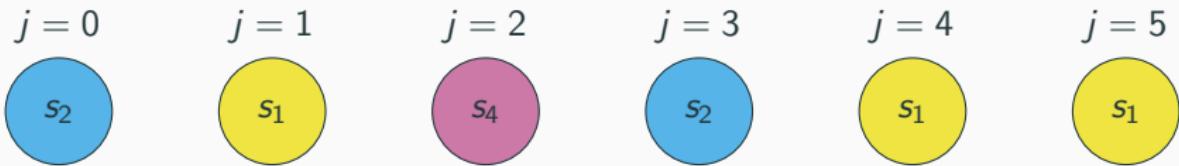
- Threshold game example:
 - Coordinating cooperators draw straws to decide who will contribute
 - The ability to coordinate entails a small cognitive cost ε
- Result:
 - Sustain cooperation where it could not otherwise be sustain
 - Can't invade, but we already know we can overcome this with homophily
- Even works in a linear game!
 - But remember, we know it never makes sense to contribute in the linear game
 - It's true the Defectors can't invade, but what about a type who participates in the lottery but doesn't follow through?
- New strategy: Liars



New notation

- Strategy indicator for player j of $\mathbf{g}_j = \mathbf{e}_x$ means j plays strategy s_x
- Focal player has subscript '0', nonfocals 'nf', and all players 'a'
- Nonfocal strategy indicator is the sum of nonfocal strategies

$$\mathbf{g}_{\text{nf}} = \mathbf{g}_1 + \dots + \mathbf{g}_{n-1}$$



The strategy indicators for each individual are

$$\mathbf{g}_1 = \mathbf{g}_4 = \mathbf{g}_5 = (1, 0, 0, 0), \quad \mathbf{g}_0 = \mathbf{g}_3 = (0, 1, 0, 0), \text{ and } \mathbf{g}_2 = (0, 0, 0, 1).$$

The whole-group strategy composition is: $\mathbf{g}_a = (3, 2, 0, 1)$

The nonfocal strategy composition is: $\mathbf{g}_{\text{nf}} = (3, 1, 0, 1)$

Many strategies

How does a trait change frequency over time?



George Robert Price

$$\Delta p_x = \text{Cov}[G_{0,x}, W_0],$$

dynamics of propn. of s_x

\downarrow

focal's strategy indicator fitness of focal

$$G_{0,x} = \begin{cases} 1 & \text{if focal strategy } s_x, \\ 0 & \text{otherwise.} \end{cases}$$

Many strategies

dynamics of propn. of s_x

$$\Delta p_x = \text{Cov}[G_{0,x}, W_0],$$

$\underbrace{\quad\quad\quad}_{1 \text{ if focal plays } s_x; 0 \text{ otherwise}}$ $\underbrace{\quad\quad\quad}_{\text{fitness of focal}}$

selection strength focal payoff random variable

$$W_0 = s + \overbrace{(1 - s)}^{\text{survival probability}} \frac{1 + \delta}{1 + \delta} \frac{\Pi_0}{\bar{\pi}}$$

$\underbrace{\quad\quad\quad}_{\text{vacancies}}$ $\underbrace{\quad\quad\quad}_{\text{avg. payoff in population}}$

Useful identity: $\text{Cov}[X, aY + b] = a \text{Cov}[X, Y]$

Substituting and rearranging:

strategy indicator focal payoff

$$\Delta p_x \propto \text{Cov}[G_{0,x}, \Pi_0]$$

Other member accounting

$$\Delta p_x \propto \text{Cov} \left[G_{0,x}, \Pi_0 \right]$$

focal's strategy indicator focal payoff

Payoff to the focal individual:

$$\Pi_0 = \sum_{i=1}^m G_{0,i} \pi(\mathbf{e}_i, \mathbf{G}_{\text{nf}})$$

1 if focal plays s_i ; 0 otherwise payoff to s_i -player

Useful identity: $\text{Cov}[X, Y] = \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]$

$$\Delta p_x = \mathbb{E} \left[G_{0,x} \pi(\mathbf{e}_x, \mathbf{G}_{\text{nf}}) \right] - p_x \sum_{i=1}^m \mathbb{E} \left[G_{0,i} \pi(\mathbf{e}_i, \mathbf{G}_{\text{nf}}) \right]$$

↑ nonfocal strategy composition ↑

Other member accounting

25

nonfocal strategy composition

$$\Delta p_x = \mathbb{E} \left[G_{0,x} \pi(\mathbf{e}_x, \mathbf{G}_{\text{nf}}) \right] - p_x \sum_{i=1}^m \mathbb{E} \left[G_{0,i} \pi(\mathbf{e}_i, \mathbf{G}_{\text{nf}}) \right]$$

The possible values of \mathbf{G}_{nf} are: $\mathcal{G}_{\text{nf}} = \{\mathbf{g}_{\text{nf}} \in \mathbb{N}_{\geq 0}^m \mid \sum_{i=1}^m g_{\text{nf},i} = n - 1\}$

So the expectations are:

$$\begin{aligned} \mathbb{E}[G_{0,i}\pi(\mathbf{e}_i, \mathbf{G}_{\text{nf}})] &= \sum_{\mathbf{g}_{\text{nf}} \in \mathcal{G}_{\text{nf}}} \pi(\mathbf{e}_i, \mathbf{g}_{\text{nf}}) \mathbb{P}[\mathbf{G}_0 = \mathbf{e}_i, \mathbf{G}_{\text{nf}} = \mathbf{g}_{\text{nf}}] \\ &= \sum_{\mathbf{g}_{\text{nf}} \in \mathcal{G}_{\text{nf}}} \pi(\mathbf{e}_i, \mathbf{g}_{\text{nf}}) \underbrace{\mathbb{P}[\mathbf{G}_0 = \mathbf{e}_i]}_{p_i} \mathbb{P}[\mathbf{G}_{\text{nf}} = \mathbf{g}_{\text{nf}} \mid \mathbf{G}_0 = \mathbf{e}_i] \\ &= p_i \underbrace{\sum_{\mathbf{g}_{\text{nf}} \in \mathcal{G}_{\text{nf}}} \pi(\mathbf{e}_i, \mathbf{g}_{\text{nf}})}_{\bar{\pi}_i} \mathbb{P}[\mathbf{G}_{\text{nf}} = \mathbf{g}_{\text{nf}} \mid \mathbf{G}_0 = \mathbf{e}_i] \end{aligned}$$

Recover replicator eqn: $\Delta p_x \propto p_x (\bar{\pi}_x - \sum_{i=1}^m p_i \bar{\pi}_i) = p_x (\bar{\pi}_x - \bar{\pi})$.

But $\mathbb{P}[\mathbf{G}_{\text{nf}} = \mathbf{g}_{\text{nf}} \mid \mathbf{G}_0 = \mathbf{e}_i]$ is not obvious.

$$\Delta p_x \propto \text{Cov} \left[G_{0,x}, \Pi_0 \right]$$

strategy indicator focal payoff

But this time, let's focus on the whole-group distribution. Idea: draw a group at random, then draw a focal individual.

new payoff fnc wrt whole-group strategy composition

$$\Pi_0 = \sum_{i=1}^m G_{0,i} \hat{\pi}(\mathbf{e}_i, \mathbf{G}_a)$$

Using a similar method to before involving covariance identities and re-arranging, we obtain

$$\Delta p_x = \sum_{\mathbf{g}_a \in \mathcal{G}_a} \left(\frac{g_{a,x}}{n} \hat{\pi}(\mathbf{e}_x, \mathbf{g}_a) - p_x \sum_{i=1}^m \frac{g_{a,i}}{n} \hat{\pi}(\mathbf{e}_i, \mathbf{g}_a) \right) \mathbb{P}[\mathbf{G}_a = \mathbf{g}_a]$$

prob. of whole-group strategy composition

Whole-group accounting

$$\Delta p_x = \sum_{\mathbf{g}_a \in \mathcal{G}_a} \left(\frac{g_{a,x}}{n} \hat{\pi}(\mathbf{e}_x, \mathbf{g}_a) - p_x \sum_{i=1}^m \frac{g_{a,i}}{n} \hat{\pi}(\mathbf{e}_i, \mathbf{g}_a) \right) \mathbb{P}[\mathbf{G}_a = \mathbf{g}_a]$$

prob. of whole-group strategy composition

- The probability $\mathbb{P}[\mathbf{G}_a = \mathbf{g}_a]$ depends on the family structure of the group
- Two rules for allocating individuals to families:
 - Individuals with different strategies → different families
 - Individuals with the same strategy → same or different families
- Introduce two new objects:
 1. z : strategywise family-size distribution
 2. y : group's family-size distribution

Strategywise family-size distribution

Example:

Consider the group illustrated below:



Each strategy's family-size distribution is

$$\mathbf{z}_1 = (1, 1, 0, 0, 0, 0), \quad \mathbf{z}_2 = (0, 1, 0, 0, 0, 0),$$

$$\mathbf{z}_3 = (0, 0, 0, 0, 0, 0), \quad \mathbf{z}_4 = (1, 0, 0, 0, 0, 0).$$

So the group's strategywise family-size distribution is

$$\mathbf{z} = ((1, 1, 0, 0, 0, 0), (0, 1, 0, 0, 0, 0), (0, 0, 0, 0, 0, 0), (1, 0, 0, 0, 0, 0)),$$

The group's family-size distribution is

$$\mathbf{y} = \text{sum}(\mathbf{z}) = (2, 2, 0, 0, 0, 0).$$

Probability of whole-group strategy composition

Let \mathcal{Z}_{g_a} be the set of all possible strategywise family-size distributions z consistent with the strategy distribution g_a . Then

$$\begin{aligned}\mathbb{P}[G_a = g_a] &= \sum_{z \in \mathcal{Z}_{g_a}} \mathbb{P}[Z = z] = \sum_{z \in \mathcal{Z}_{g_a}} \mathbb{P}[Y = y] \mathbb{P}[Z = z \mid Y = y], \\ &= \sum_{z \in \mathcal{Z}_{g_a}} F_y \mathbb{P}[Z = z \mid Y = y],\end{aligned}$$

↑ get from homophilic group-formation model

Probability of strategywise family-size distribution:

nbr. ways to order strategies across families

$$\mathbb{P}[Z = z \mid Y = y] = C(z) A(z, p)$$

↑ count of multiset permutations ↑ prob. families' strategies

$$A(z, p) = \prod_{i=1}^m p_i^{\|z_i\|}$$

↑ nbr. families pursuing strategy s_i

Whole-group accounting

Bringing it all together:

$$\Delta p_x \propto \sum_{\mathbf{g}_a \in \mathcal{G}_a} \left(\frac{g_{a,x}}{n} \hat{\pi}(\mathbf{e}_x, \mathbf{g}_a) - p_x \sum_{i=1}^m \frac{g_{a,i}}{n} \hat{\pi}(\mathbf{e}_i, \mathbf{g}_a) \right) \left(\sum_{z \in \mathcal{Z}_{\mathbf{g}_a}} C(z) A(z, \mathbf{p}) F_{\text{sum}(z)} \right)$$

prob. focal pursues s_x over strategywise family-sizes

↑ sum over group strategy compositions

- Not as intuitive as the traditional replicator equation
 - $\Delta p_x \propto p_x (\bar{\pi}_x - \bar{\pi})$
- Useful from computational perspective because we've split homophily calculations off from strategy identity
 - Components of $C(z)$ and $A(z, \mathbf{p})$ can be precalculated and stored
 - But computational explosion as group size and number of strategies increase
- Now it's clearer how to calculate $\mathbb{P}[\mathbf{G}_{\text{nf}} = \mathbf{g}_{\text{nf}} \mid \mathbf{G}_0 = \mathbf{e}_i]$

Payoff transformation 2 players

- Payoff transformation
 - Transform payoff matrix by writing as a function of relatedness
 - Solve the well-mixed dynamics
- Example: 2 players many strategies
 - Define matrix with elements equal to payoffs $a_{i,j} = \pi(\mathbf{e}_i, \mathbf{e}_j)$
 - Replicator dynamics in matrix form:
$$\dot{p}_i = p_i(\bar{\pi}_i - \bar{\pi}) = p_i((A\mathbf{p})_i - \mathbf{p}^T A\mathbf{p})$$
 - For example, we can write out $\bar{\pi} = A\mathbf{p}$ in detail

$$\begin{pmatrix} \bar{\pi}_1 \\ \vdots \\ \bar{\pi}_i \\ \vdots \\ \bar{\pi}_m \end{pmatrix} = \underbrace{\begin{pmatrix} & \xrightarrow{\text{nonfocal's strategy}} \\ \text{focal's strategy} \downarrow & \end{pmatrix}}_{\text{matrix } A} \begin{pmatrix} a_{1,1} & \dots & a_{1,m} \\ \vdots & & \vdots \\ a_{i,1} & \dots & a_{i,m} \\ \vdots & & \vdots \\ a_{m,1} & \dots & a_{m,m} \end{pmatrix} \begin{pmatrix} p_1 \\ \vdots \\ p_i \\ \vdots \\ p_m \end{pmatrix} = \begin{pmatrix} a_{1,1}p_1 + \dots + a_{1,m}p_m \\ \vdots \\ a_{i,1}p_1 + \dots + a_{i,m}p_m \\ \vdots \\ a_{m,1}p_1 + \dots + a_{m,m}p_m \end{pmatrix}$$

Payoff transformation 2 players

- Example: 2 players many strategies continued
 - We are up to here; $\bar{\pi} = A\mathbf{p}$:

$$\begin{pmatrix} \bar{\pi}_1 \\ \vdots \\ \bar{\pi}_i \\ \vdots \\ \bar{\pi}_m \end{pmatrix} = \begin{pmatrix} a_{1,1} & \dots & a_{1,m} \\ \vdots & & \vdots \\ a_{i,1} & \dots & a_{i,m} \\ \vdots & & \vdots \\ a_{m,1} & \dots & a_{m,m} \end{pmatrix} \begin{pmatrix} p_1 \\ \vdots \\ p_i \\ \vdots \\ p_m \end{pmatrix} = \begin{pmatrix} a_{1,1}p_1 + \dots + a_{1,m}p_m \\ \vdots \\ a_{i,1}p_1 + \dots + a_{i,m}p_m \\ \vdots \\ a_{m,1}p_1 + \dots + a_{m,m}p_m \end{pmatrix}$$

- Now consider the game under dyadic relatedness $\theta_{2 \rightarrow 1}$

$$B = \theta_{2 \rightarrow 1} \underbrace{\begin{pmatrix} a_{1,1} & \dots & a_{1,1} \\ \vdots & & \vdots \\ a_{i,i} & \dots & a_{i,i} \\ \vdots & & \vdots \\ a_{m,m} & \dots & a_{m,m} \end{pmatrix}}_{i \text{ matched with } i \text{ with prob. } \theta_{2 \rightarrow 1}} + (1 - \theta_{2 \rightarrow 1}) \underbrace{\begin{pmatrix} a_{1,1} & \dots & a_{1,m} \\ \vdots & & \vdots \\ a_{i,1} & \dots & a_{i,m} \\ \vdots & & \vdots \\ a_{m,1} & \dots & a_{m,m} \end{pmatrix}}_{i \text{ matched with random with prob. } 1 - \theta_{2 \rightarrow 1}}$$

- Dynamics of A with homophily \equiv dynamics of B well-mixed

$$\dot{p}_i = p_i((B\mathbf{p})_i - \mathbf{p}^T B\mathbf{p})$$

Payoff transformation n players

Seeking a solution to:

$$B = \begin{array}{c} \text{player 2} \\ \xrightarrow{\text{focal player 0}} \\ \left[\begin{array}{cccc} b_{m,1,1} & b_{m,1,2} & \dots & b_{m,1,m} \\ b_{m,2,1} & b_{m,2,2} & \dots & b_{m,2,m} \\ \vdots & \vdots & & \vdots \\ b_{2,1,1} & b_{2,1,2} & \dots & b_{2,1,m} \\ b_{2,2,1} & b_{2,2,2} & \dots & b_{2,2,m} \\ \vdots & \vdots & \ddots & \vdots \\ b_{1,1,1} & b_{1,1,2} & \dots & b_{1,1,m} \\ b_{1,2,1} & b_{1,2,2} & \dots & b_{1,2,m} \\ \vdots & \vdots & \ddots & \vdots \\ b_{1,m,1} & b_{1,m,2} & \dots & b_{1,m,m} \end{array} \right] \\ \text{player 1} \end{array}$$

Payoff transformation n players

$$B = \begin{bmatrix} & & & & \text{player 2} \\ & b_{m,1,1} & b_{m,1,2} & \dots & b_{m,1,m} \\ \text{focal player 0} & b_{m,n+1} & b_{m,n+2} & \dots & b_{m,2,m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ & b_{m,m} & & & \vdots \\ \hline & b_{2,1,1} & b_{2,1,2} & \dots & b_{2,1,m} \\ b_{2,n+1} & b_{2,n+2} & \dots & b_{2,2,m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ & b_{2,m} & & & \vdots \\ \hline & b_{1,1,1} & b_{1,1,2} & \dots & b_{1,1,m} \\ b_{1,2,1} & b_{1,2,2} & \dots & b_{1,2,m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ & b_{1,m} & & & \vdots \\ \hline & b_{1,m,1} & b_{1,m,2} & \dots & b_{1,m,m} \end{bmatrix}$$

$$b_u = \sum_{q \vdash n} F_q \left(\sum_{q_0 \in q} \frac{q_0}{n|\mathcal{J}_{q_0, q}|} \left(\sum_{j \in \mathcal{J}_{q_0, q}} a_{uj} \right) \right)$$

get from group-formation model

Code to calculate it on Github:

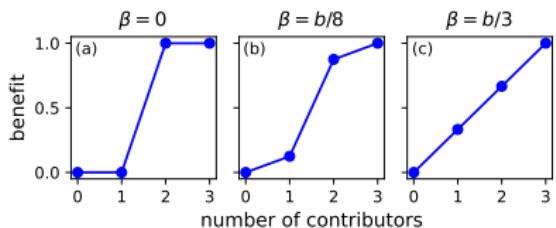
1. Numerically: TransmatBase class functions/transmat_base.py.
2. Symbolically: functions/symbolic_transformed.py.

But why would you want to do this?

- B is expensive to calculate, but matrix multiplication is optimised, can be worth the trade-off when finding steady states
- Use maths from well-mixed case, e.g., Jorge Peña's analysis techniques (example in appendix)

Coordinated cooperation

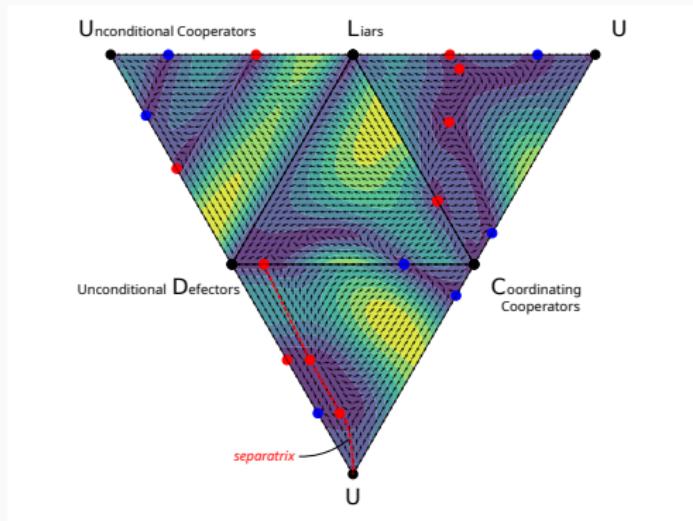
- Game with 4 strategies:
 1. D : unconditional Defector, never contributes
 2. C : Coordinating cooperator, hold lottery, follow through if chosen
 - Nbr. contributors τ = threshold, or inflection point if sigmoid
 3. L : Liar, participate in lottery, never contributes
 4. U : Unconditional cooperator, always contributes
- C and L pay cognitive cost ε regardless of game outcome
- U and C pay contribution cost c if contributing
- Explore the range from linear to threshold game



Example 3-player - symbolic analysis

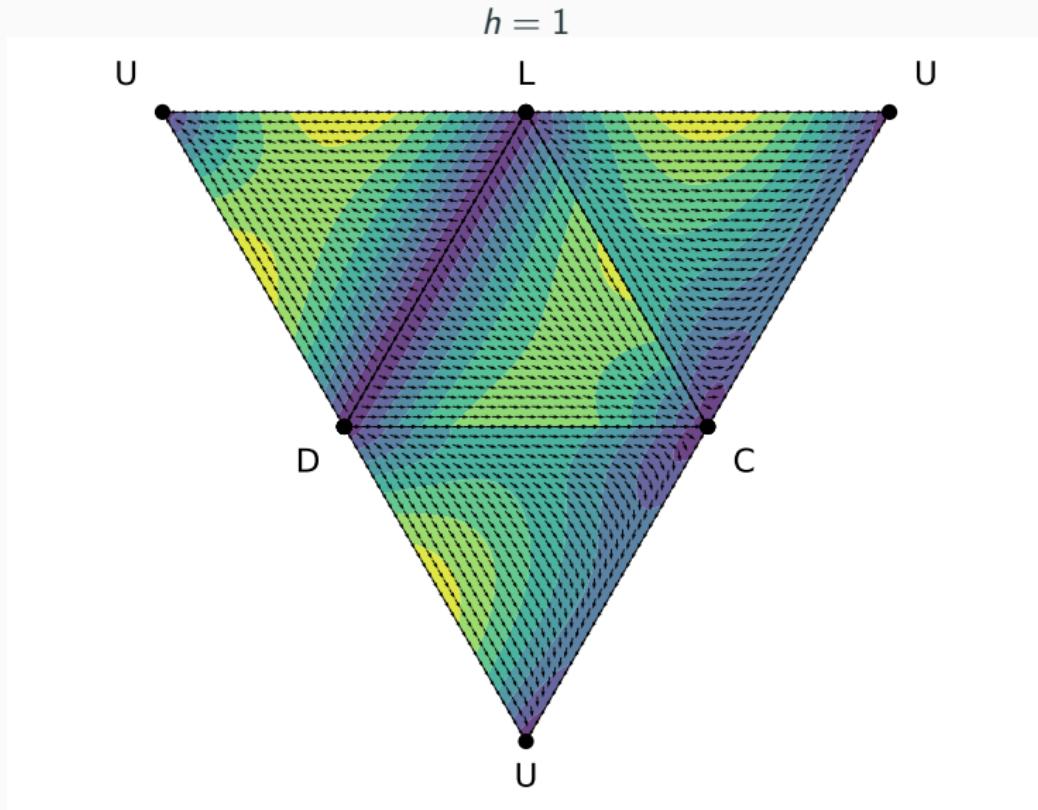
Example 8-player - numerical analysis

How to read results

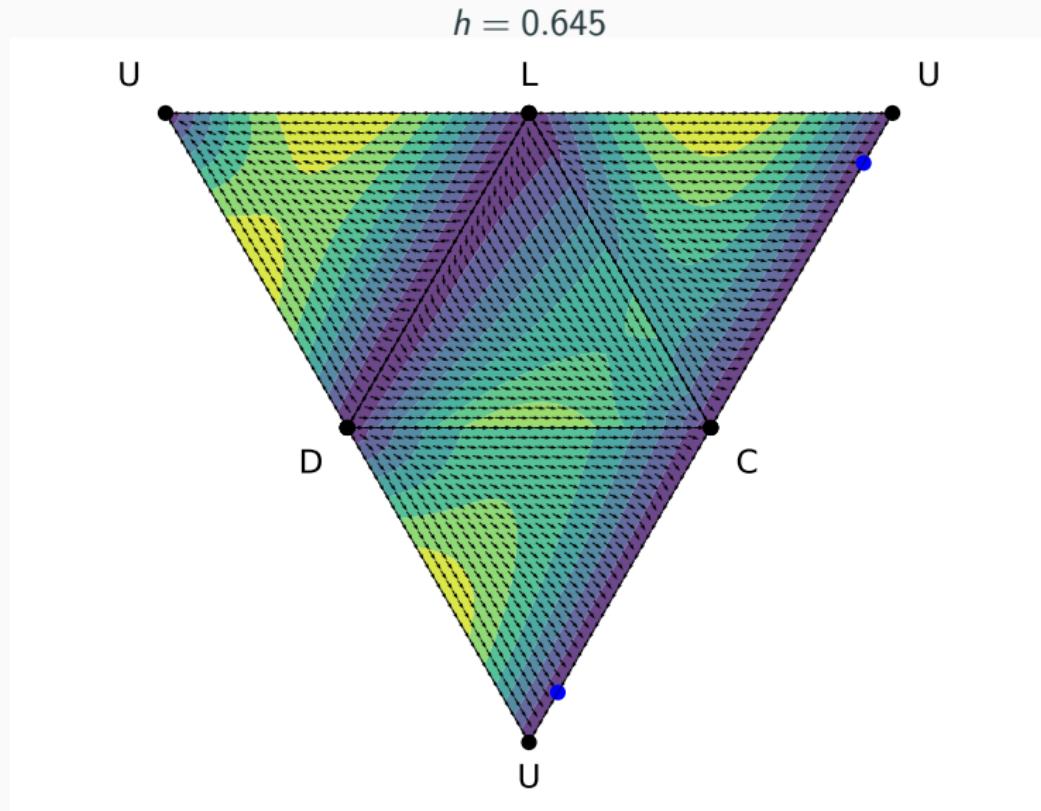


- Evolutionary dynamics for a given homophily level h
 - Dynamics inside a triangular pyramid
 - The points represent a population with just one strategy, lines 2 strategies, triangles 3
 - Blue points are stable in that dimension, red points unstable

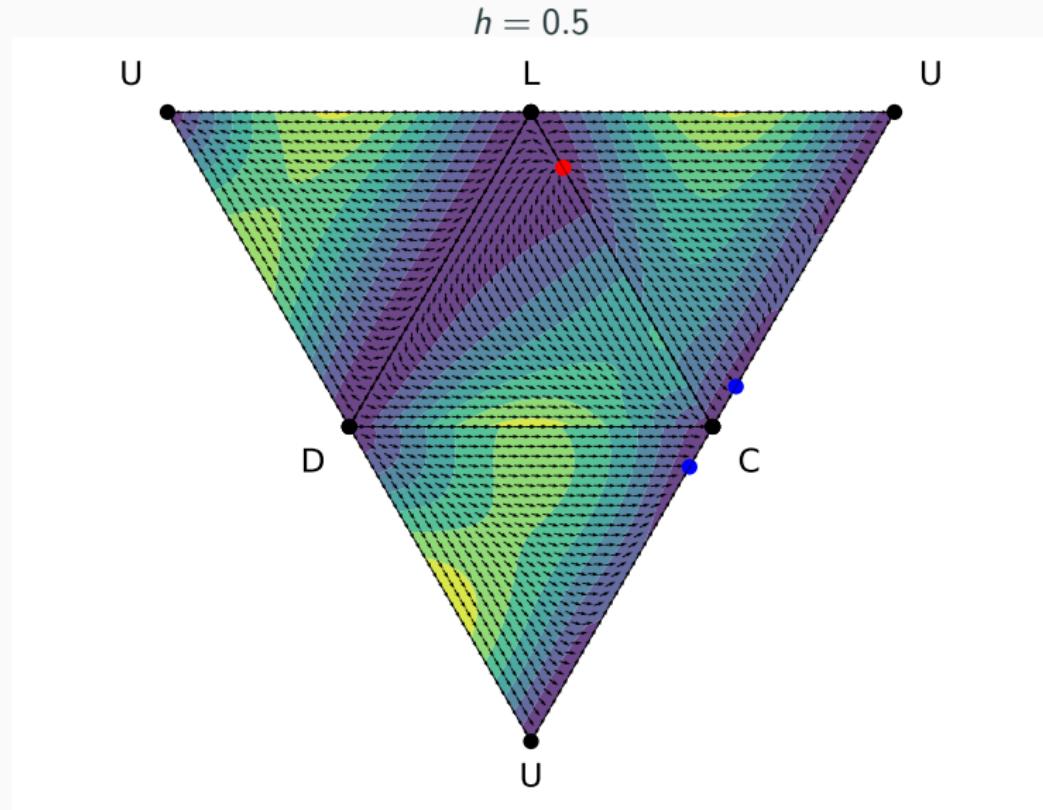
Results 1: fairly nonlinear benefits function



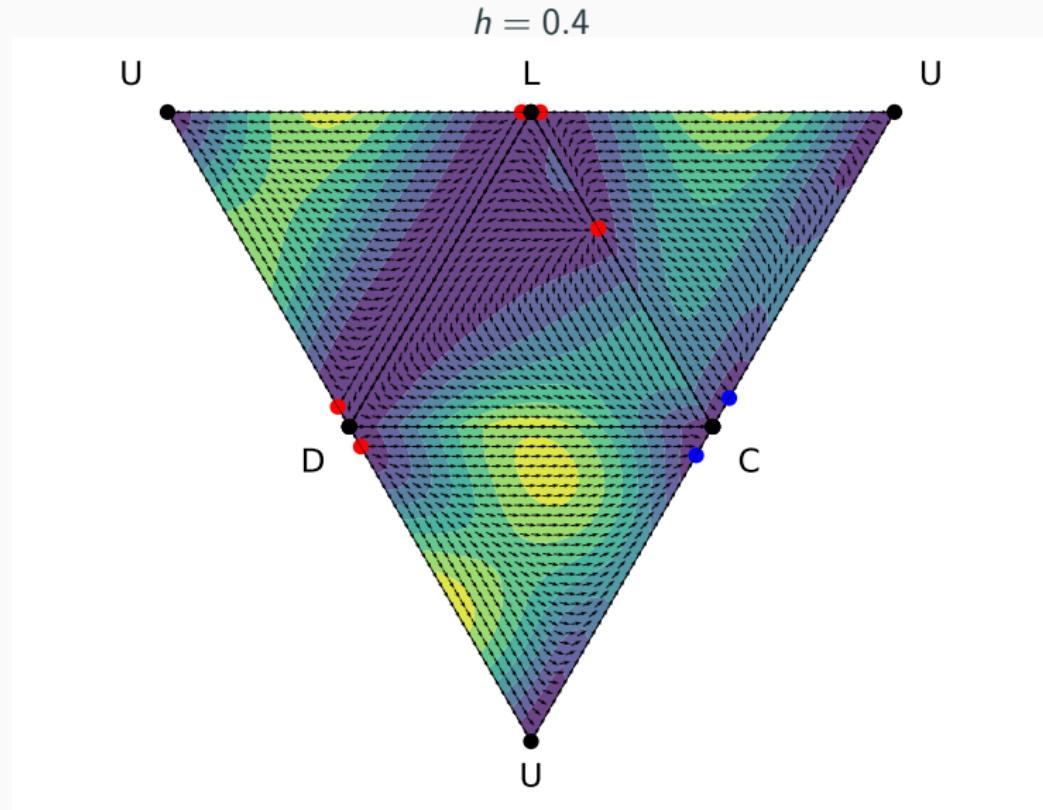
Results 1: fairly nonlinear benefits function



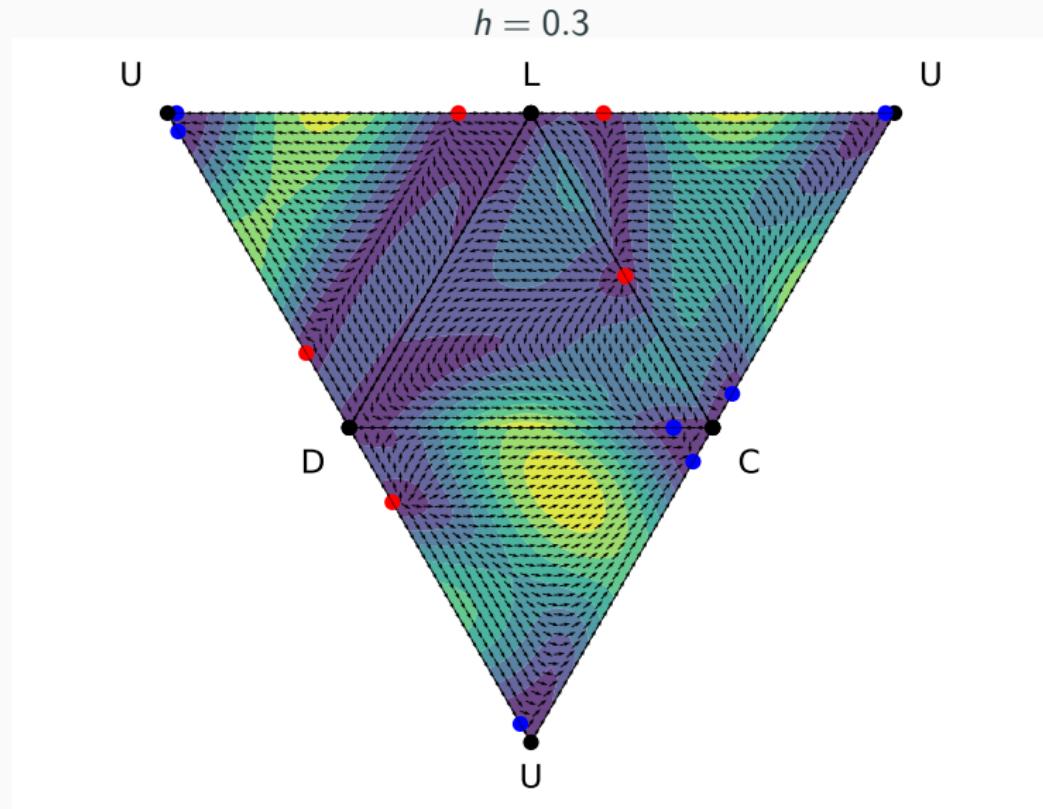
Results 1: fairly nonlinear benefits function



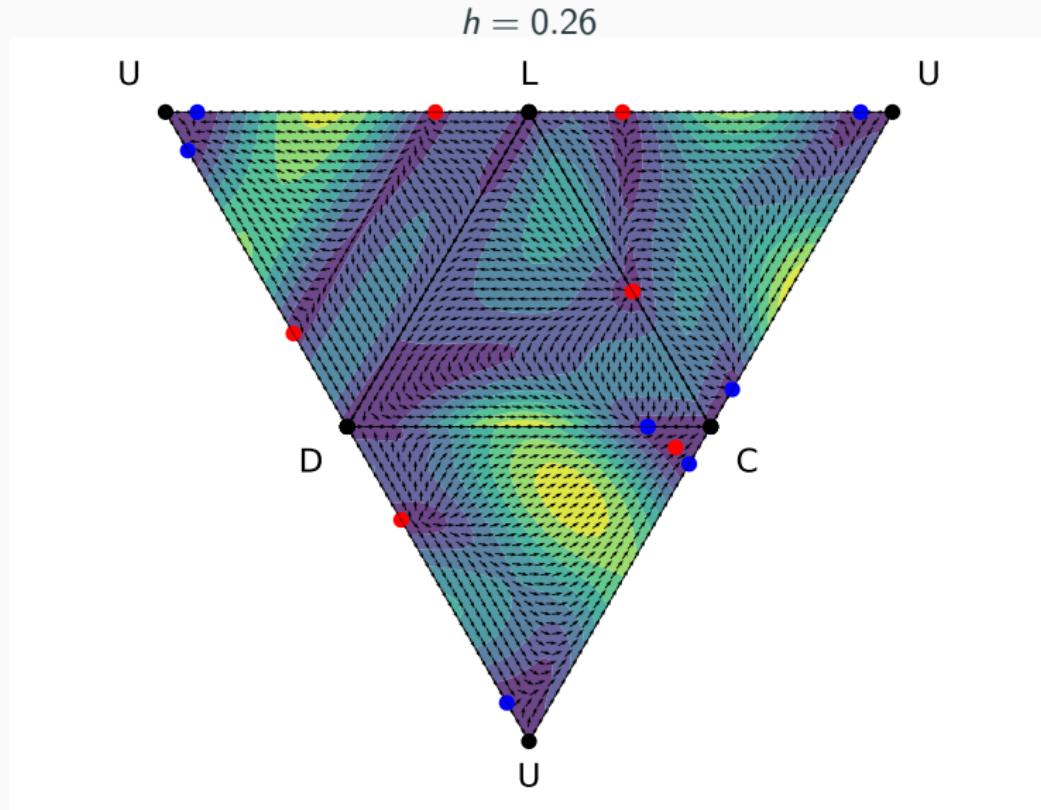
Results 1: fairly nonlinear benefits function



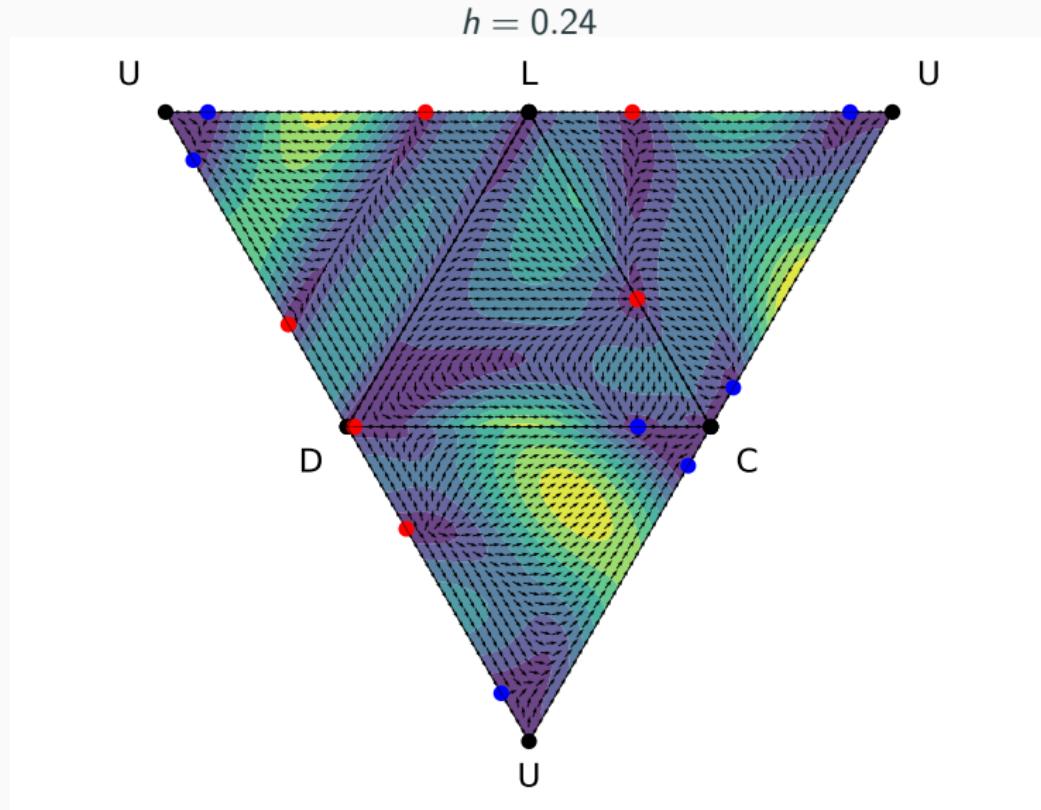
Results 1: fairly nonlinear benefits function



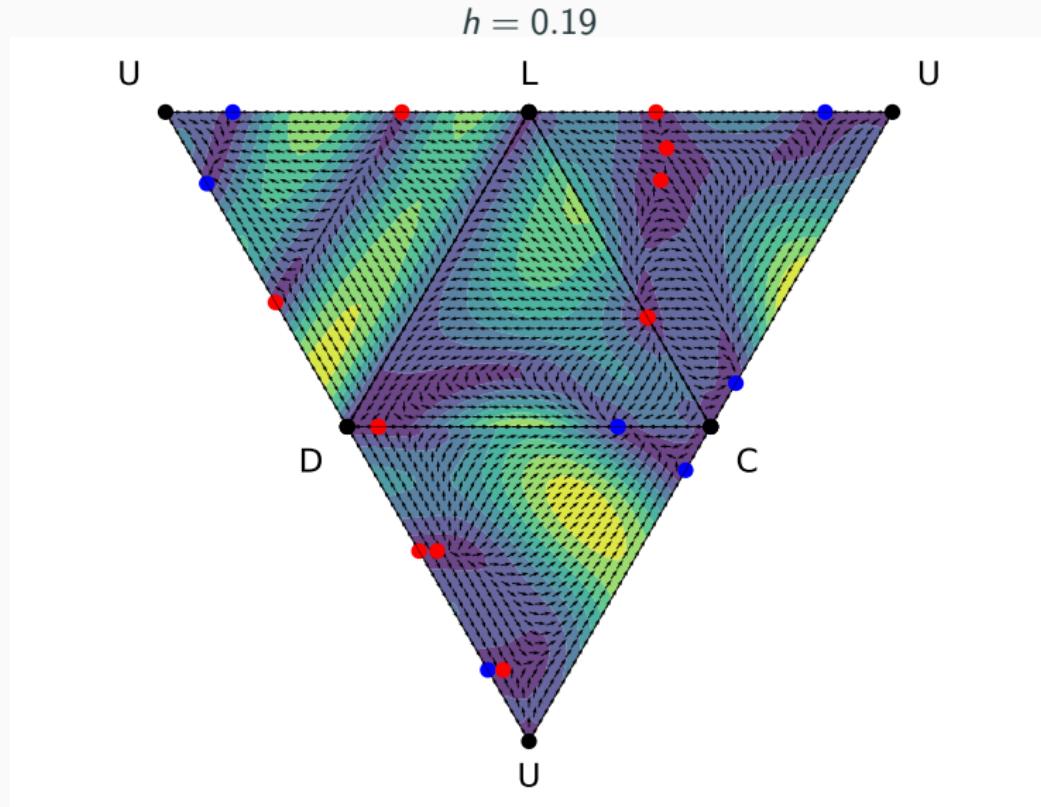
Results 1: fairly nonlinear benefits function



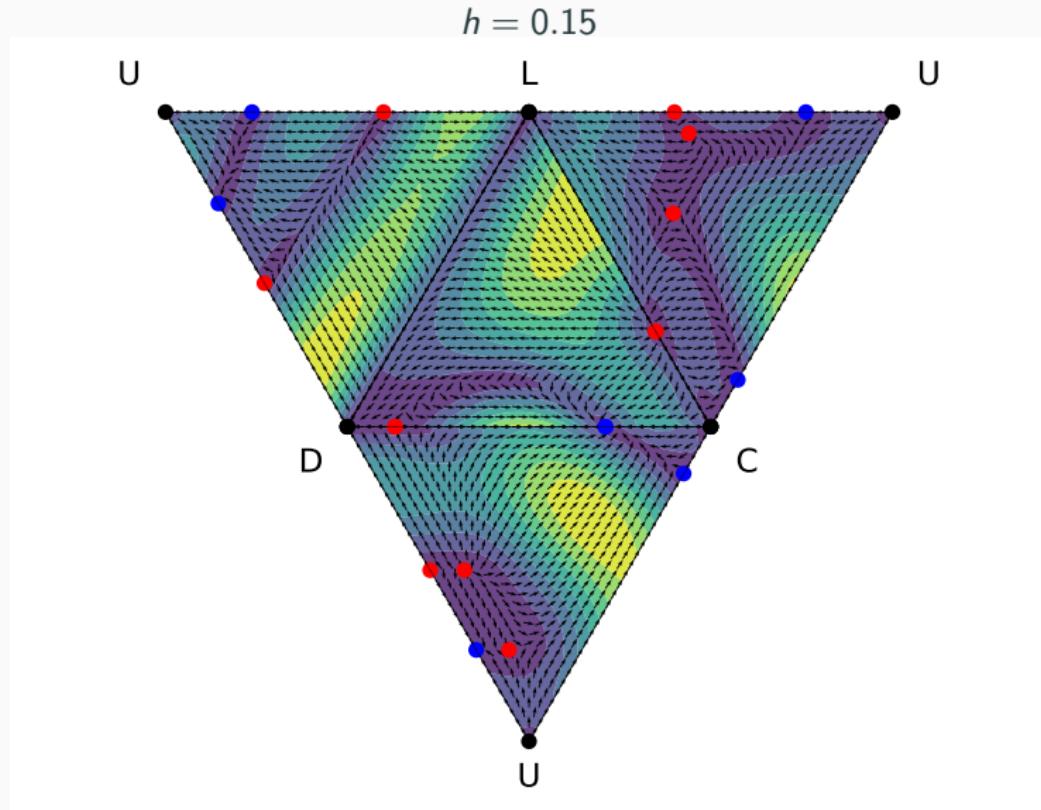
Results 1: fairly nonlinear benefits function



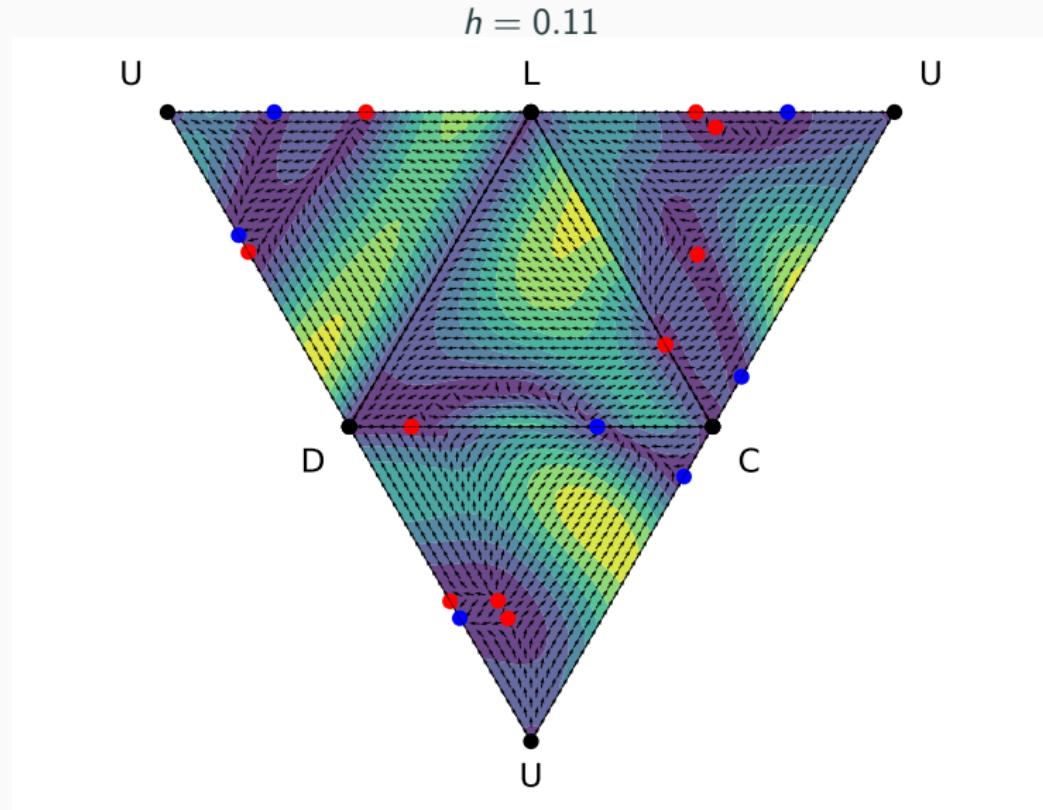
Results 1: fairly nonlinear benefits function



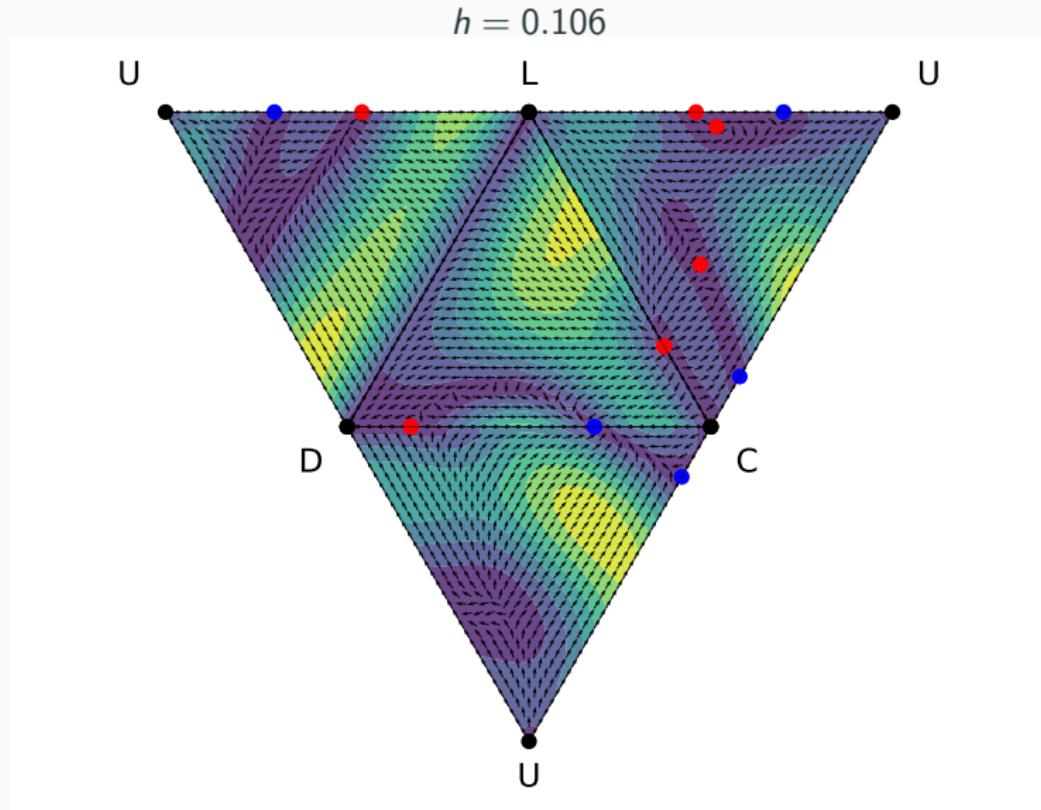
Results 1: fairly nonlinear benefits function



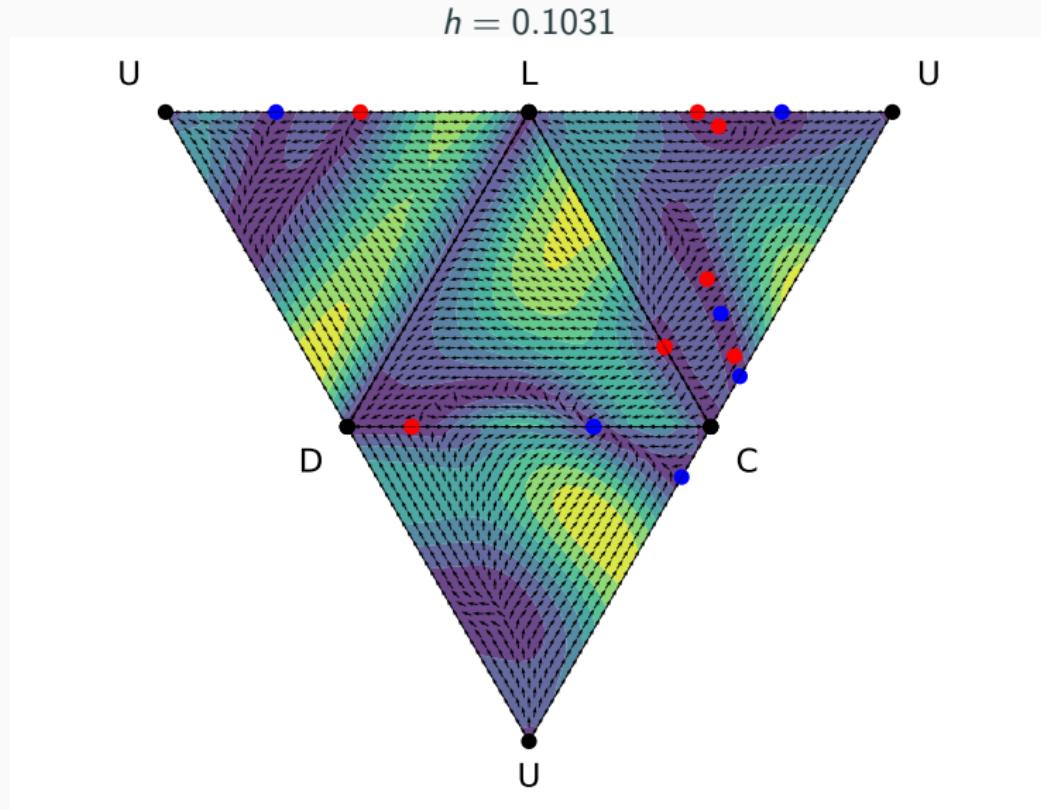
Results 1: fairly nonlinear benefits function



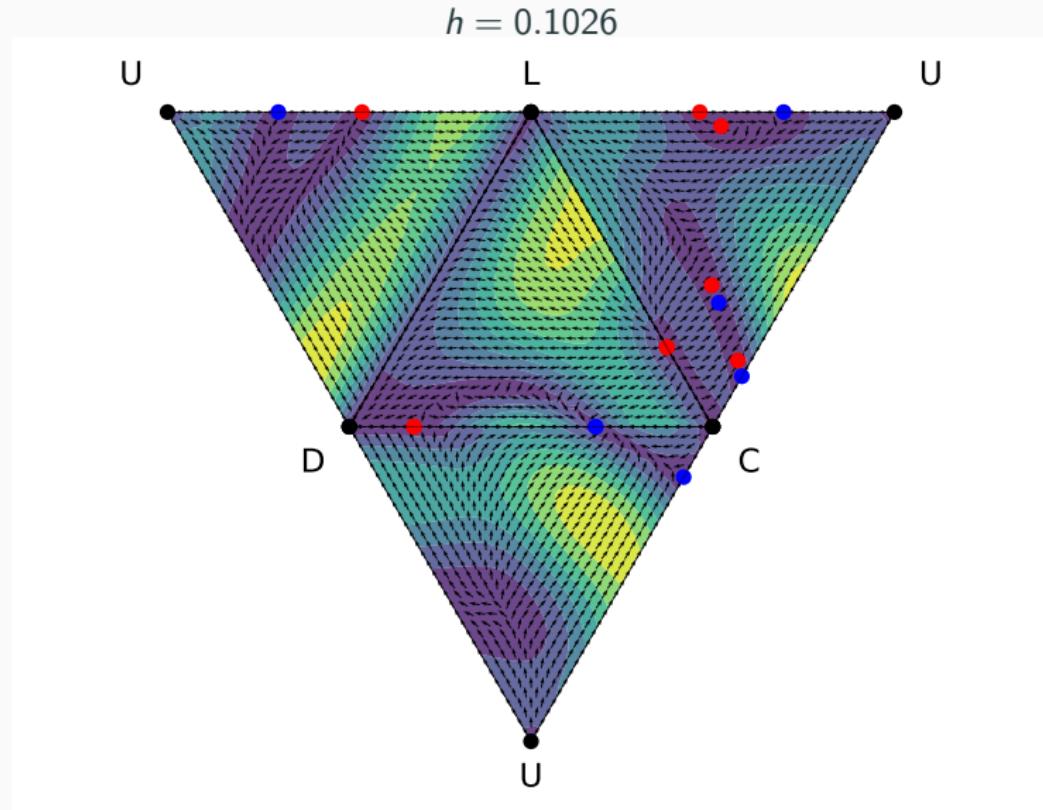
Results 1: fairly nonlinear benefits function



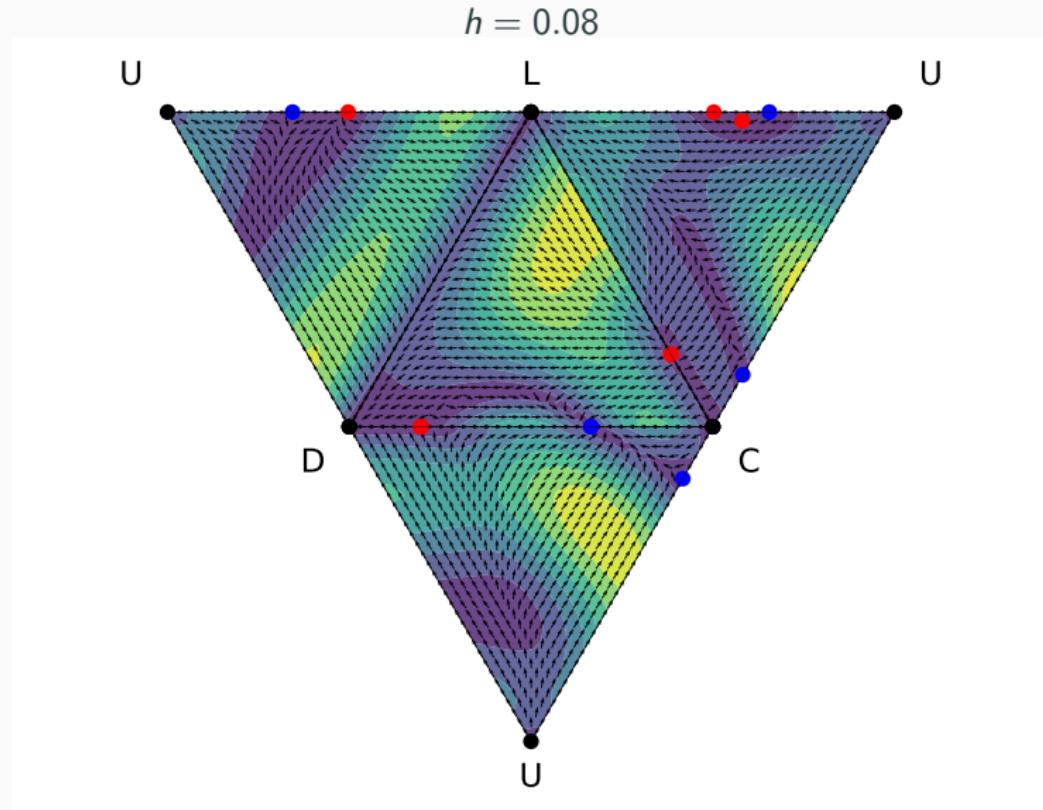
Results 1: fairly nonlinear benefits function



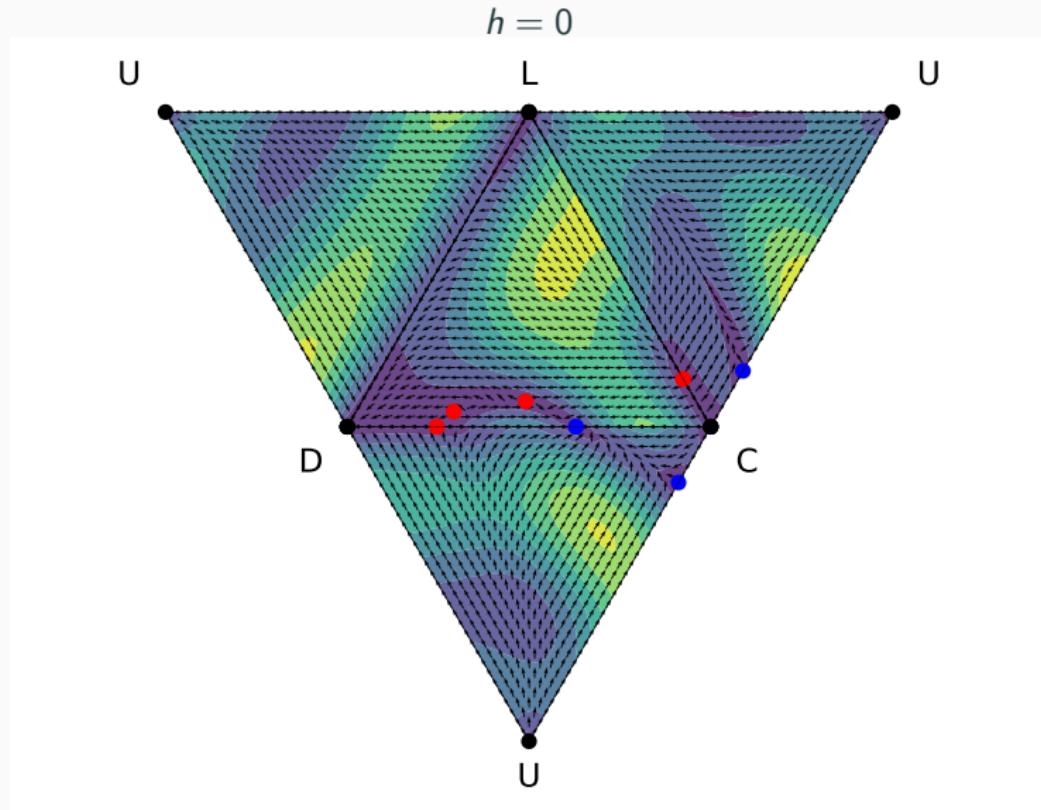
Results 1: fairly nonlinear benefits function



Results 1: fairly nonlinear benefits function

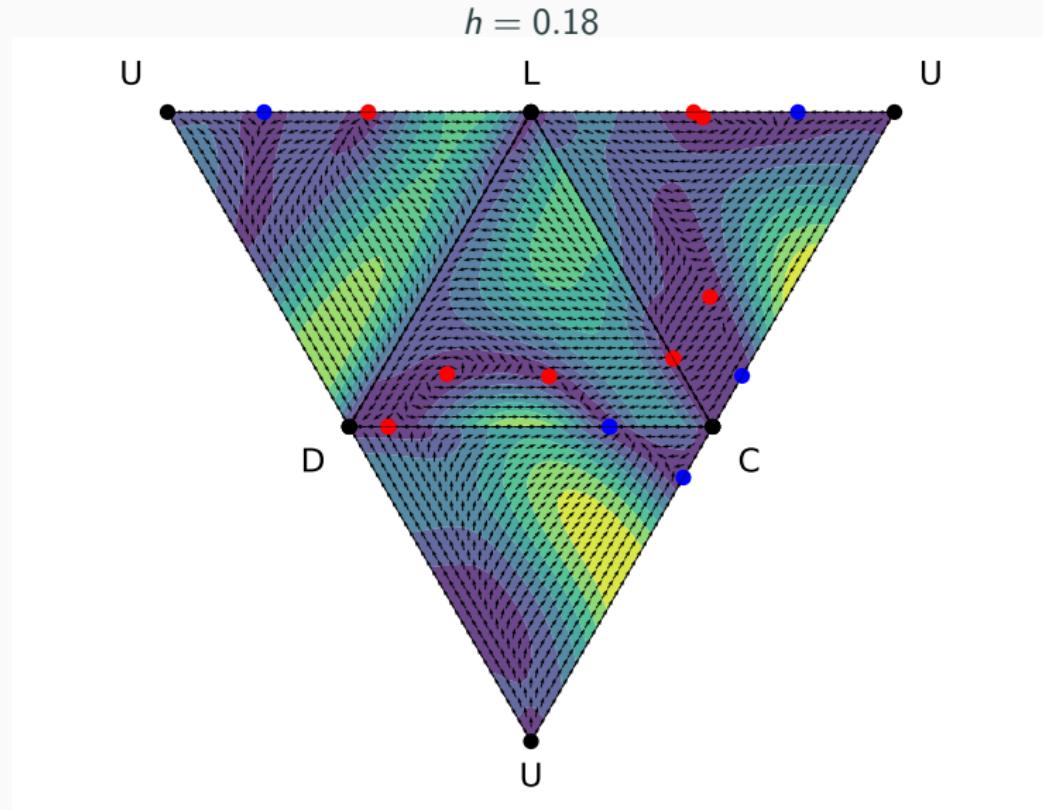


Results 1: fairly nonlinear benefits function



Results 2: more linear benefits function

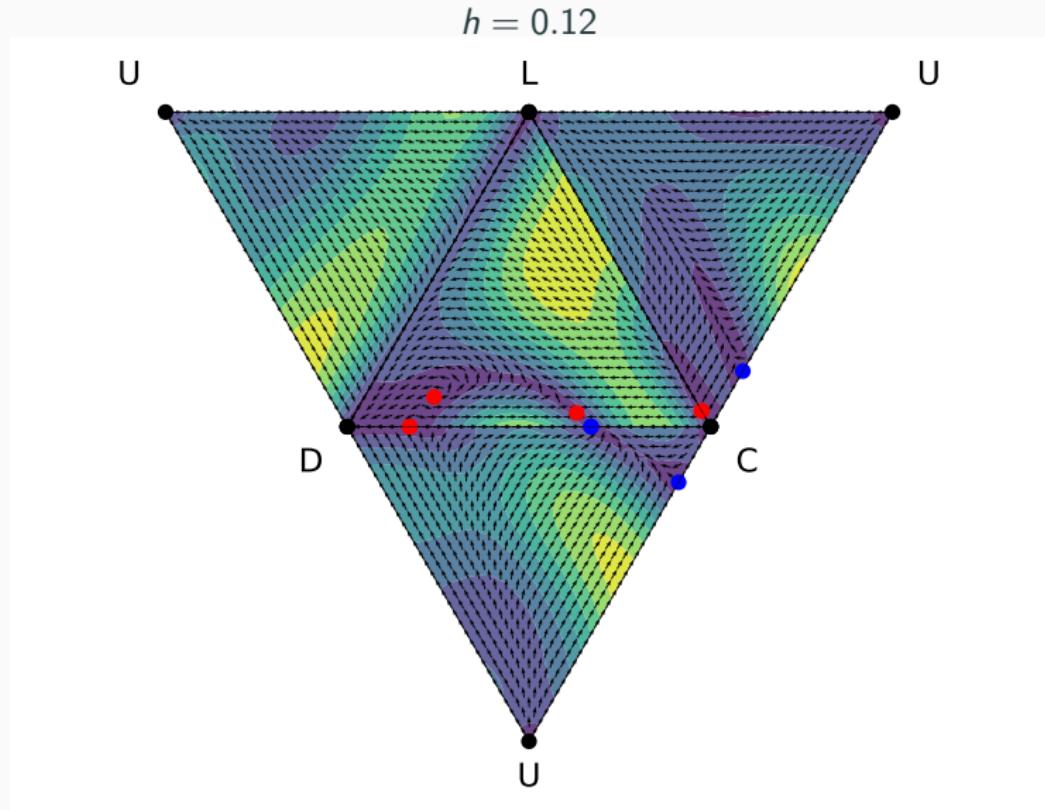
40



45

Results 2: more linear benefits function

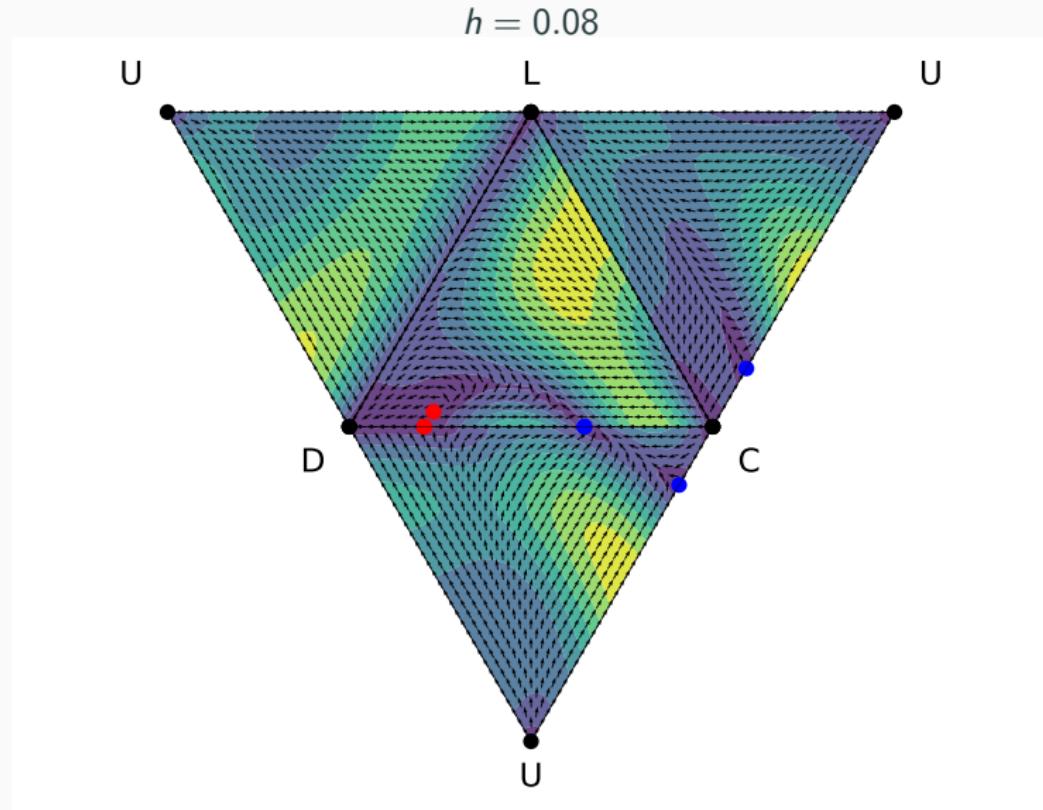
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45

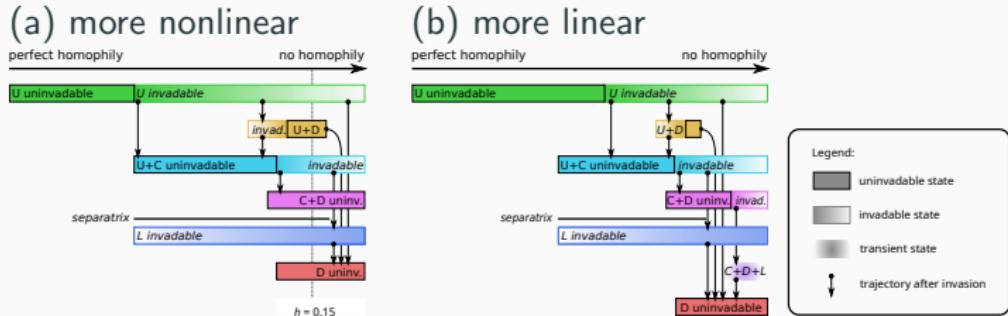
Results 2: more linear benefits function

40



45

Results summary



- Commitment beyond self-interest?
 - Interdependence is not enough
 - Enough experimentally, e.g., children's task
 - But evolutionary, must be some kind of 'mistake'
 - Reputation?
 - Players with good reputation only cooperate with others with good reputation
 - A kind of strategy-based homophily
 - Prisoner's Dilemma model by Kellner & Han (2025, arXiv)
 - Good reputation obtained by commitment adherence

Summary

- Built a mathematical framework to combine discrete-strategy group games with kin selection (or ‘matching rules’)
- Past homophily can explain how cooperation first arose
- Conditioned, coordinated cooperation allows cooperation to persist where unconditional cooperation cannot
- Commitment beyond self-interest either requires some level of homophily or some other mechanism

Thank you!

github.com/nadiahpk

Nadia Pardede Kristensen
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Models for ecology, evolution, and cooperation. Code or it didn't happen.
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#metacommunity #author #quantitative_modelling #learning #unfolded_extinctions

Oct 15, 2024
Check if an iterated Prisoner's Dilemma strategy is a subgame perfect Nash equilibrium

I recently read a paper by Kristensen et al. (2022). The effect of environmental information on evolution of cooperation in stochastic games, in particular, on iterated Prisoner's Dilemma games. In particular, the problem they investigated admits 64 possible strategies in the possible strategy space, and was solved in writing a script that could automate the analysis. I eventually landed on (Github rep) and used a combination of SymPy, NetworkX, Segenfish, the 23 Thomas Prover, and PyEDA for Boolean minimization, but I think my approach could be improved. Background to the paper Kristensen et al.

Filed under: [Nadiah](#)

Oct 15, 2024
A summary of Richard Joyce's 'The Evolution of Morality'

Moral. To learn more about the evolution of cooperation from a philosopher's perspective, I recently read Richard Joyce's book The Evolution of Morality. Joyce's book makes the case for the evolutionary debunking argument, which holds that moral beliefs are the product of evolutionary processes rather than tracking moral truths. While evolution has equipped us with the capacity for moral judgment, this doesn't necessarily mean that our moral beliefs are true or justified. Instead, our moral sense evolved because it was useful for our ancestors' survival and reproduction, regardless of whether moral facts actually exist. Joyce bases his arguments on the well-known