

Mariehamn, Åland

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Statistical mechanics of distributed information systems

Evolutionary games in finite populations



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Outline

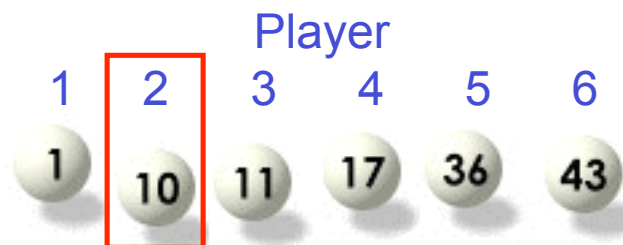
- Introduction to evolutionary games
- Stochastic game dynamics in finite populations
- From finite to infinite populations
- The evolution of cooperation
- The emergence of punishment

Beauty contest



- N players choose a number between 1 and 100
- Calculate the average a
- The number closest to $a/2$ wins

Example



$$N = 6$$
$$a \approx 20$$
$$a/2 \approx 10$$

What is the optimal choice?

● Classical game theory

- Strategic interactions of humans
- Assumption of rationality

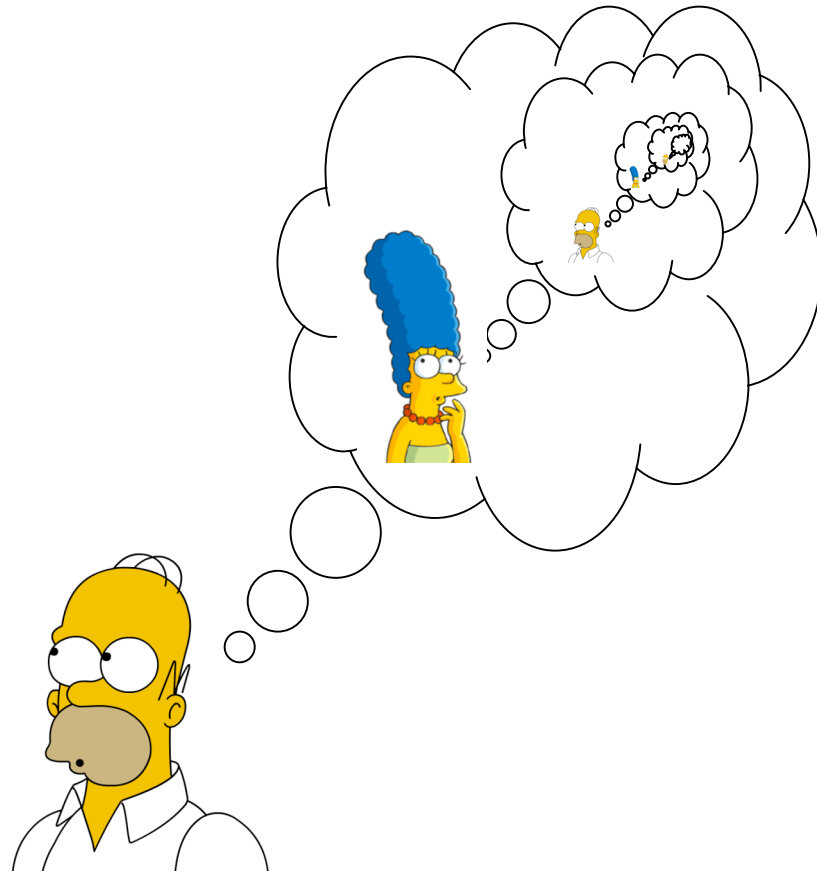


John von Neumann Oskar Morgenstern



- Random numbers $\Rightarrow a = 50 \Rightarrow 25$
- Everybody knows $\Rightarrow a = 25 \Rightarrow 12$
- Everybody knows $\Rightarrow a = 12 \Rightarrow 6$
- ... $\Rightarrow 1$

Rationality



Coordination games



1

-100



-100

1

Matrix games



a_{AA}

a_{AB}



a_{BA}

a_{BB}



Left player gets

Evolutionary game theory

- Populations
- Reproductive fitness = payoff
- Evolutionary dynamics



John Maynard Smith

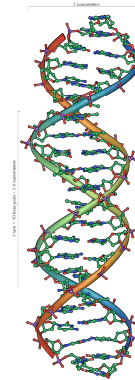
Left side drivers
get extinct



Interpretations of evolutionary game dynamics

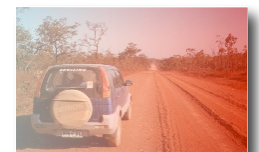
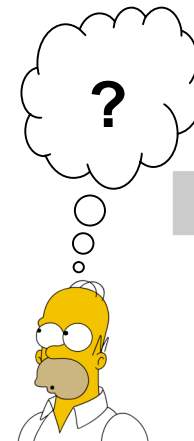
● Genetic

- Strategies are determined genetically
- More successful individuals produce more offspring
- Less successful strategies go extinct



● Cultural

- Strategies are choices of individuals
- More successful strategies are imitated more often



Replicator dynamics

$$\dot{x}_i = x_i (\pi_i - \langle \pi \rangle)$$

Abundance of
strategy i

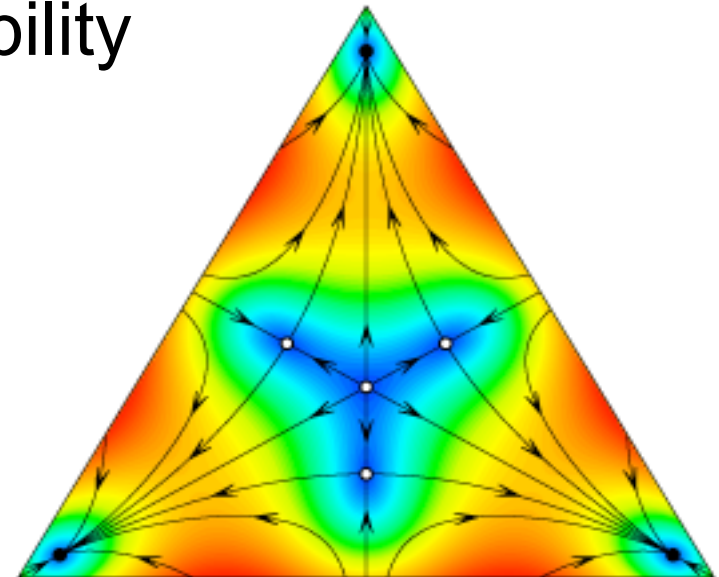
Payoff of
strategy i

Average payoff in
the population

Replicator dynamics

$$\dot{x}_i = x_i (\pi_i - \langle \pi \rangle)$$

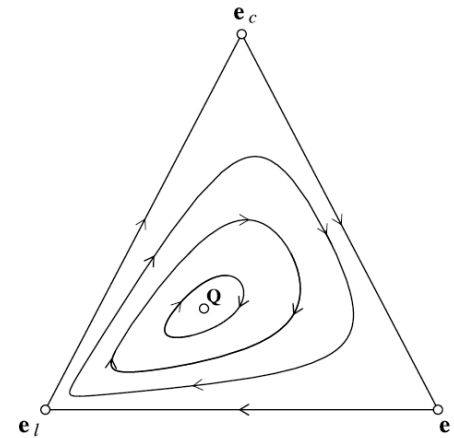
- Fixed points & their stability



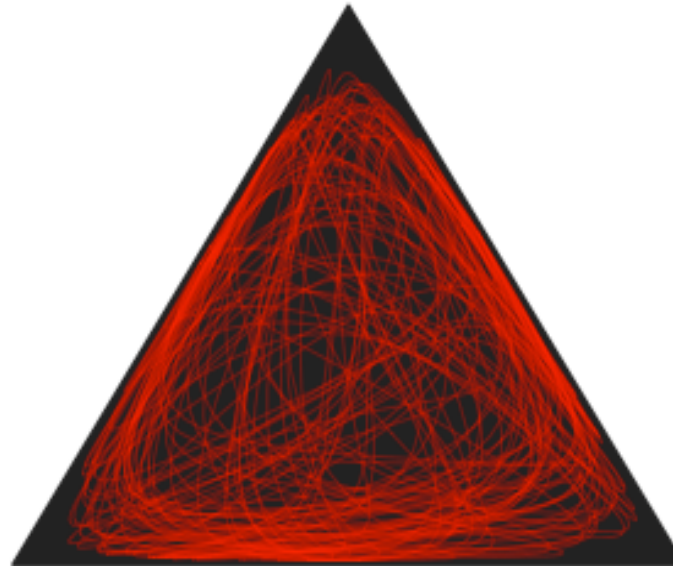
Replicator dynamics

$$\dot{x}_i = x_i (\pi_i - \langle \pi \rangle)$$

- Constants of motion



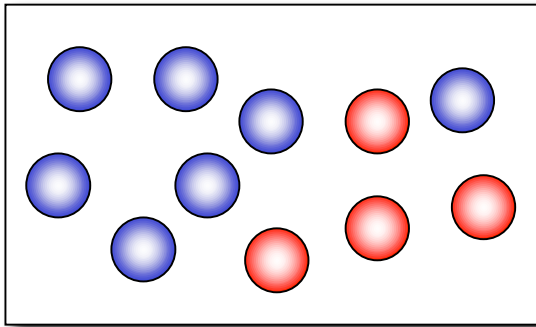
- Deterministic Chaos



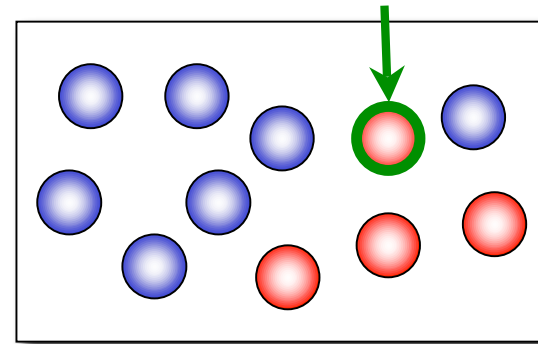
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Pairwise comparison

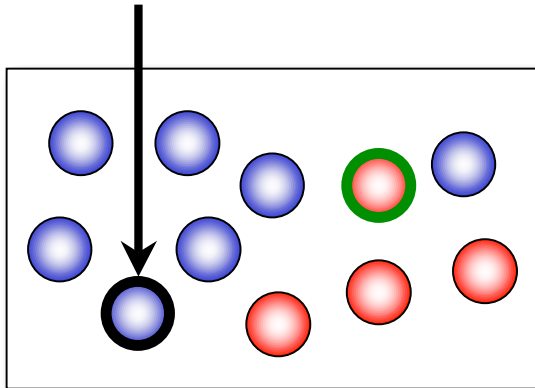
Initial population



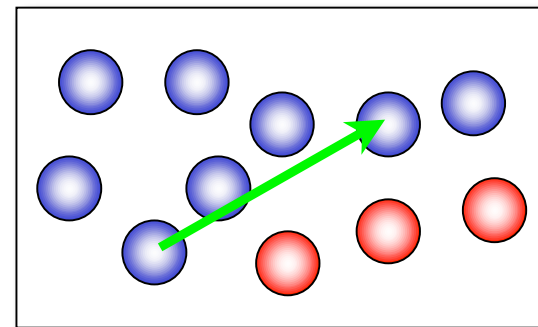
Select focal individual
(at random)



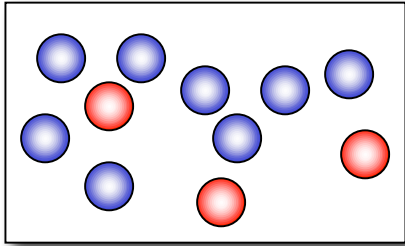
Select (random) role model



Adopt strategy according to
payoff difference



The “Fermi” process



$$T_{i,i+1} = \frac{N-i}{N} \frac{i}{N} \frac{1}{1 + e^{-\beta(\pi_A - \pi_B)}}$$

A **B** player updates its strategy

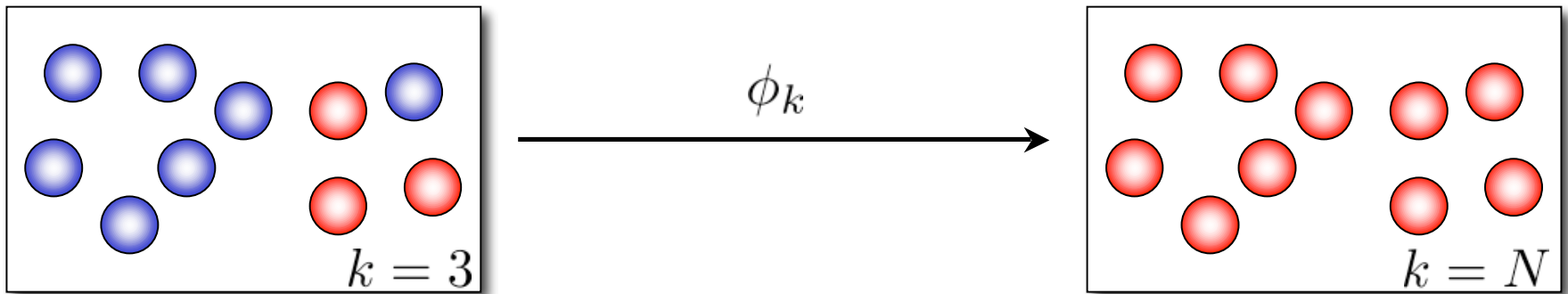
The role model is **A**

Probability to switch to **A**

Inverse temperature of selection

$$T_{i,i-1} = \frac{i}{N} \frac{N-i}{N} \frac{1}{1 + e^{+\beta(\pi_A - \pi_B)}}$$

Fixation probabilities



$$\phi_k = T_{k,k-1}\phi_{k-1} + T_{k,k+1}\phi_{k+1} + T_{k,k}\phi_k$$

$$\phi_k = \frac{\sum_{i=0}^{k-1} \prod_{j=1}^i T_{j,j-1}/T_{j,j+1}}{\sum_{i=0}^{N-1} \prod_{j=1}^i T_{j,j-1}/T_{j,j+1}}$$

Ratio of transition probabilities

$$\phi_k = \frac{\sum_{i=0}^{k-1} \prod_{j=1}^i T_{j,j-1}/T_{j,j+1}}{\sum_{i=0}^{N-1} \prod_{j=1}^i T_{j,j-1}/T_{j,j+1}}$$

$$\begin{aligned} \frac{T_{i,i-1}}{T_{i,i+1}} &= \frac{\frac{i}{N} \frac{N-i}{N} \frac{1}{1+e^{+\beta(\pi_A-\pi_B)}}}{\frac{N-i}{N} \frac{i}{N} \frac{1}{1+e^{-\beta(\pi_A-\pi_B)}}} \\ &= \frac{1 + e^{-\beta(\pi_A-\pi_B)}}{1 + e^{+\beta(\pi_A-\pi_B)}} \\ &= e^{-\beta(\pi_A-\pi_B)} \end{aligned}$$

Fixation probability for the Fermi process

$$\begin{aligned}\phi_k &= \frac{\sum_{i=0}^{k-1} \prod_{j=1}^i \frac{T_{j,j-1}}{T_{j,j+1}}}{\sum_{i=0}^{N-1} \prod_{j=1}^i \frac{T_{j,j-1}}{T_{j,j+1}}} && \text{General equation} \\ &= \frac{\sum_{i=0}^{k-1} \prod_{j=1}^i e^{-\beta(\pi_A - \pi_B)}}{\sum_{i=0}^{N-1} \prod_{j=1}^i e^{-\beta(\pi_A - \pi_B)}} && \text{Fermi process} \\ &= \frac{\sum_{i=0}^{k-1} e^{-\beta \sum_{j=1}^i (\pi_A - \pi_B)}}{\sum_{i=0}^{N-1} e^{-\beta \sum_{j=1}^i (\pi_A - \pi_B)}} \\ &\approx \frac{\int_{i=-0.5}^{k-0.5} e^{-\beta \sum_{j=1}^i (\pi_A - \pi_B)} di}{\int_{i=-0.5}^{N-0.5} e^{-\beta \sum_{j=1}^i (\pi_A - \pi_B)} di} && \text{Large populations}\end{aligned}$$

Calculation of payoffs

$$\begin{array}{c} A \\ B \end{array} \begin{array}{cc} A & B \\ \left(\begin{array}{cc} a & b \\ c & d \end{array} \right) \end{array}$$

$$\begin{aligned} \pi_A &= a \frac{i-1}{N-1} + b \frac{N-i}{N-1} \\ \pi_B &= c \frac{i}{N-1} + d \frac{N-i-1}{N-1} \end{aligned}$$

$$\pi_A - \pi_B = \underbrace{\frac{(a-b-c+d)}{N-1}}_u i + \underbrace{\frac{-a+bN-dN+d}{N-1}}_v$$

$$u = 0 \quad \text{“equal gains from switching”}$$

Fixation probabilities for arbitrary N and β

$$\pi_A - \pi_B = u \cdot i + v$$

 Frequency dependence

1. Frequency independent, $u=0$

$$\phi_k = \frac{1 - r^{-k}}{1 - r^{-N}}$$

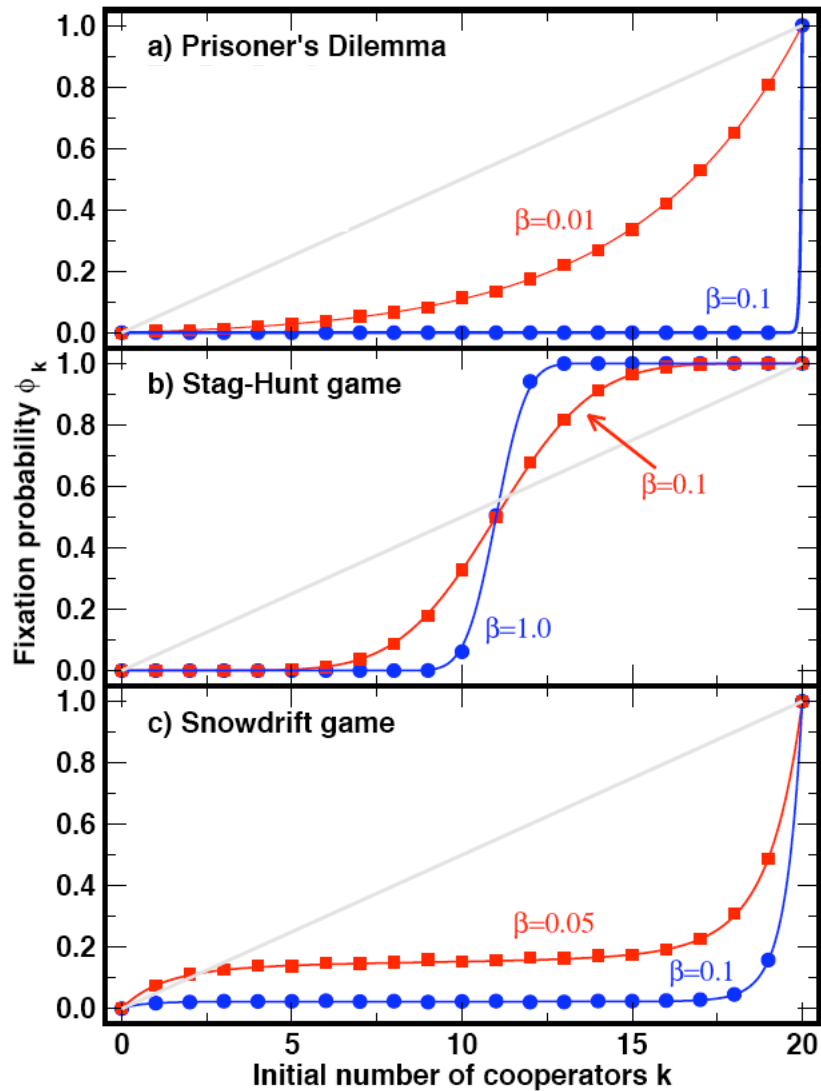
$$r = e^{2\beta v}$$

2. Frequency dependent

$$\phi_k = \frac{\text{erf} [\xi_k] - \text{erf} [\xi_0]}{\text{erf} [\xi_N] - \text{erf} [\xi_0]}$$

$$\xi_k = \sqrt{\frac{\beta}{u}}(ku + v)$$

Numerical results for fixation probabilities

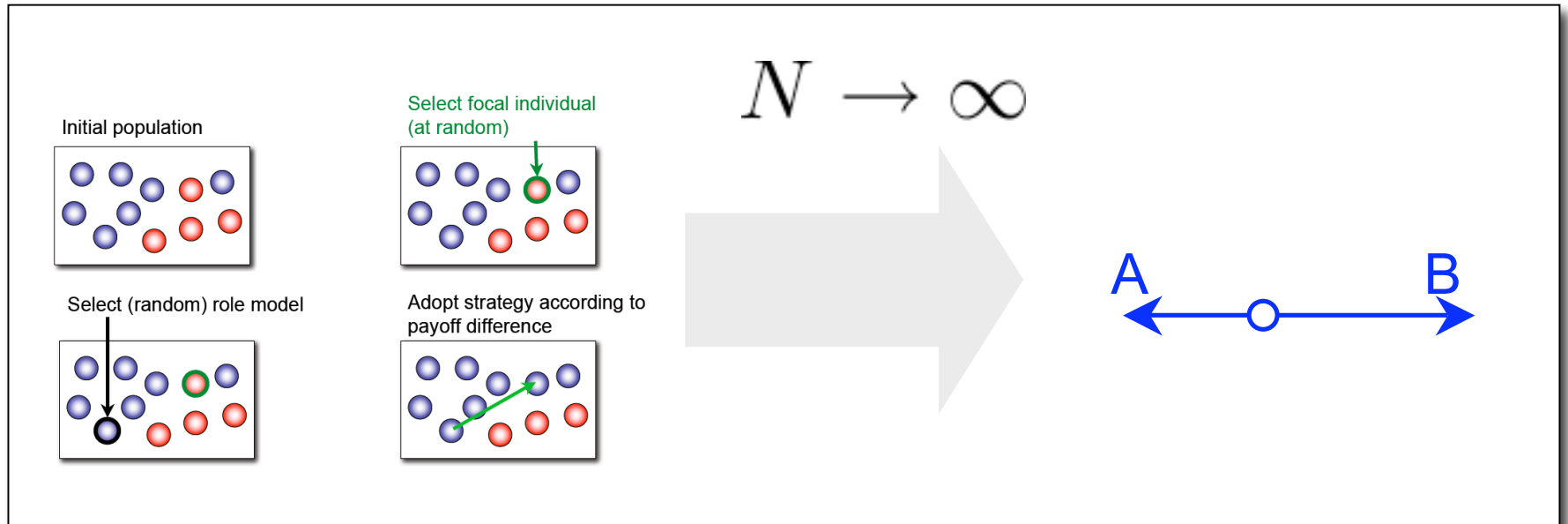


Very good agreement between theory and simulation

$$\sum_i \dots \rightarrow \int \dots di$$

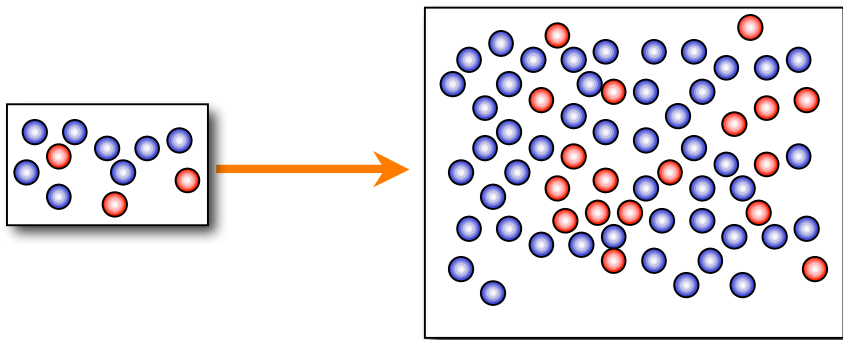
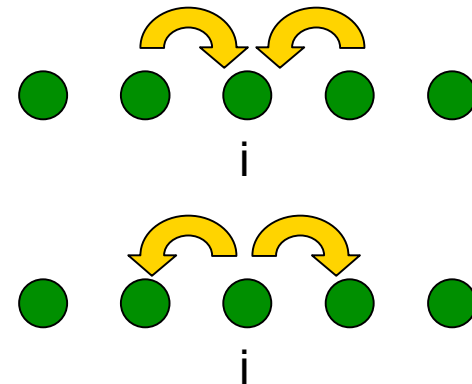
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From finite to infinite populations



Temporal dynamics

$$\begin{aligned}
 P_i^{\tau+1} &= P_i^{\tau} \\
 &+ P_{i-1}^{\tau} T_{i-1}^+ + P_{i+1}^{\tau} T_{i+1}^- \\
 &- P_i^{\tau} T_i^+ - P_i^{\tau} T_i^-
 \end{aligned}$$



Large populations:
 Approximate Master
 equation by Fokker-Planck
 equation

Fokker-Planck equation

$$\dot{\rho}(x, t) \approx -\frac{\partial}{\partial x} [a(x)\rho(x, t)] + \frac{\partial^2}{\partial x^2} [b^2(x)\rho(x, t)]$$

Drift

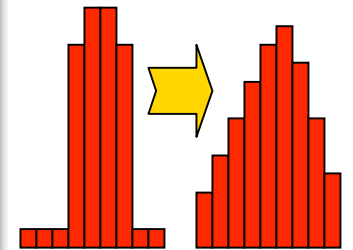
$$a(x) = T^+(x) - T^-(x)$$

Diffusion

$$b(x) = \sqrt{\frac{T^+(x) + T^-(x)}{N}}$$

From Fokker-Planck to Langevin

$$\dot{\rho}(x, t) \approx -\frac{\partial}{\partial x} [a(x)\rho(x, t)] + \frac{\partial^2}{\partial x^2} [b^2(x)\rho(x, t)]$$



Stochastic Calculus

$$\dot{x} = a(x) + b(x)\xi$$



Deterministic replicator equation for large N

$$\dot{x} = a(x) + \cancel{b(x)\xi}$$

$$b(x) = \sqrt{\frac{T^+(x) + T^-(x)}{N}}$$

$$N \rightarrow \infty$$

$$\begin{aligned}\dot{x} &= a(x) \\ &= T^+(x) - T^-(x) \\ &= (1-x)x \frac{1}{1 + e^{-\beta(\pi_A - \pi_B)}} - x(1-x) \frac{1}{1 + e^{+\beta(\pi_A - \pi_B)}} \\ &= x(1-x) \tanh \left[\frac{\beta}{2} (\pi_A - \pi_B) \right]\end{aligned}$$

Back to the usual replicator equation

$$\dot{x} = x(1-x) \tanh \left[\frac{\beta}{2} (\pi_A - \pi_B) \right] \quad \text{Weak selection}$$

$$\approx x(1-x) \frac{\beta}{2} (\pi_A - \pi_B)$$

$$= x(1-x) (\pi_A - \pi_B) \quad \text{New timescale}$$

$$= x [\pi_A - x\pi_A - (1-x)\pi_B]$$

$$= x [\pi_A - \langle \pi \rangle]$$

$$\dot{x}_i = x_i (\pi_i - \langle \pi \rangle)$$

Advantages of the Fermi process

- Arbitrary intensity of selection



- Closed expression for fixation probabilities

$$\phi_k = \frac{\operatorname{erf} [\xi_k] - \operatorname{erf} [\xi_0]}{\operatorname{erf} [\xi_N] - \operatorname{erf} [\xi_0]}$$

- Simple replicator equation for large N

$$\dot{x} = x(1 - x) \tanh \left[\frac{\beta}{2} (\pi_A - \pi_B) \right]$$

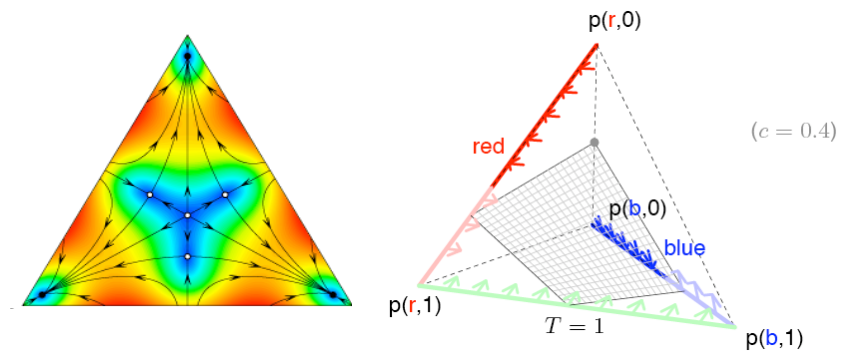
Generalization to more strategies

- Use d instead of 2 strategies

$$\dot{\rho}(\mathbf{x}) = - \sum_{k=1}^{d-1} \frac{\partial}{\partial x_k} \rho(\mathbf{x}) a_k(\mathbf{x}) + \frac{1}{2} \sum_{j,k=1}^{d-1} \frac{\partial^2}{\partial x_k \partial x_j} \rho(\mathbf{x}) b_{jk}(\mathbf{x})$$

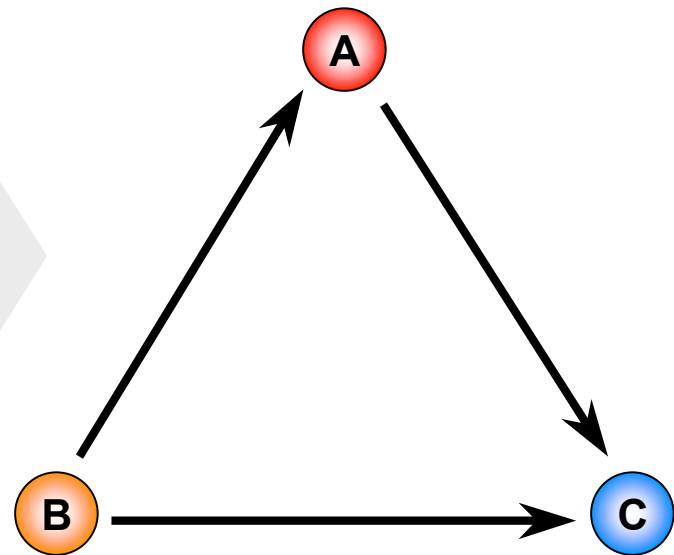
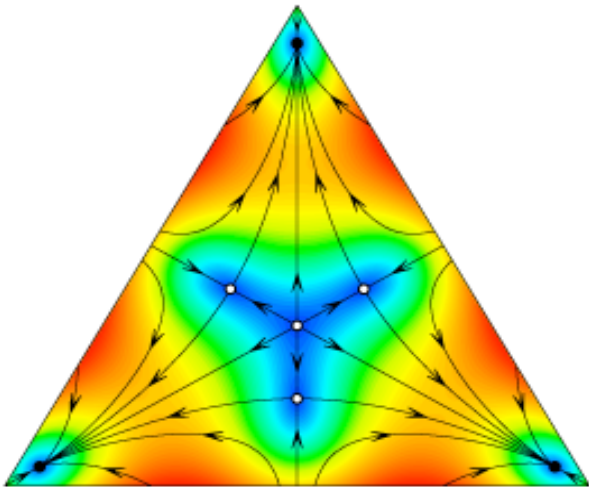
$$b_{jk}(\mathbf{x}) = \frac{1}{N} \left[-T_{jk}(\mathbf{x}) - T_{kj}(\mathbf{x}) + \delta_{jk} \sum_{l=1}^d T_{jl}(\mathbf{x}) + T_{lj}(\mathbf{x}) \right]$$

$$a_k(\mathbf{x}) = \sum_{j=1}^d T_{jk}(\mathbf{x}) - T_{kj}(\mathbf{x})$$



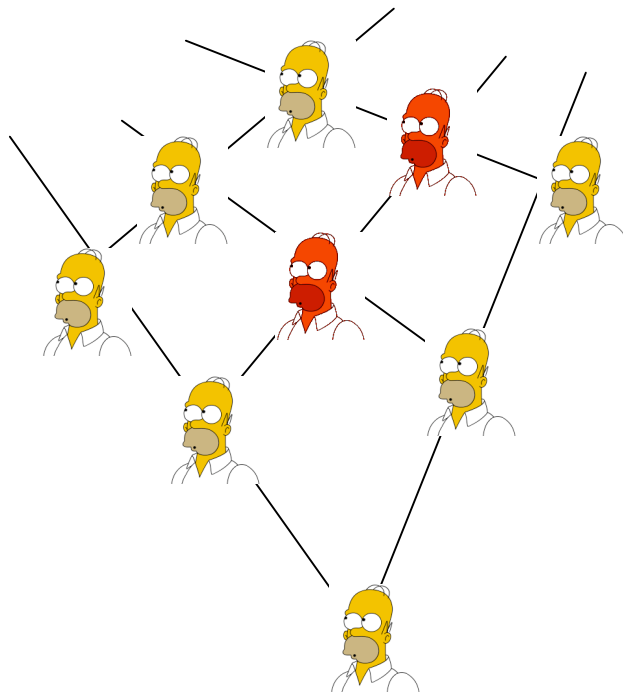
More strategies in finite populations

- Small mutation rates



Simulations for larger
mutation rates

Population structure



- Cellular automata
- Pair approximation
- Cluster methods
- Mapping to Ising models
- Kin selection
- ...

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The problem of cooperation

- Alarm calls



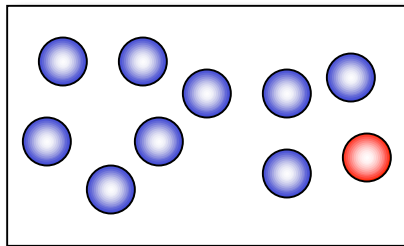
- Environment preservation



The Prisoner's Dilemma

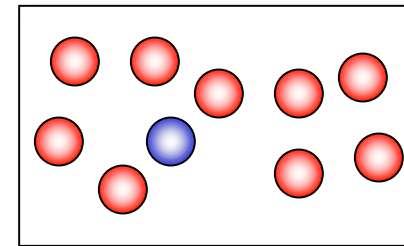
Player 1 cooperation	$-c$
Player 2 cooperation	$+b$
Player 1 payoff	$b-c$

		Player 2	
		C	D
Player 1	C	$b - c$	$-c$
	D	b	0



Cooperation

Natural Selection



Defection

Mechanisms for the evolution of cooperation

- Kin selection
- Direct reciprocity
- Indirect reciprocity
- Group selection
- Spatial reciprocity
- ...

Kin selection



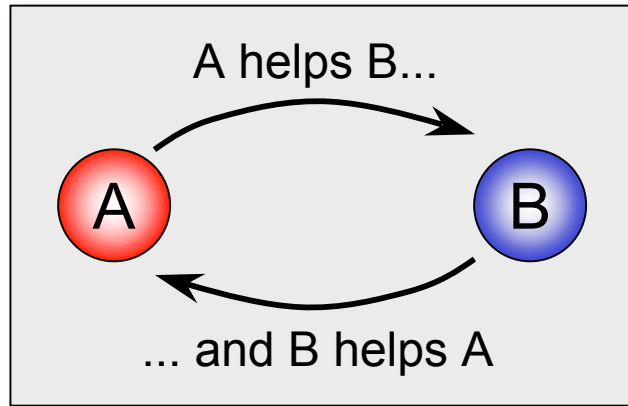
J.B.S Haldane

“I will jump into the river to save two brothers or eight cousins.”

$$\frac{b}{c} > \frac{1}{r}$$

↑
relatedness

Direct reciprocity



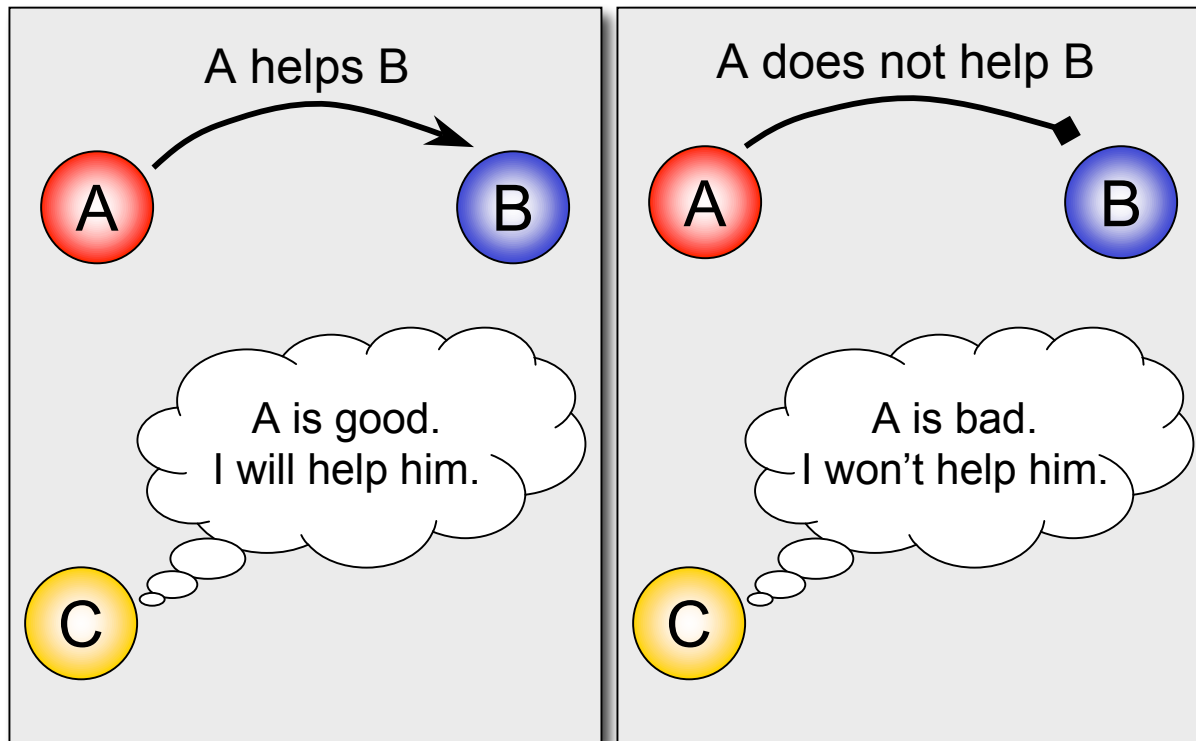
“I scratch your back
and you scratch mine.”

- Tit for Tat
- Pavlov
- ...

$$\frac{b}{c} > \frac{1}{w}$$

probability for further
interaction

Indirect reciprocity (reputation)

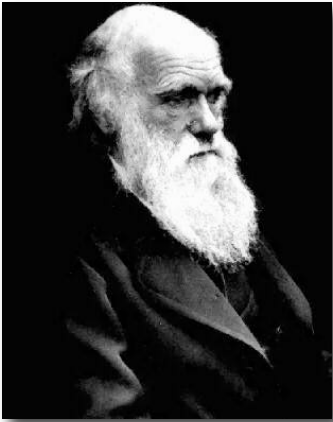


“I scratch your back and you scratch some-one else’s.”

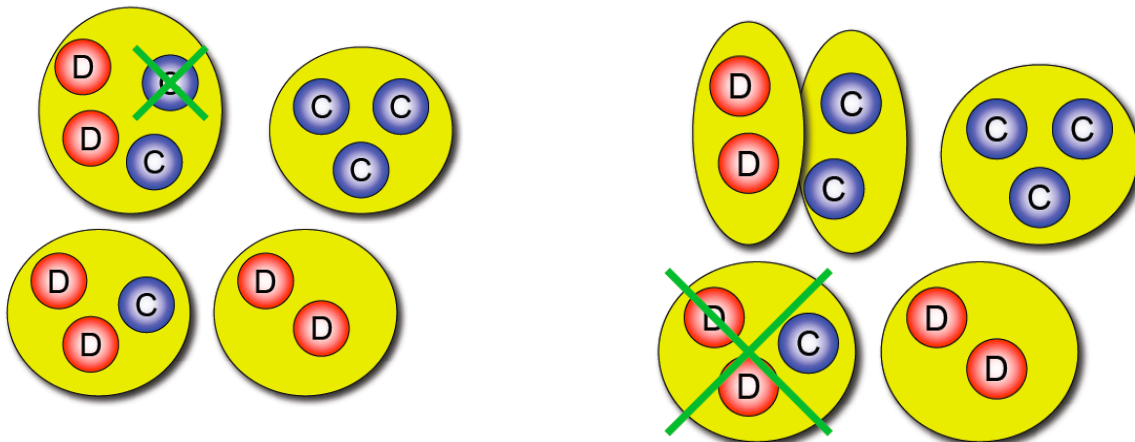
$$\frac{b}{c} > \frac{1}{q}$$

probability to know the reputation

Group selection



"[...] A tribe including many members who [...] were always ready to give aid to each other [...], would be victorious over other tribes; and this would be natural selection"

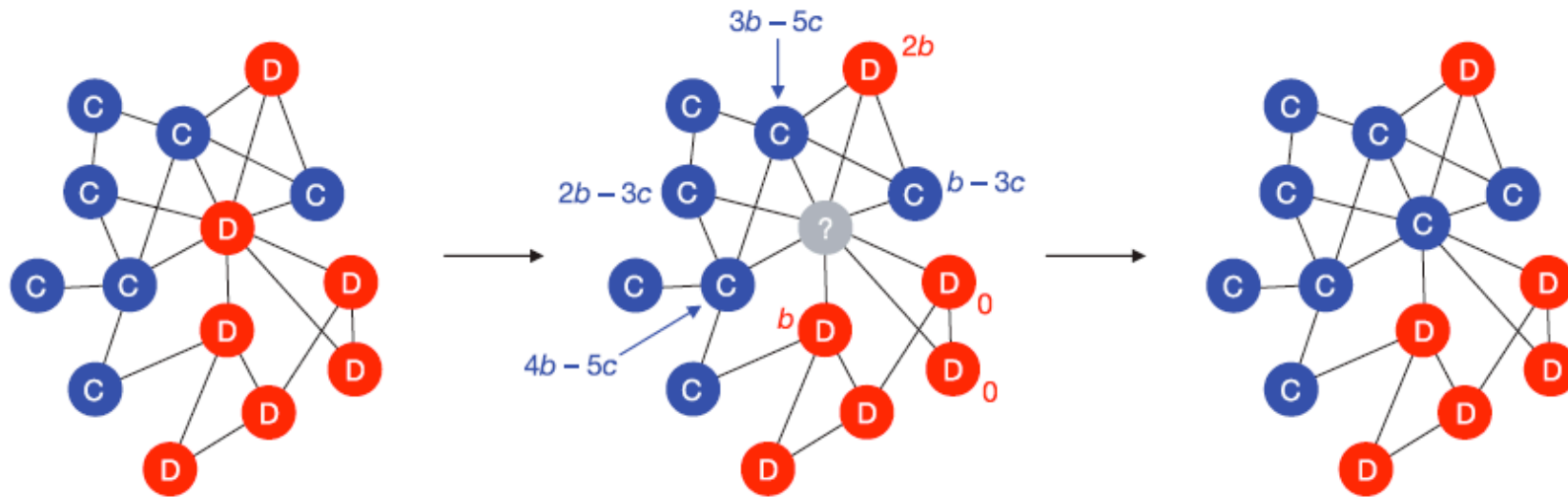


$$\frac{b}{c} > 1 + \frac{n}{m-2}$$

Size of groups
↓
 n

↑
Number of groups
 $m-2$

Network reciprocity

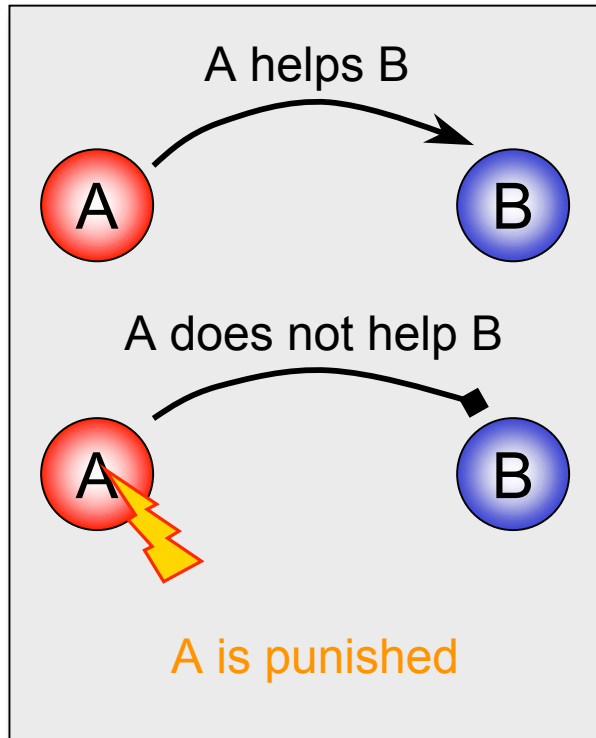


“Help your neighbors!”

$$\frac{b}{c} > k$$

Degree of the graph

Punishment ?



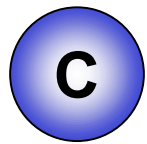
“If you don’t cooperate,
I will punish you!”

- Punishment can stabilize cooperation
- Punishment cannot lead to the evolution of cooperation
- How can punishment emerge?

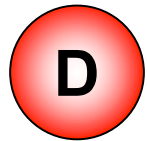
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The emergence of costly punishment

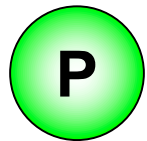
- Why are humans willing to punish others at a personal cost to themselves?



Play the game. Always cooperate with others.



Play the game. Exploit others that cooperate.

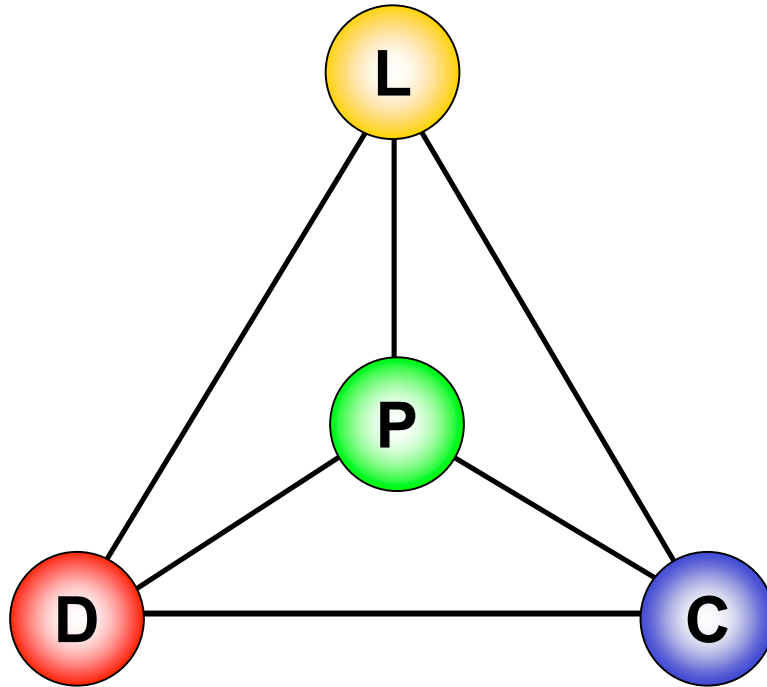


Play the game. Always cooperate and punish those who do not cooperate.



Do not join the game.

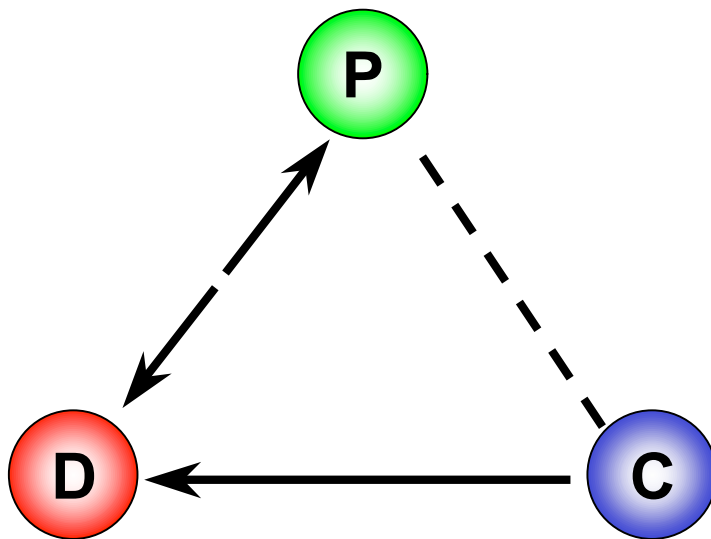
Interaction of four strategies



- 2 Types:
 $(C, D) = (i, N-i)$
- 3 Types:
 $(C, D, L) = (i, j, N-i-j)$
- 4 Types:
 $(C, D, L, P) = (i, j, k, N-i-j-k)$

Punishers in the Prisoner's Dilemma

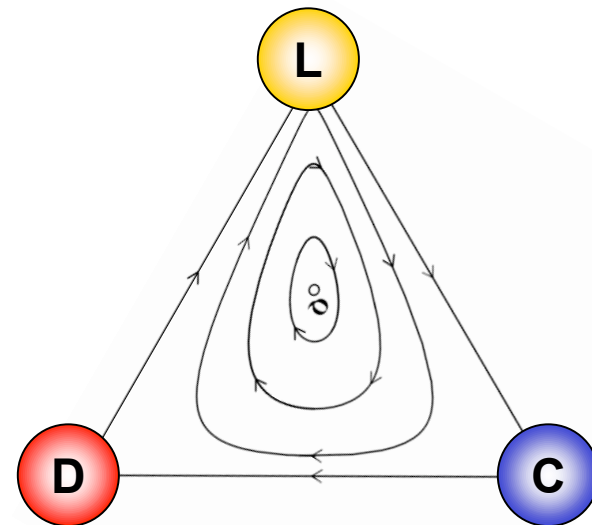
- Adding a punishment option does not lead to the emergence of punishment



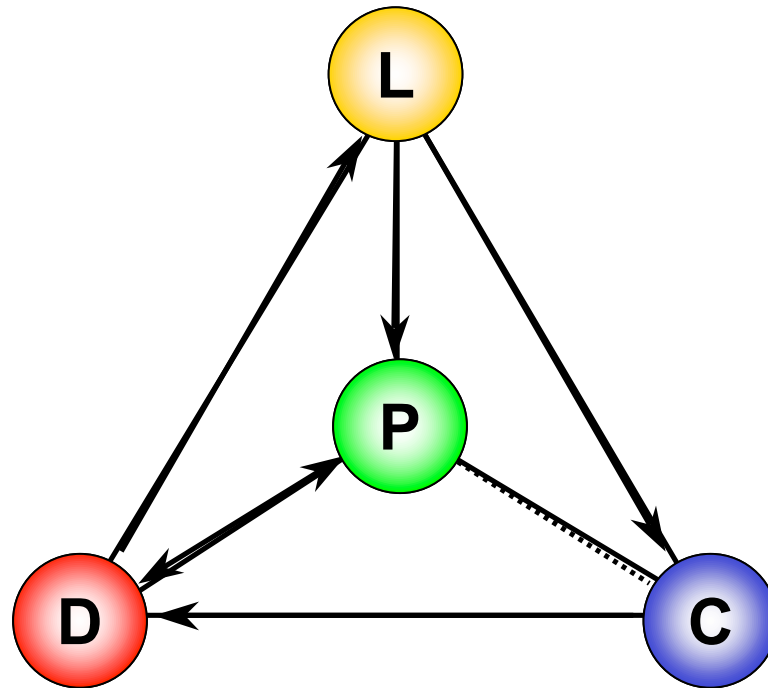
Only D is stable!

Voluntary participation

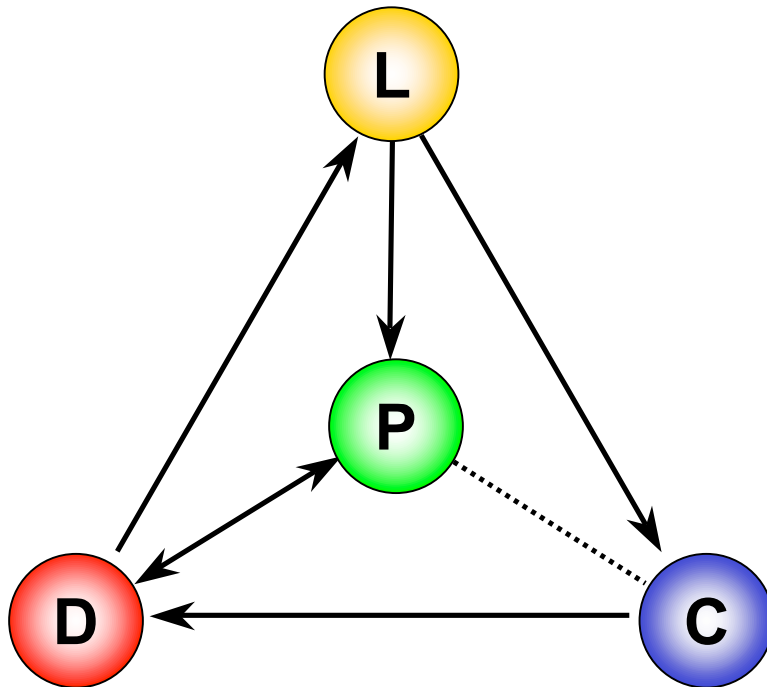
- “Loners” avoid the cost of joining the game σ
 - Loners are advantageous among defectors
 - Cooperation is advantageous if there are no defectors
 - Defection is advantageous among cooperators



Adding a fourth strategy



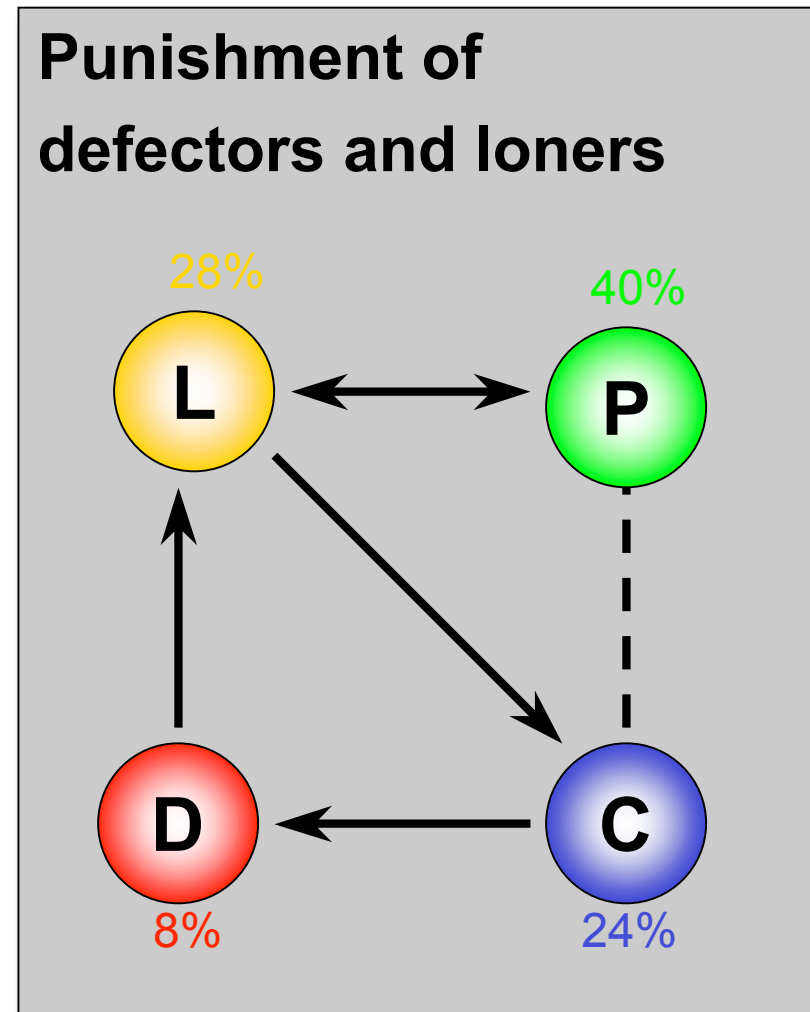
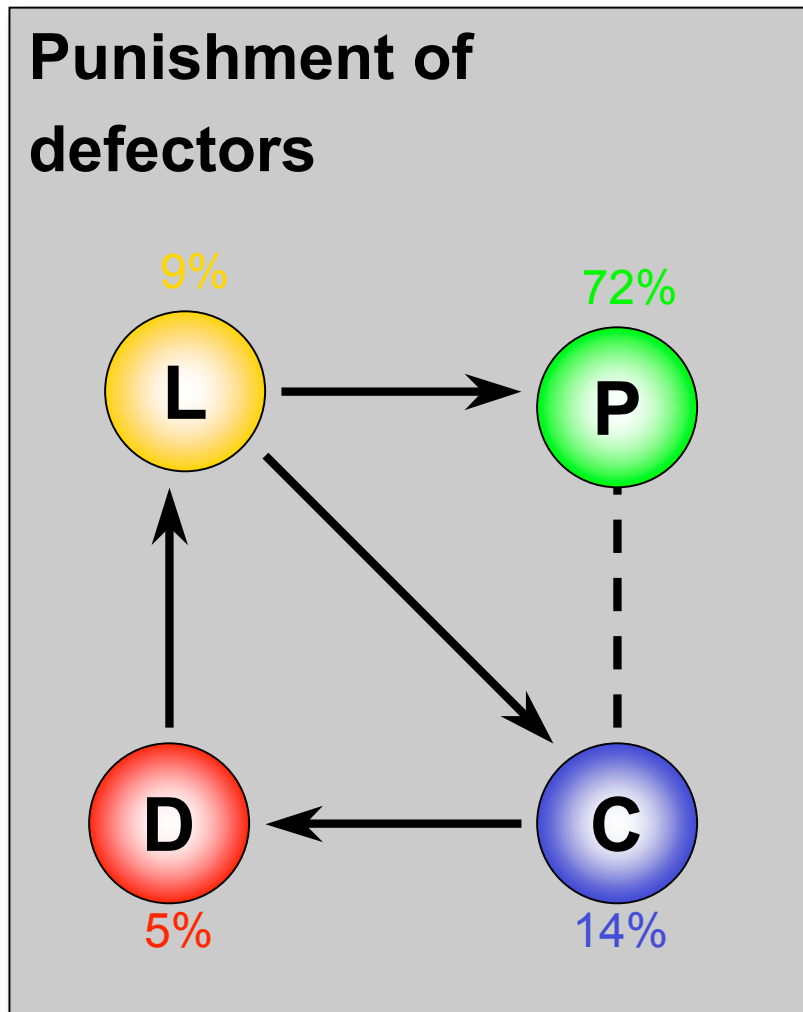
Stochastic dynamics in finite populations



Where will the system spend most of the time?

- Small mutation rates
- Embedded Markov chain on pure states
- Solve for stationary distribution

Punishment can evolve in finite populations

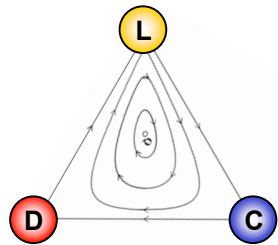


Game theoretic experiments

SCIENCE VOL 296 10 MAY 2002

Volunteering as Red Queen Mechanism for Cooperation in Public Goods Games

Christoph Hauert,^{1,2} Silvia De Monte,^{1,3} Josef Hofbauer,¹
Karl Sigmund^{1,4*}

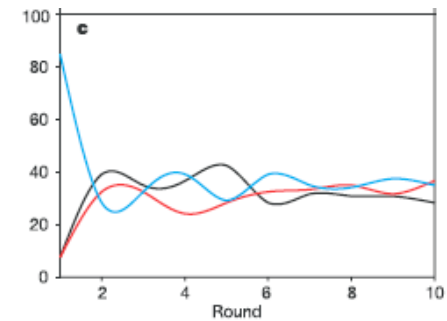


NATURE | VOL 425 | 25 SEPTEMBER 2003 |

Volunteering leads to rock-paper-scissors dynamics in a public goods game

Dirk Semmann, Hans-Jürgen Krambeck & Manfred Milinski

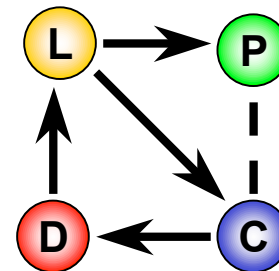
Department of Evolutionary Ecology, Max Planck Institute of Limnology,
24306 Plön, Germany



SCIENCE VOL 316 29 JUNE 2007

Via Freedom to Coercion: The Emergence of Costly Punishment

Christoph Hauert,¹ Arne Traulsen,¹ Hannelore Brandt,² Martin A. Nowak,¹ Karl Sigmund^{3,4*}

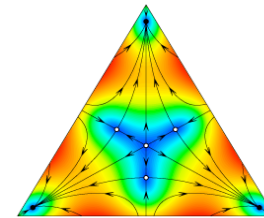


*work in
progress*

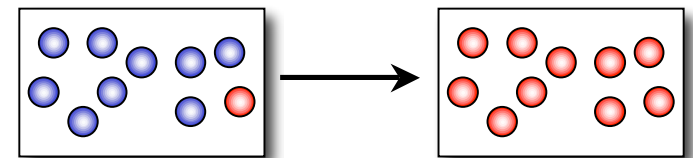
Summary

- Evolutionary game dynamics in finite populations

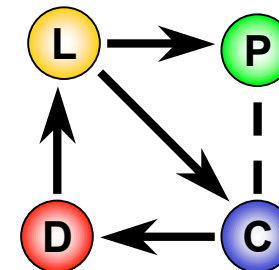
- Converges to replicator equation for large populations



- Equivalent to classical population dynamics for constant selection



- Applications: The evolution of cooperation, punishment,...



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