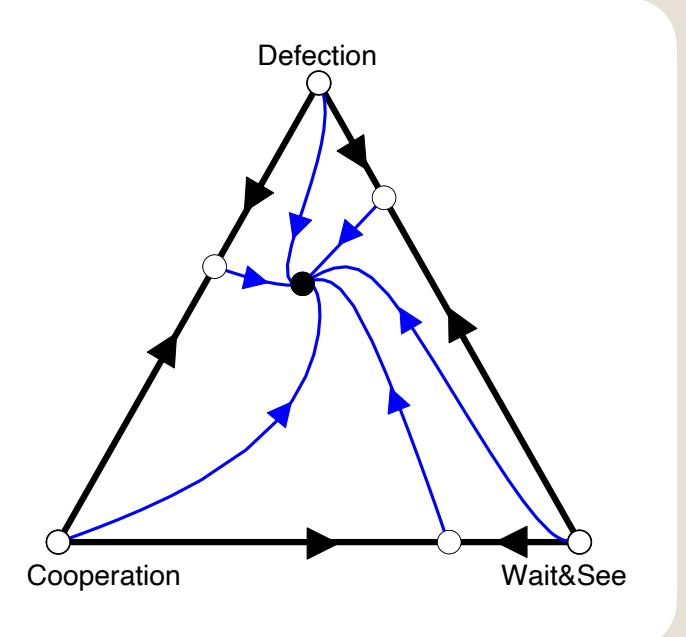


# An overview

## Yesterday's class (March 11, 2025)

- An introduction to evolutionary game theory  
(Replicator dynamics, games in finite populations)



## Today's classes (March 12, 2025)

- Evolution of cooperation & direct reciprocity
- Social norms & indirect reciprocity

## Tomorrow's class (March 13, 2025)

- Some current research: Reciprocity in complex environments

## Evolution of cooperation: Motivation

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## Reciprocal food sharing in the vampire bat

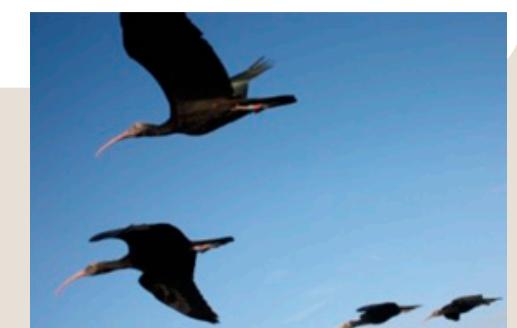
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Matching times of leading and following suggest cooperation through direct reciprocity during V-formation flight in ibis

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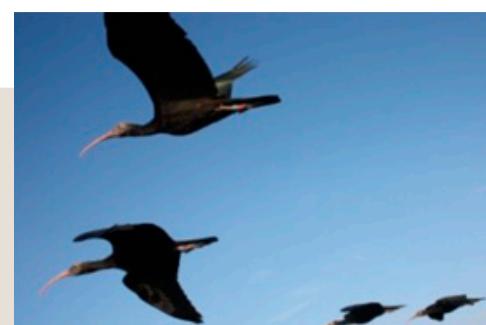
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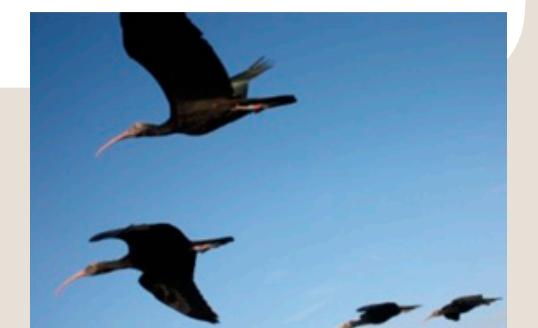
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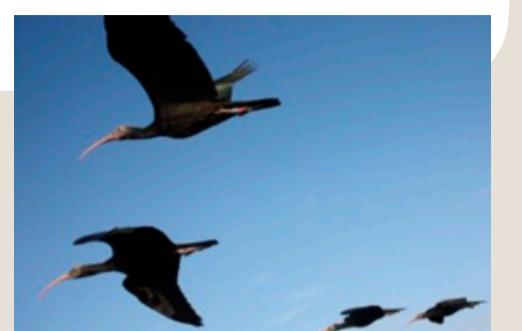
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Still, one might raise the question: Why would people have these particular emotions, social norms, and institutions?

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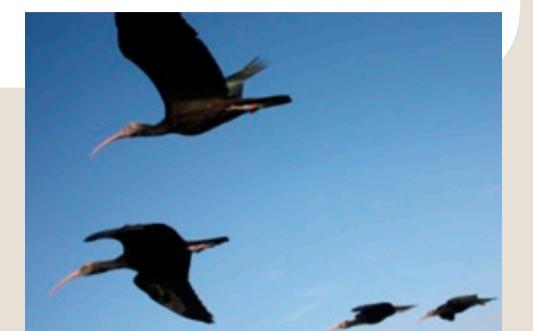
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## Evolution of cooperation: Prisoner's dilemma

### Remark 2.2. Cooperation in a Prisoner's dilemma

The problem of cooperation is usually illustrated with a prisoner's dilemma. Cooperation means to pay a cost  $c$ , for the co-player to get a payoff of  $b$ .

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Cooperate	$b-c$	$-c$
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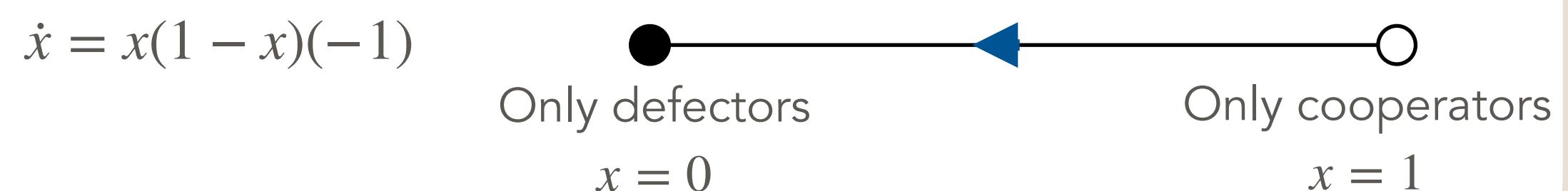
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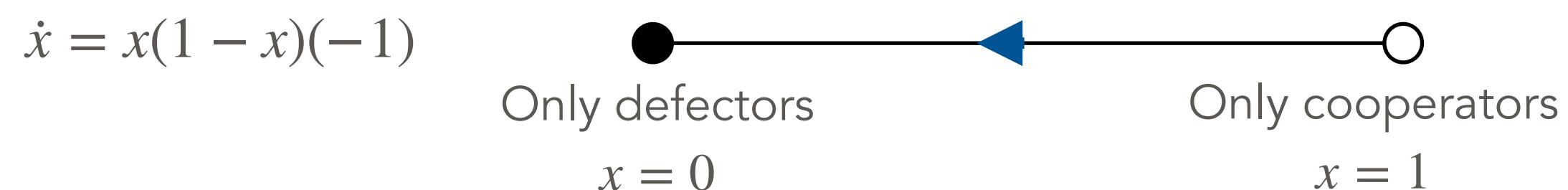
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The diagram illustrates a flow from a state labeled "Only defectors" to a state labeled "Only cooperators". It consists of two nodes connected by a horizontal arrow pointing from left to right. The left node is a solid black circle labeled "Only defectors" below it. The right node is an empty white circle labeled "Only cooperators" below it. A blue arrow points from the black circle to the white circle.

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TARGET REVIEW

## **The evolution of cooperation and altruism – a general framework and a classification of models**

L. LEHMANN\*† & L. KELLER\*

# Five Rules for the Evolution of Cooperation

Martin A. Nowak

# **Eleven mechanisms for the evolution of cooperation**

MICHAEL A. ZAGGI

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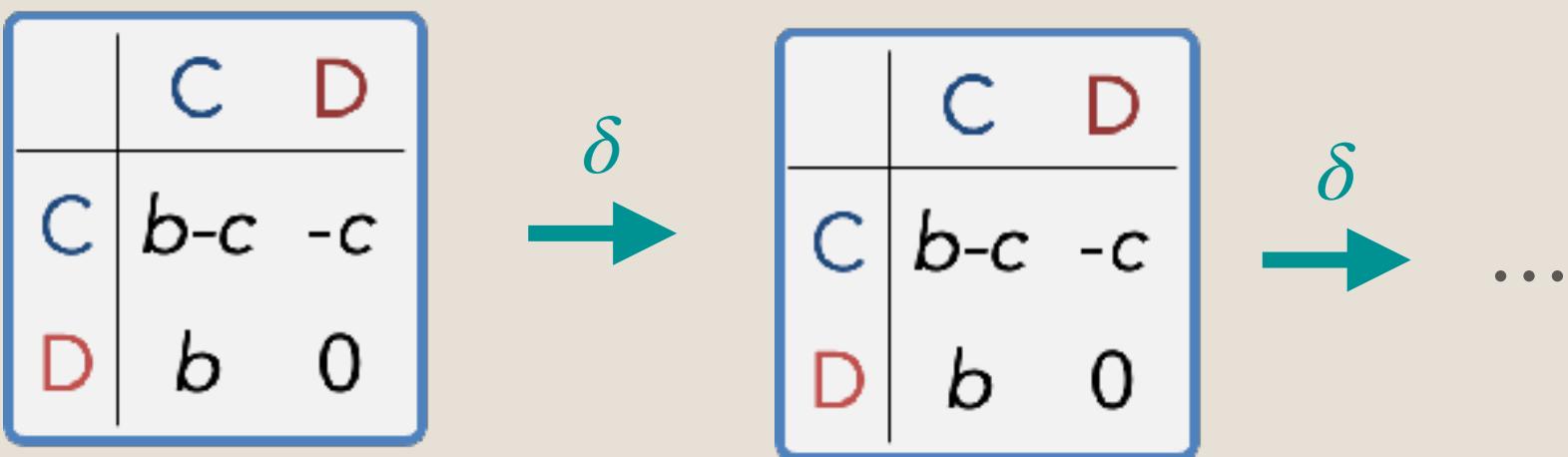
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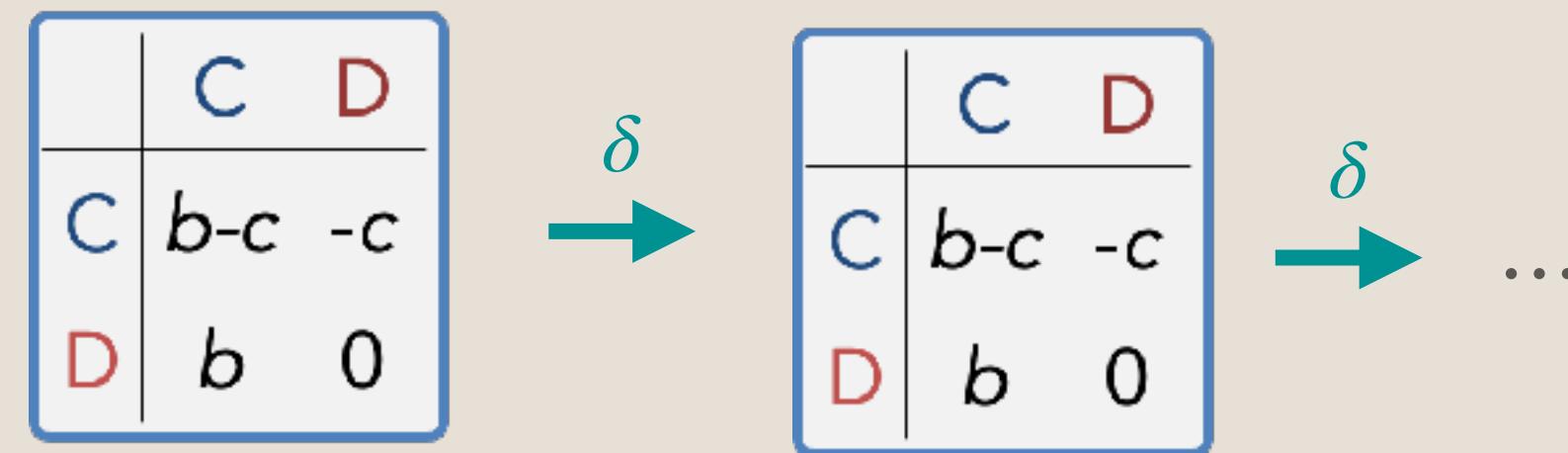
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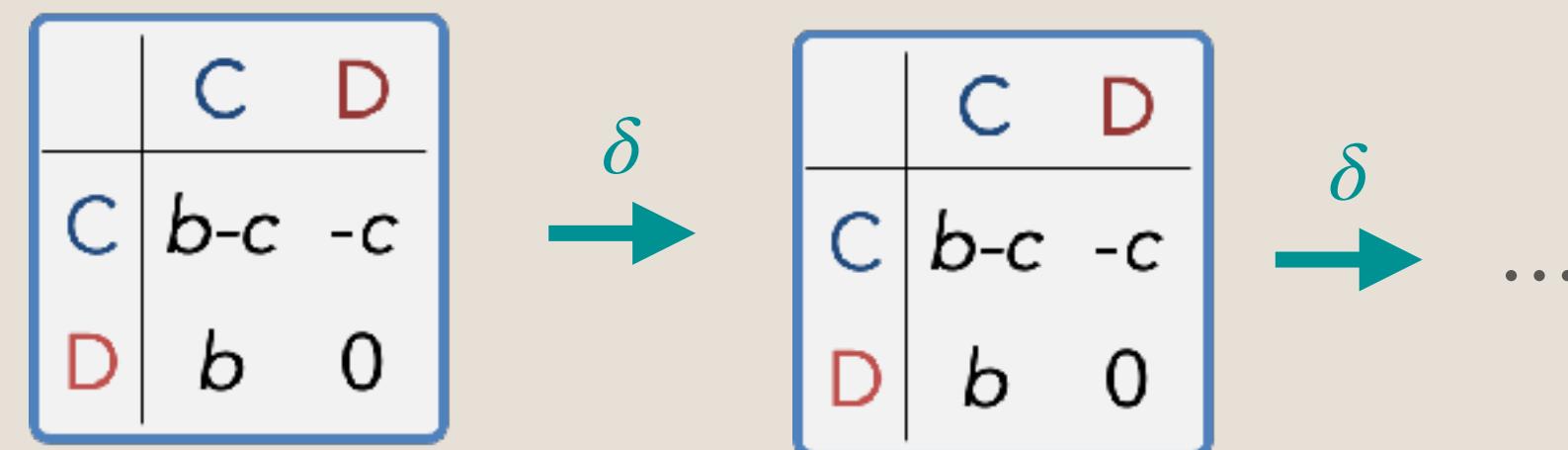
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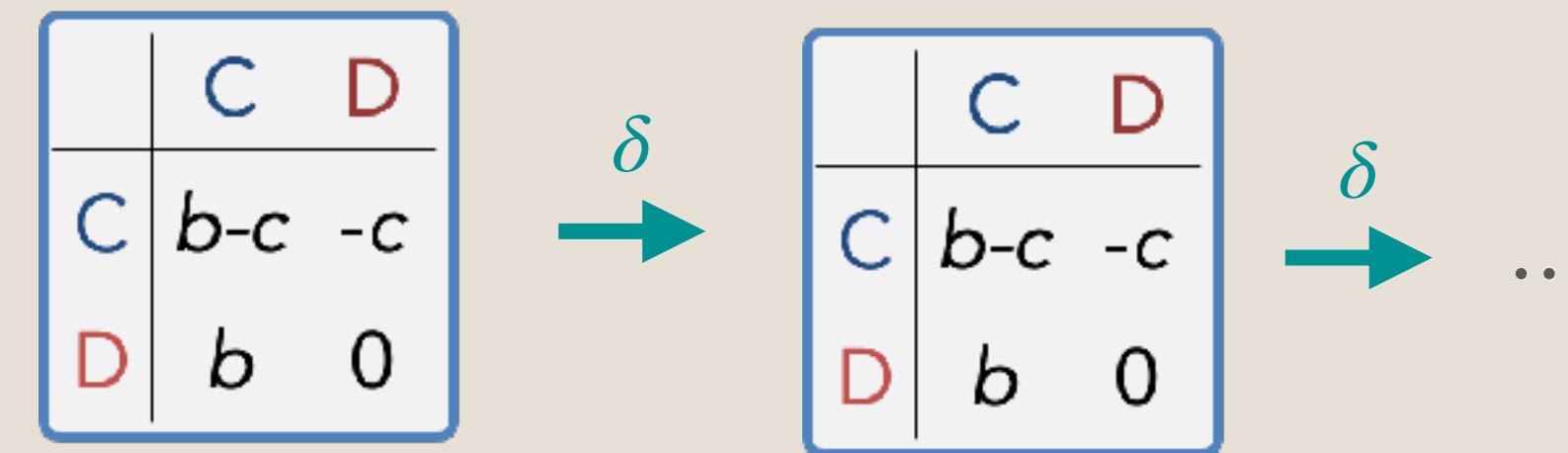
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- Let  $\mathbf{a}(t) = (a_1(t), a_2(t))$  denote the outcome of round  $t$ , with  $a_1(t), a_2(t) \in \{C, D\}$ .

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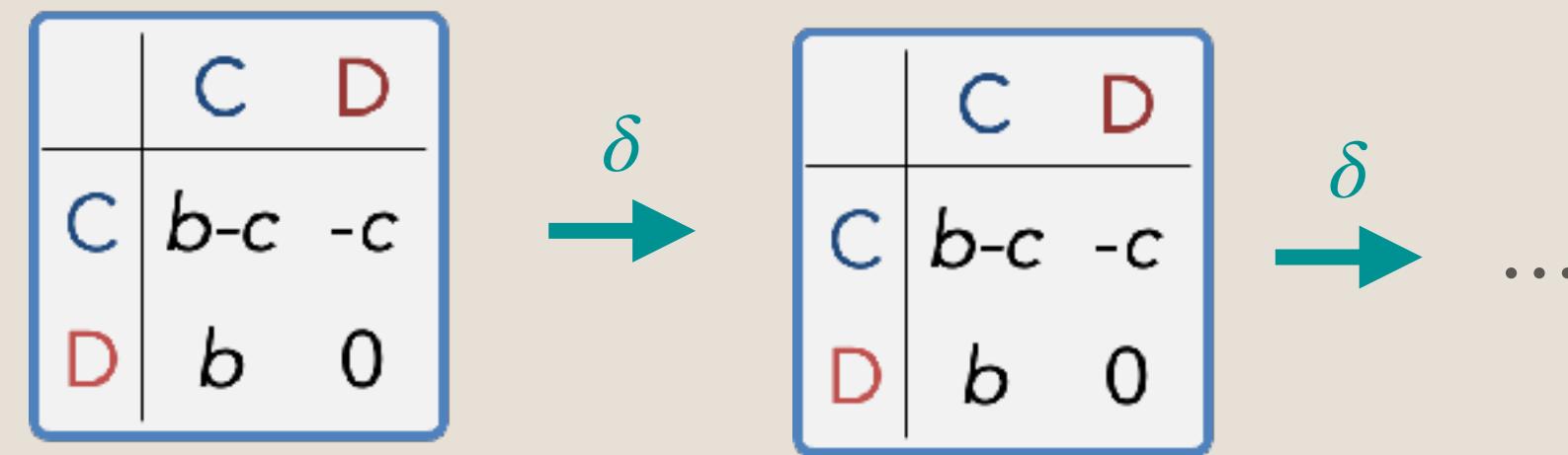
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$$\pi_i = (1 - \delta) \sum_{t=1}^{\infty} \delta^t \cdot \pi_i(t)$$

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### Remark 2.5. Strategies for the repeated prisoner's dilemma

In bi-matrix games, strategies were just probability distributions over the actions of the game.

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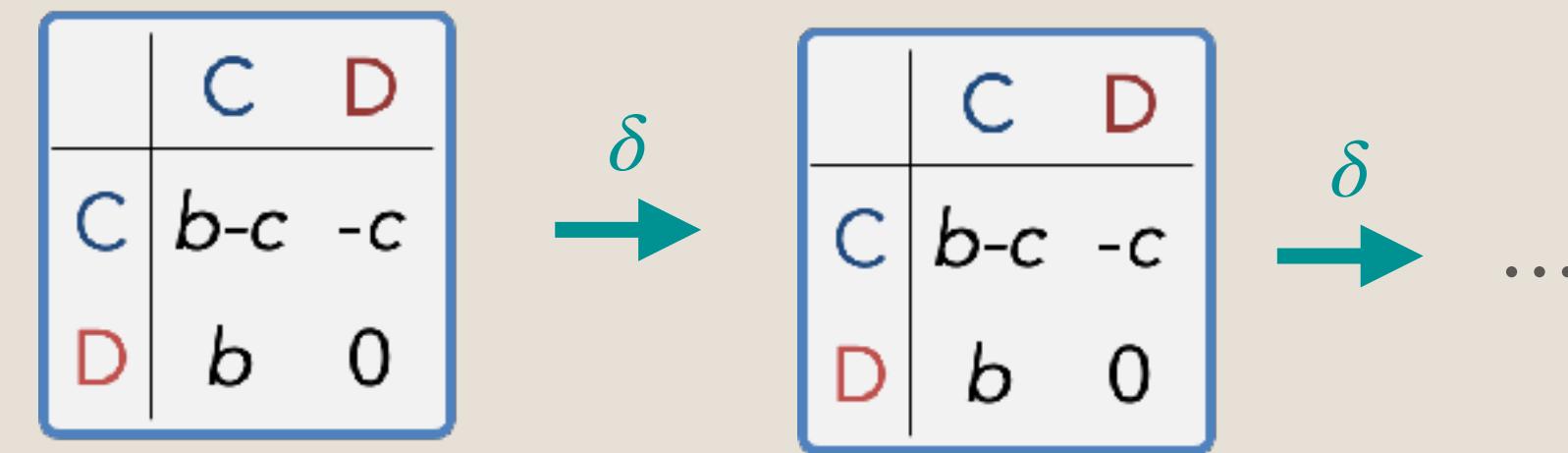
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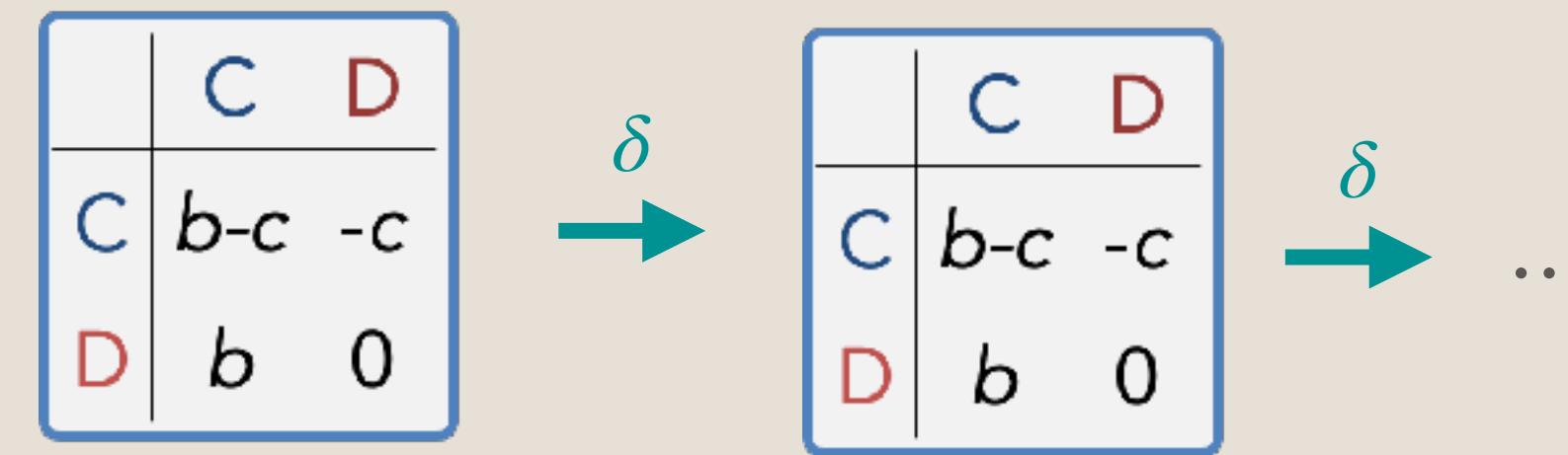
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- A strategy  $\sigma$  is now a function that takes as input the possible histories, and as output an action:

$$\sigma : H \rightarrow \{C, D\}$$

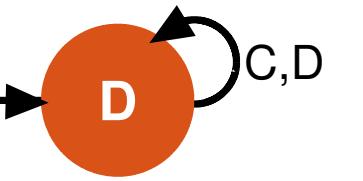
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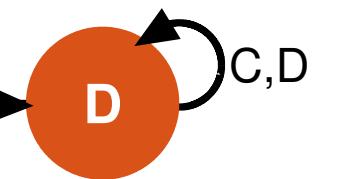
- **ALLD:** Defect in every single round, after all histories.



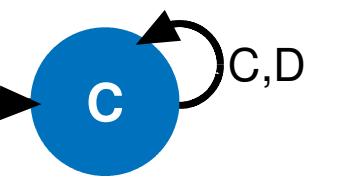
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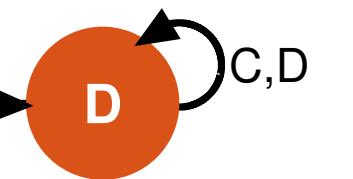
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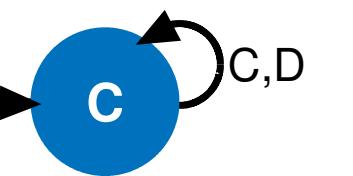
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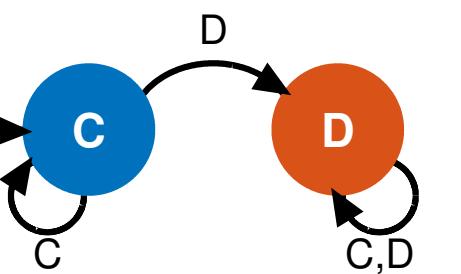
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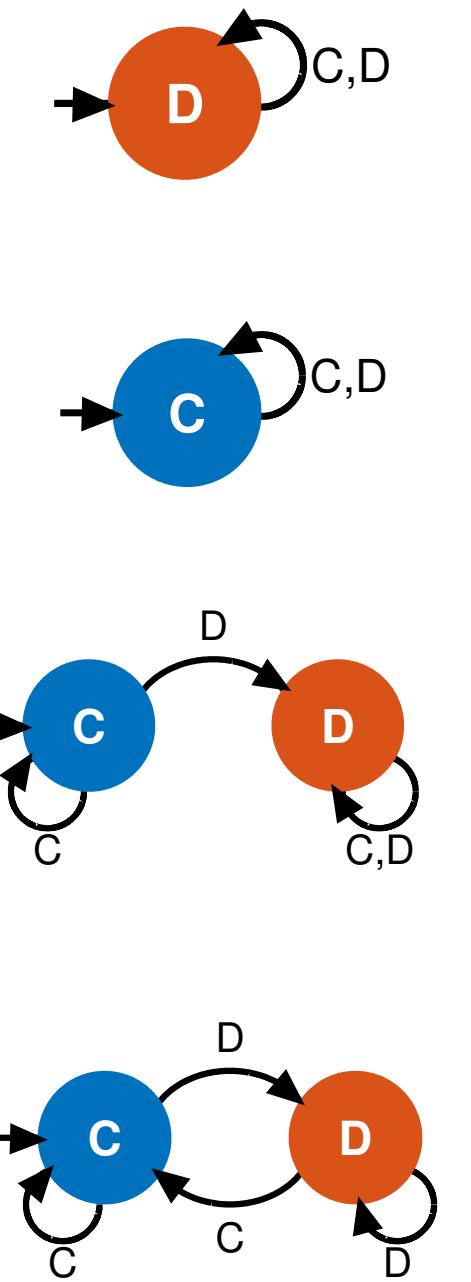
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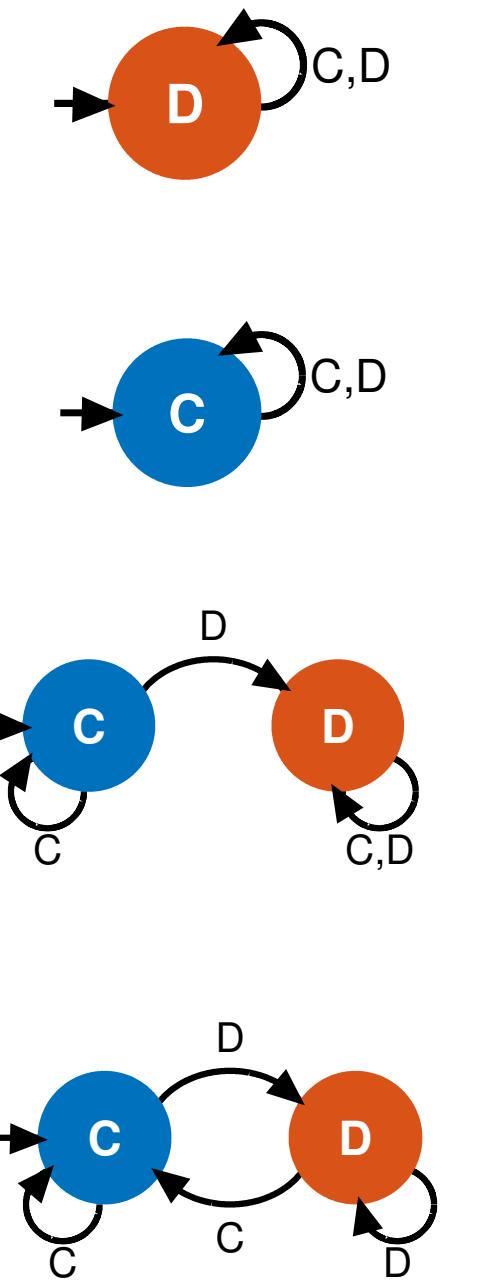
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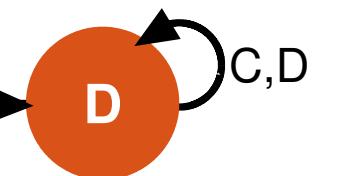


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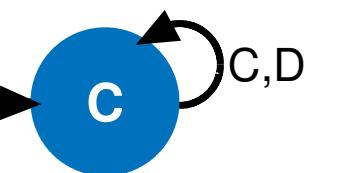
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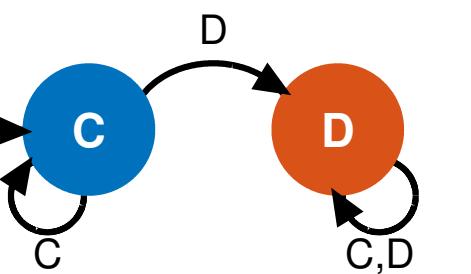
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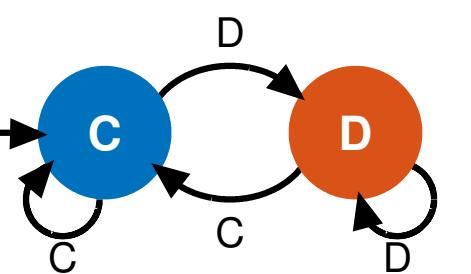
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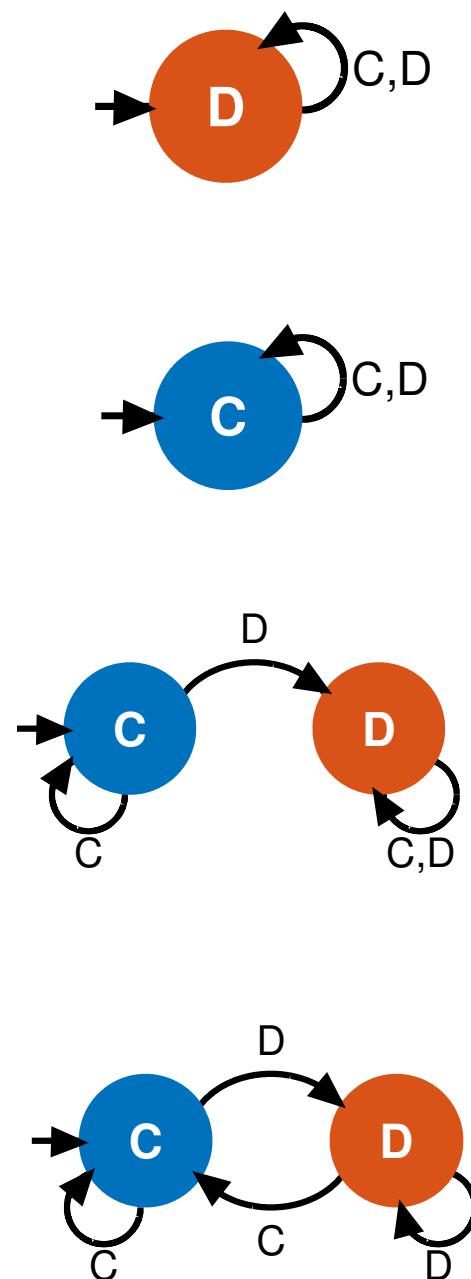
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## Definition 2.7. Nash equilibrium

A strategy profile  $(\sigma_1^*, \sigma_2^*)$  is a Nash equilibrium for the repeated game if

$$\pi_1(\sigma, \sigma_2^*) \leq \pi_1(\sigma_1^*, \sigma_2^*) \quad \text{and} \quad \pi_2(\sigma_1^*, \sigma) \leq \pi_2(\sigma_1^*, \sigma_2^*) \quad \text{for all } \sigma$$

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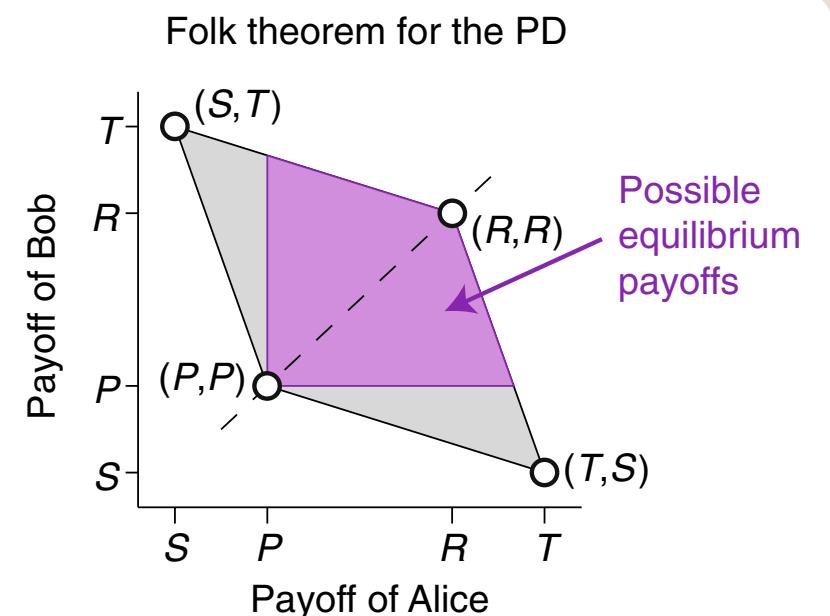
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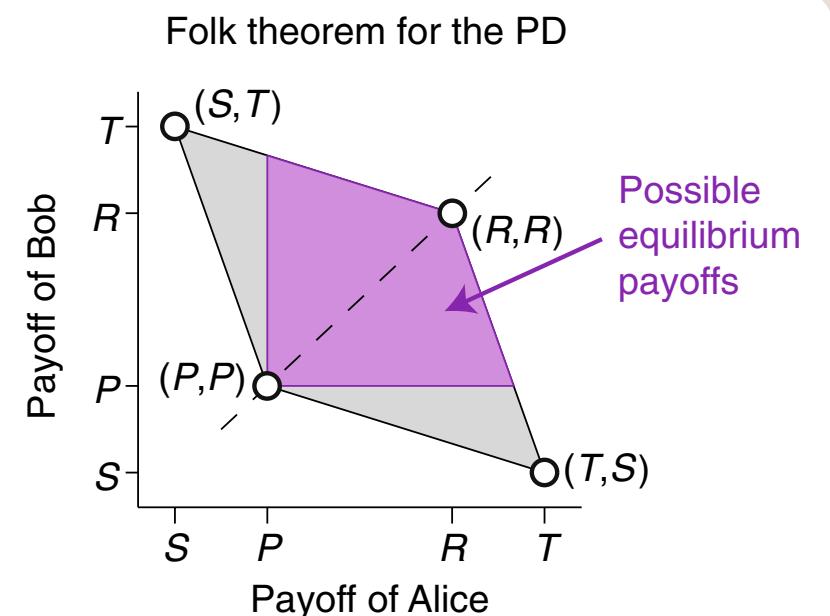
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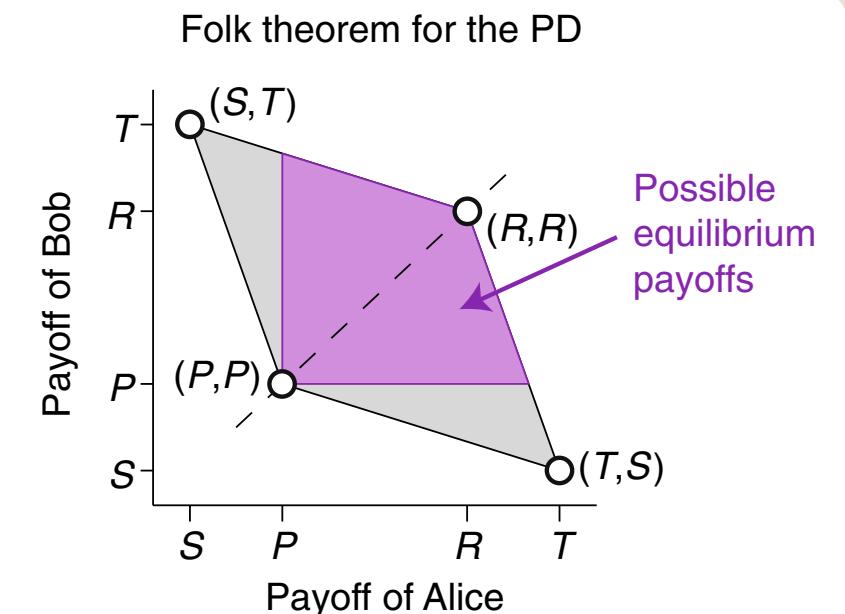
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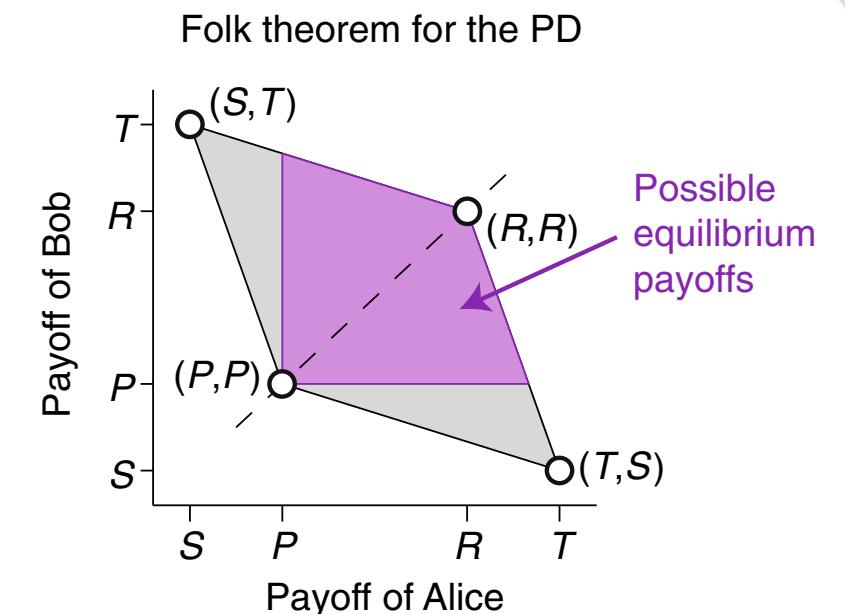
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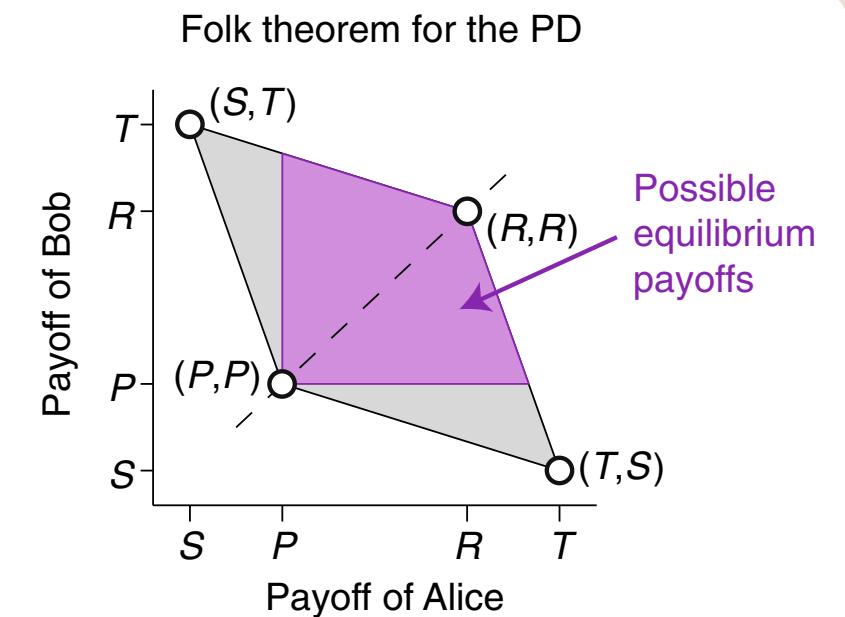
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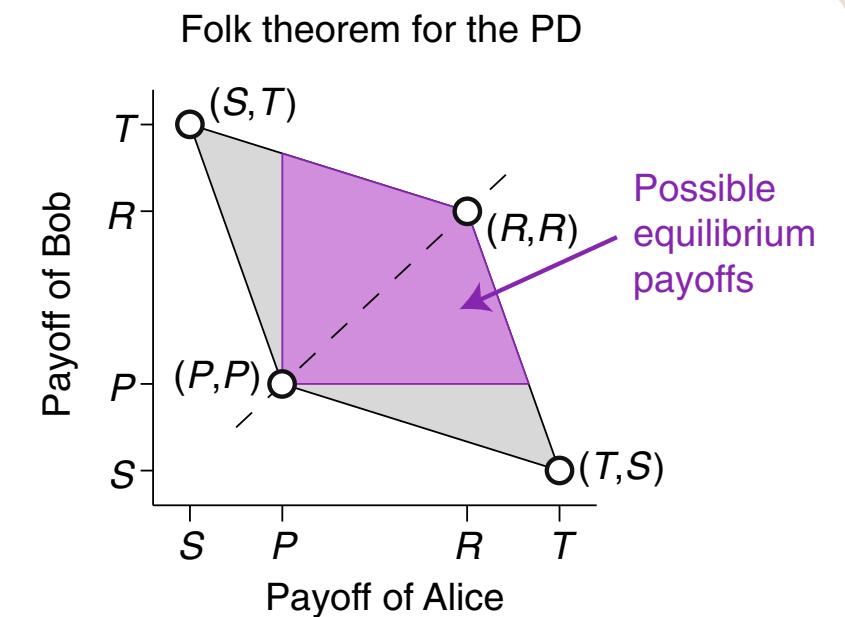
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  2. It remains unclear what strategy one should use in the repeated prisoner's dilemma in general.

## Direct reciprocity: Axelrod's tournament

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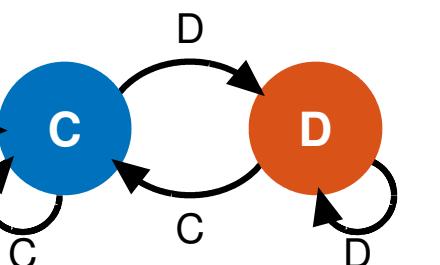
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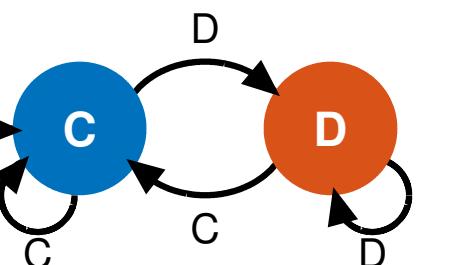
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TFT

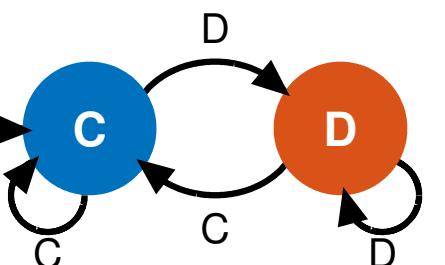
ALLD

Payoffs

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## The Evolution of Cooperation

Robert Axelrod and William D. Hamilton

Round 1

TFT C

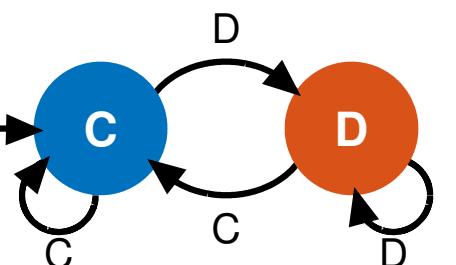
ALLD D

Payoffs (-c,b)

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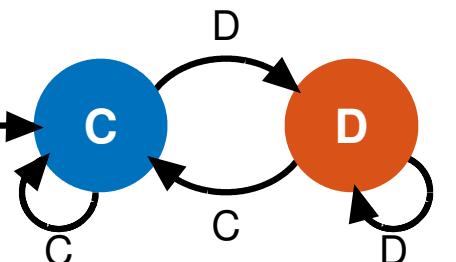
Robert Axelrod and William D. Hamilton

	Round 1	Round 2
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## The Evolution of Cooperation

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	Round 1	Round 2	Round 3
TFT	C	D	D
ALLD	D	D	D
Payoffs	(-c,b)	(0,0)	(0,0)

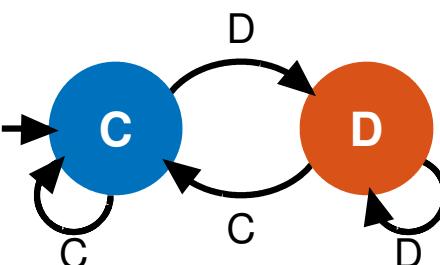
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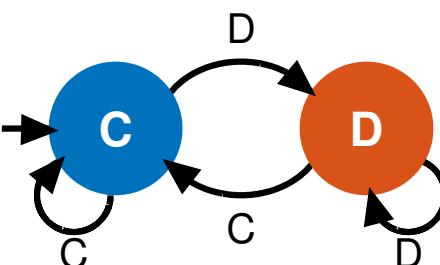
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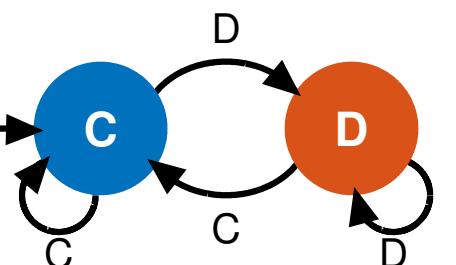


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Average payoffs:

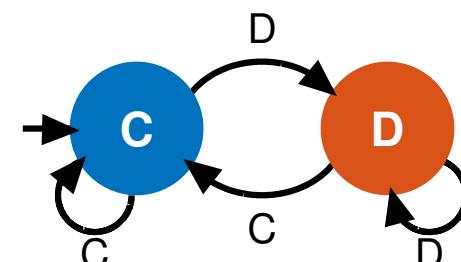
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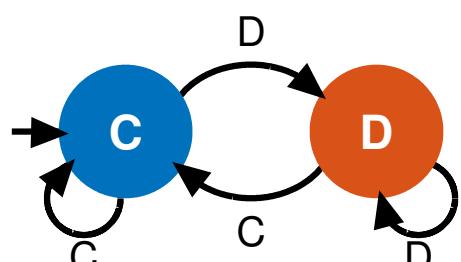
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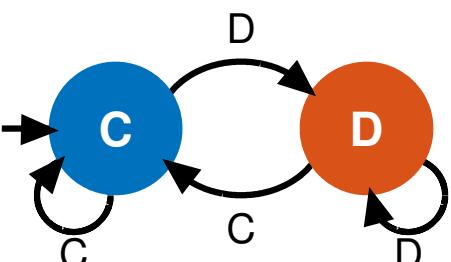
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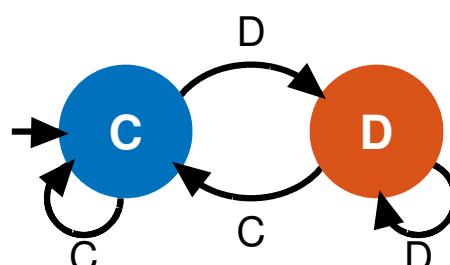
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Tit-for-Tat is a best response against itself if  $b - c > (1 - \delta)b$ , or if  $\delta > c/b$ .

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LETTERS

### TIT FOR TAT in sticklebacks and the evolution of cooperation

Manfred Milinski

Arbeitsgruppe für Verhaltensforschung, Fakultät für Biologie,  
Ruhr-Universität, Postfach 102148, 4630 Bochum 1, FRG



### Reciprocal food sharing in the vampire bat

Gerald S. Wilkinson

Department of Biology, C-016, University of California,  
San Diego, La Jolla, California 92093



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- In particular, for any pair of memory-1 strategies  $\mathbf{p}$  and  $\mathbf{q}$ , one can compute the players' payoffs  $\pi_1(\mathbf{p}, \mathbf{q})$  and  $\pi_2(\mathbf{p}, \mathbf{q})$ .

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- Approach: Let evolution determine an “optimal” strategy, by running individual-based simulations.

**A strategy of win-stay,  
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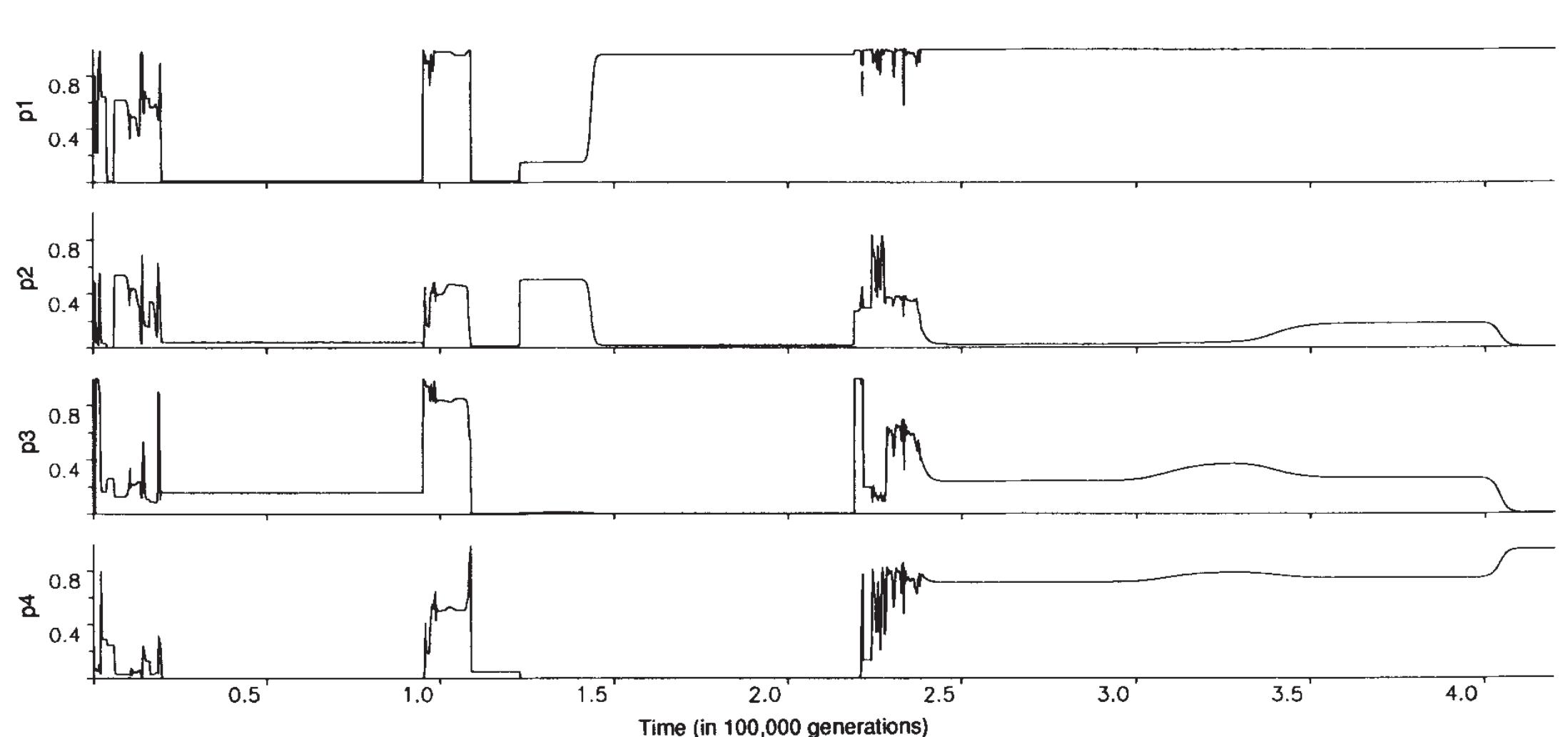
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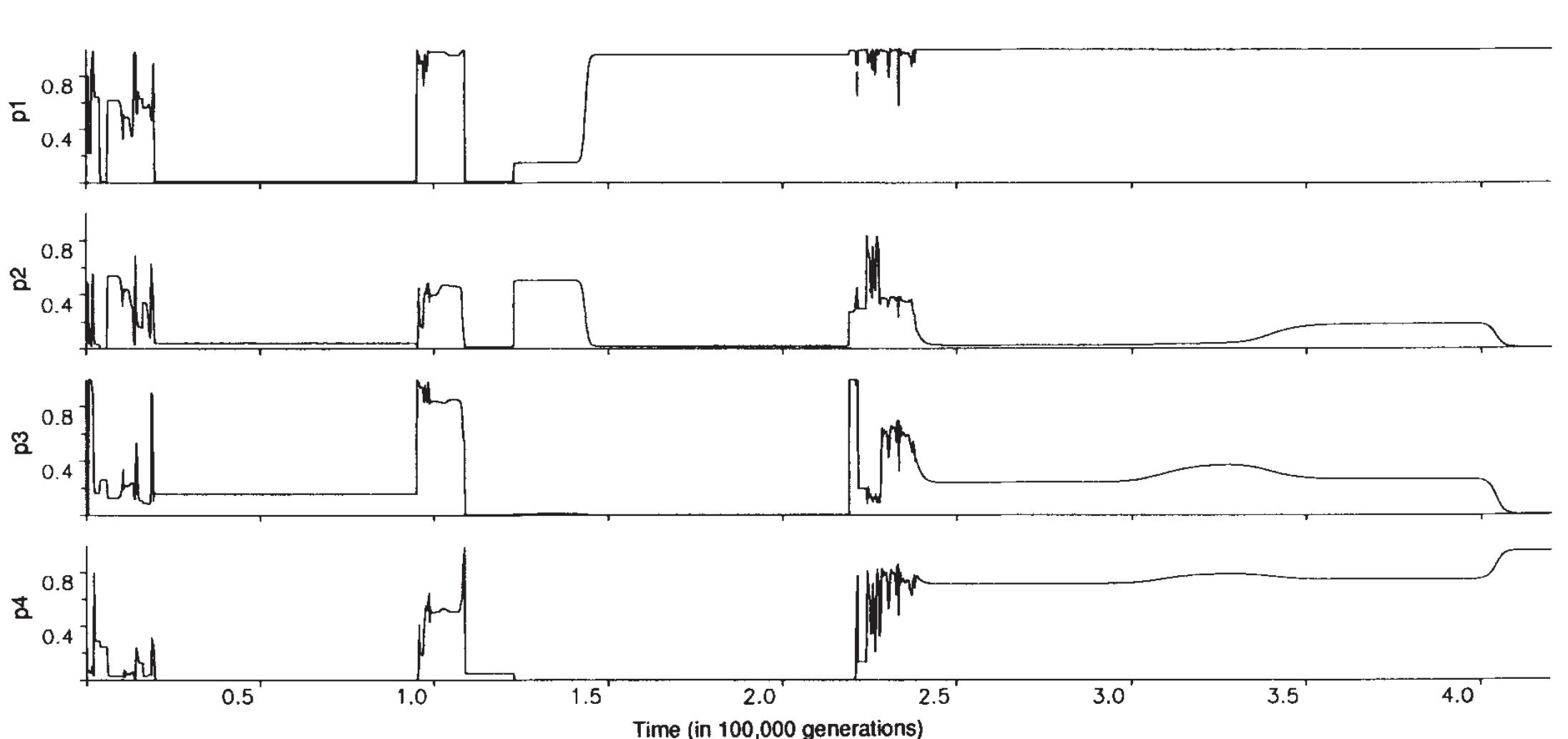
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**Win-stay Lose-shift WSLS = (1,1,0,0,1)**

Round 1: Cooperate

All other rounds:

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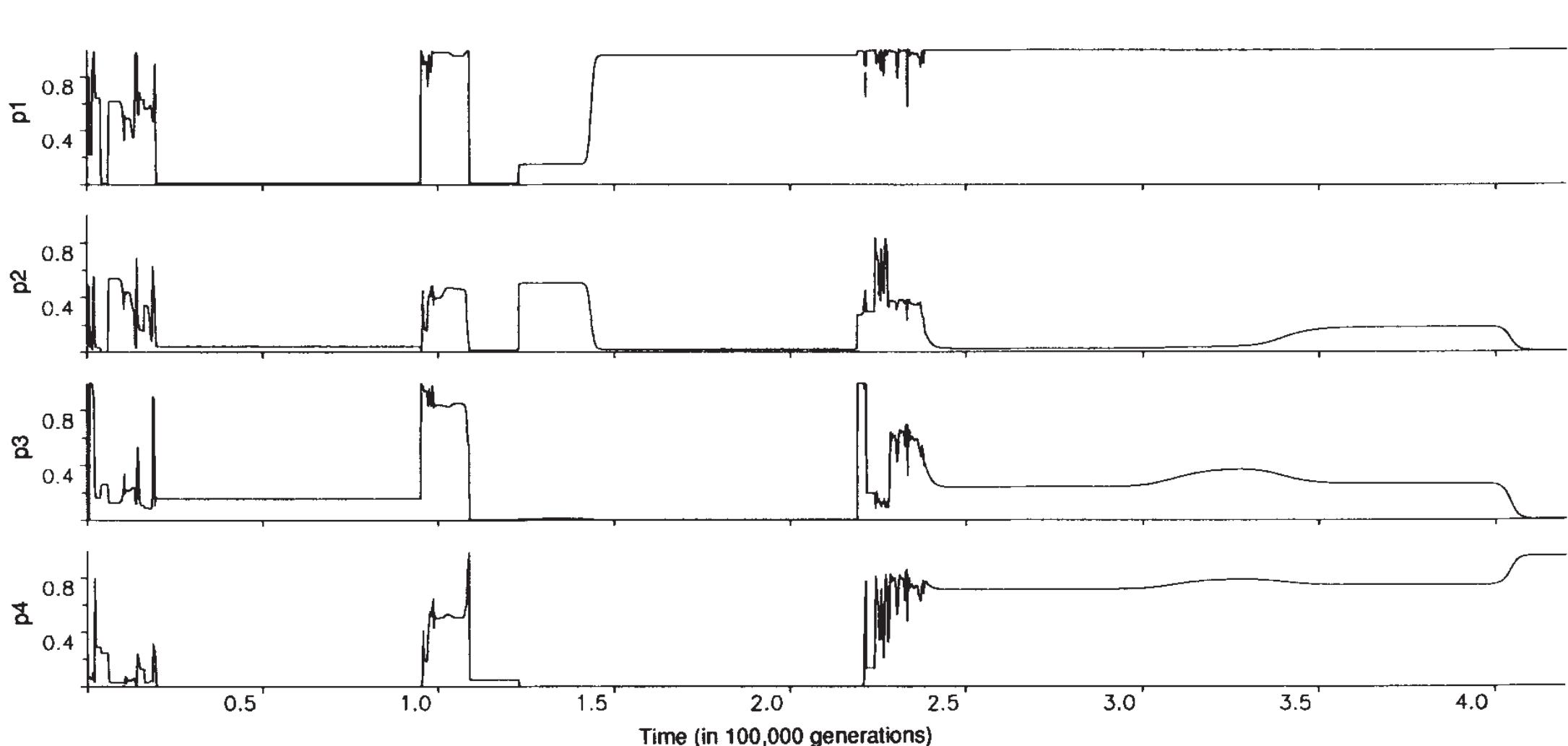
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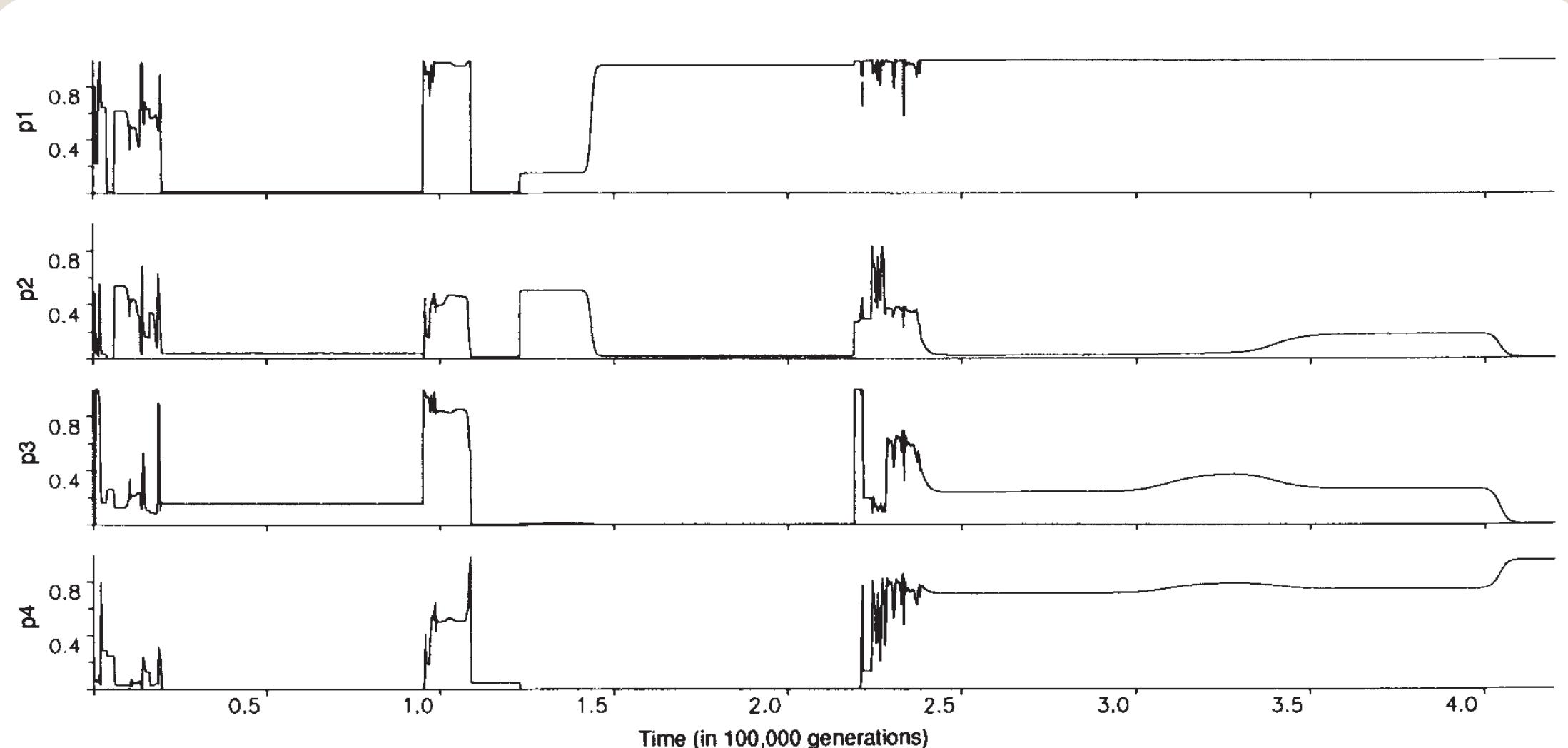
[For memory- $k$  strategies, one can find generalised versions of this strategy that are stable for  $b > \frac{k+1}{k}c$ ]

**A strategy of win-stay, lose-shift that outperforms tit-for-tat in the Prisoner’s Dilemma game**

Martin Nowak\* & Karl Sigmund†

\*Department of Zoology, University of Oxford, South Parks Road, Oxford OX1 3PS, UK

†Institut für Mathematik, Universität Wien, Strudlhofgasse 4, A-1090 Vienna, Austria



**Win-stay Lose-shift WSLS = (1,1,0,0,1)**

Round 1: Cooperate

All other rounds:

- After obtaining  $b$  or  $b-c$ , do what you did last round
- Otherwise do the opposite of what you did.

## Direct reciprocity: Direct reciprocity and extortion

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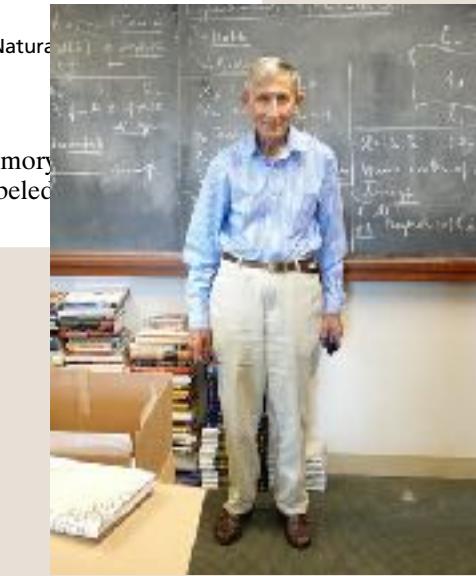
## Iterated Prisoner's Dilemma contains strategies that dominate any evolutionary opponent

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Contributed by William H. Press, April 19, 2012 (sent for review March 14, 2012)

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Then, irrespective of player 2's strategy, payoffs satisfy  $\alpha\pi_1 + \beta\pi_2 + \gamma = 0$ . Such a strategy  $\mathbf{p}$  is called a zero-determinant (ZD) strategy.



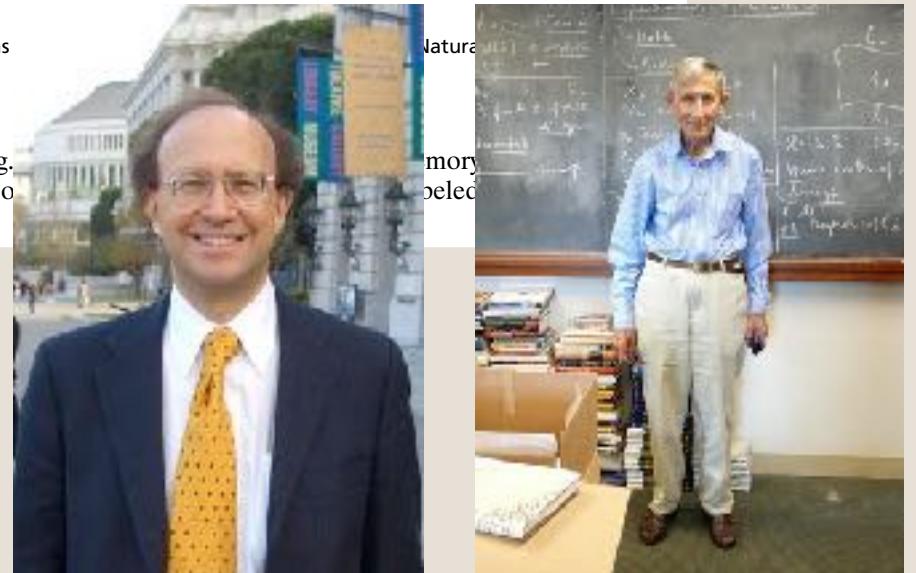
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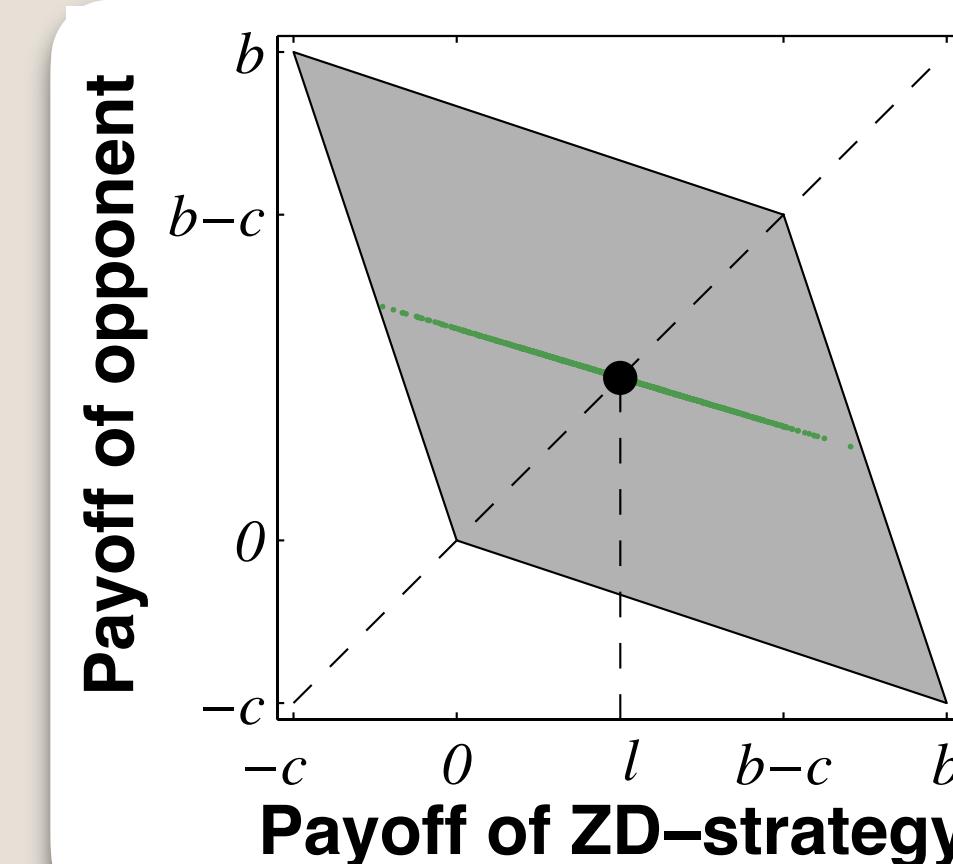
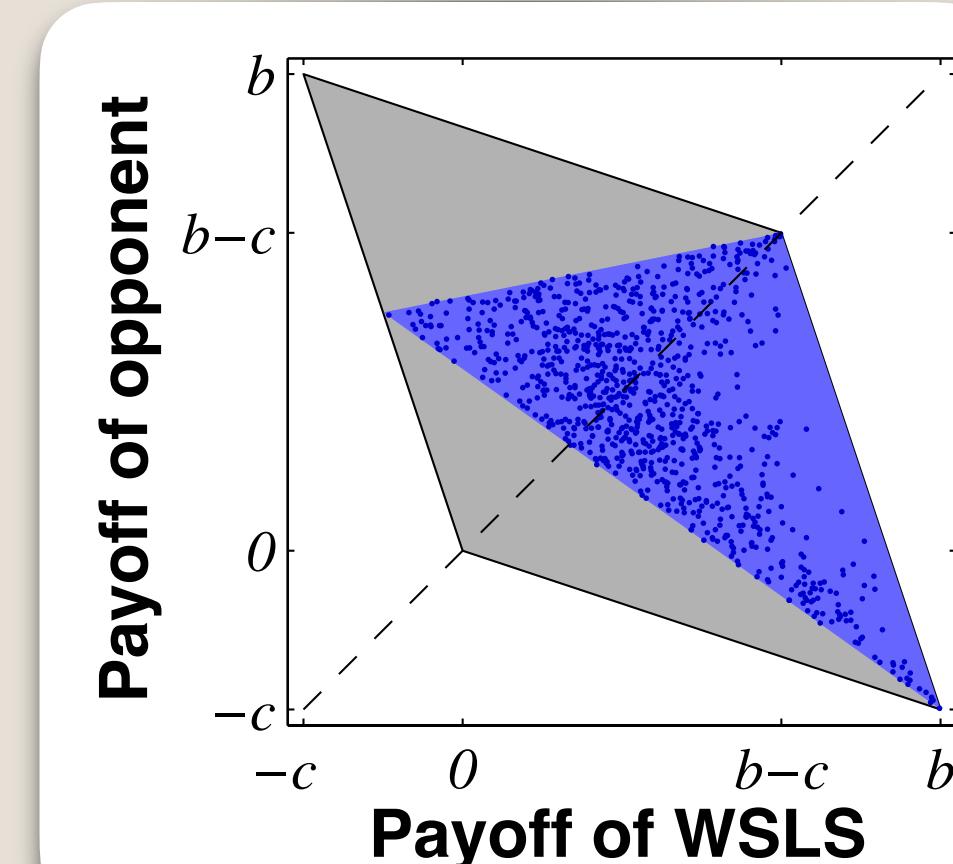
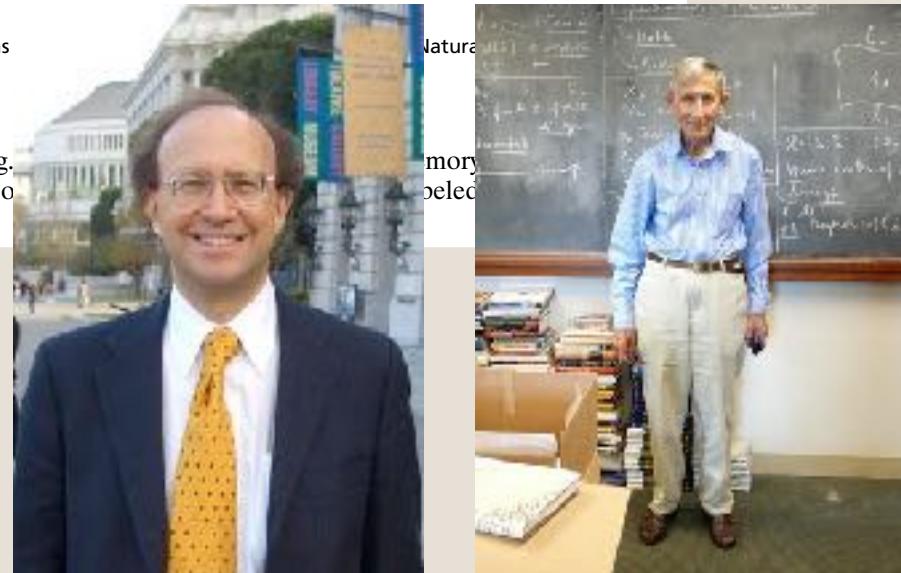
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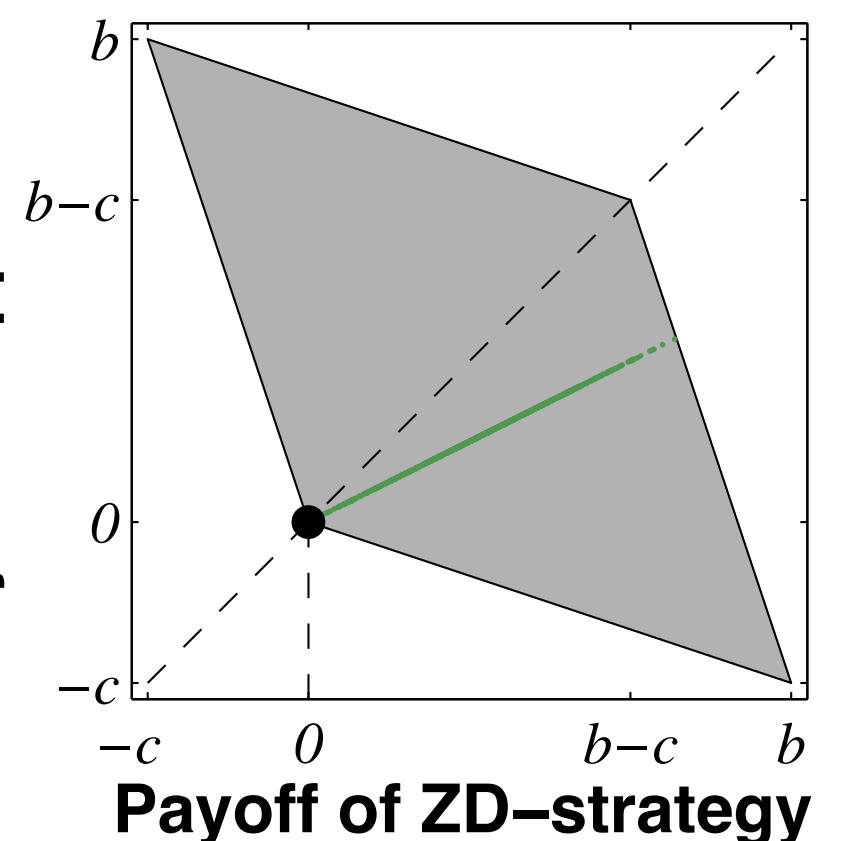
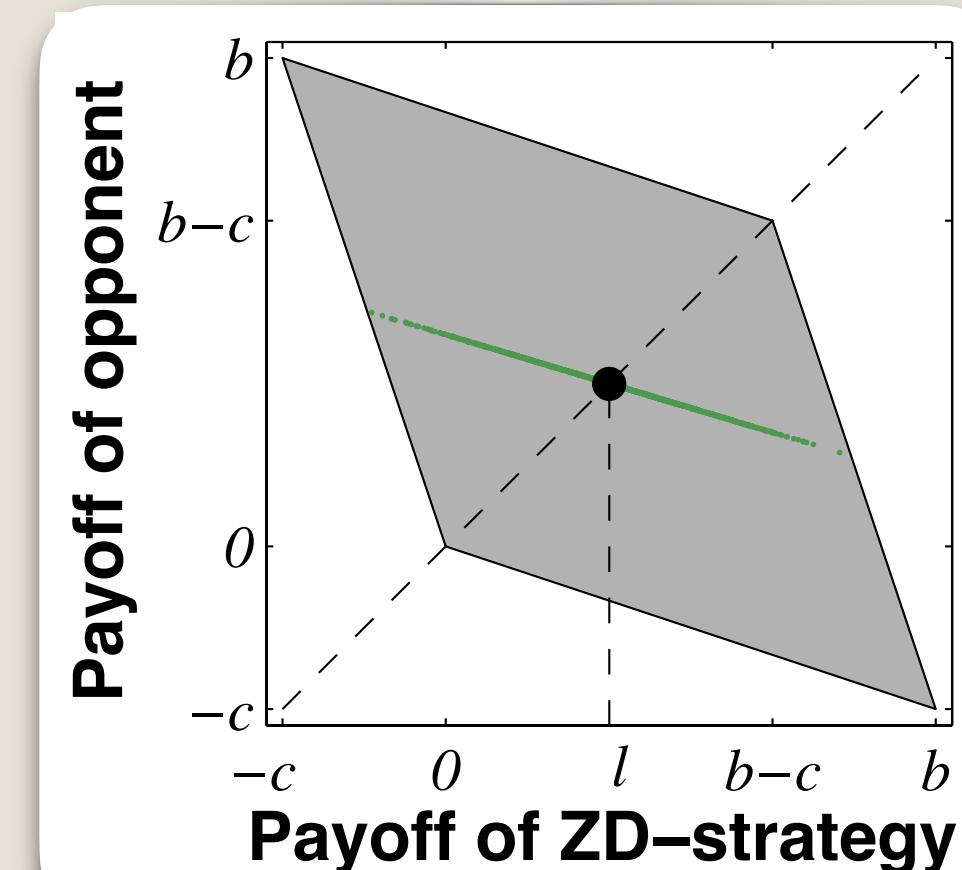
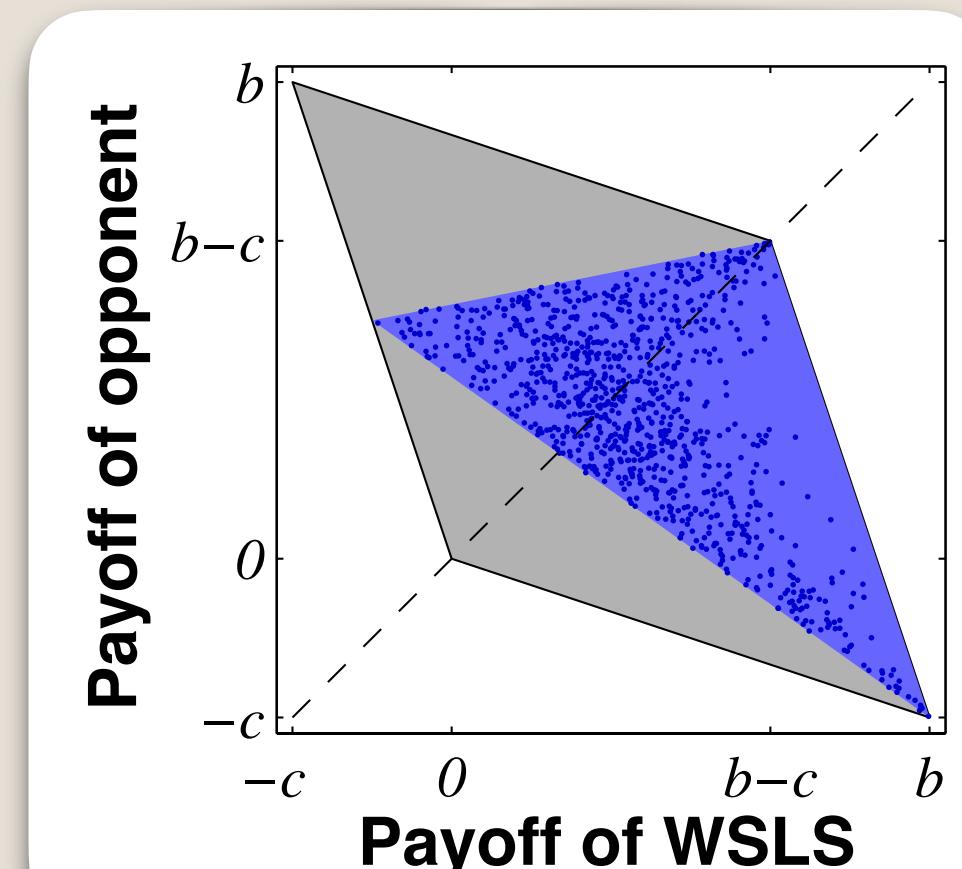


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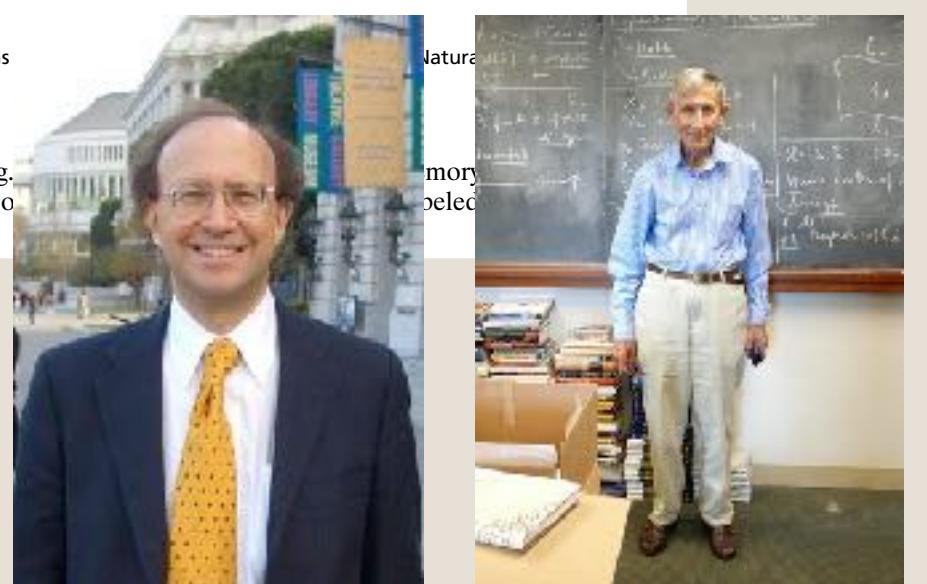
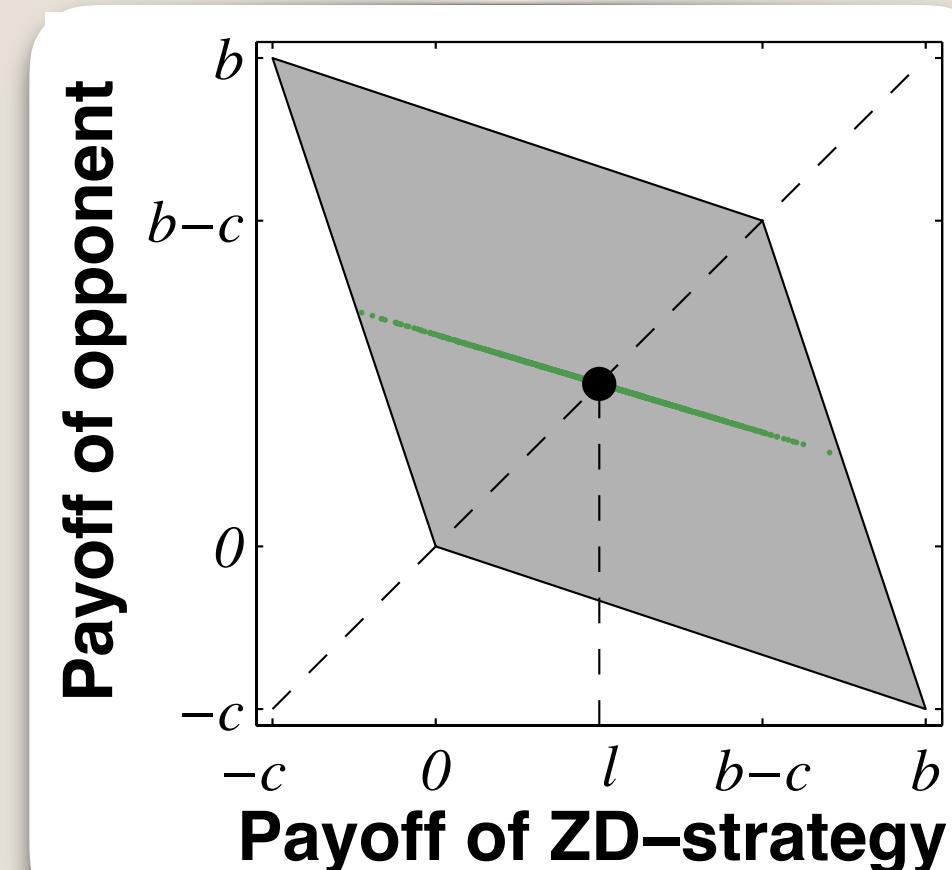
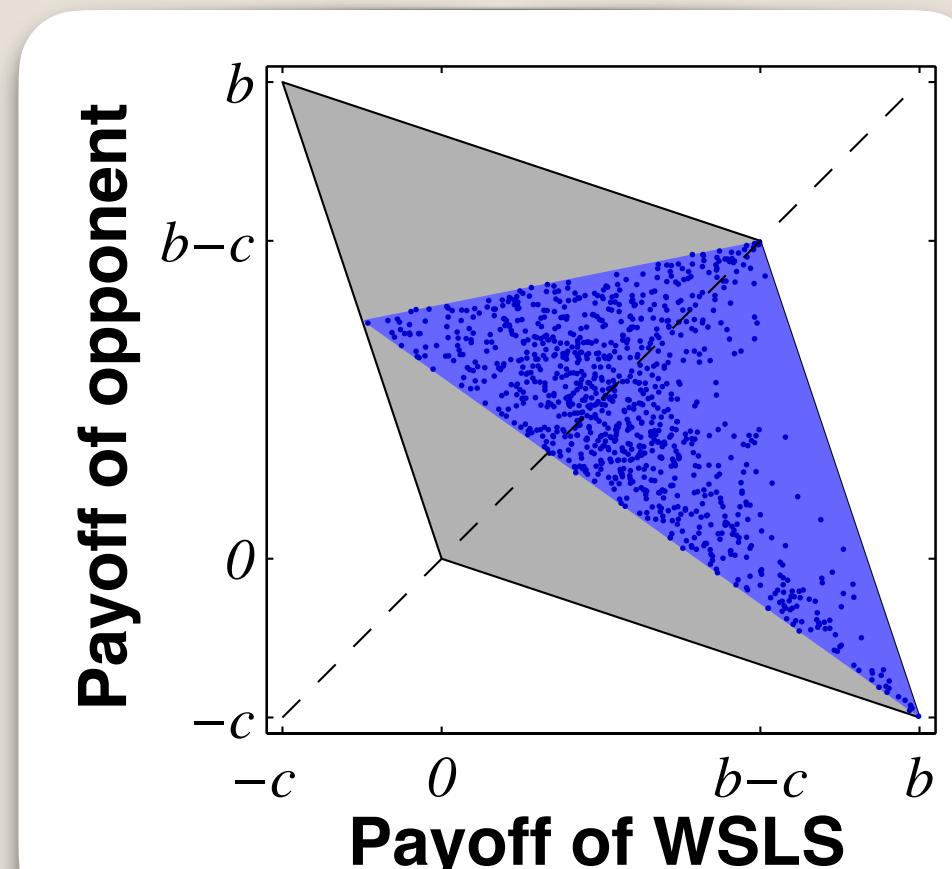
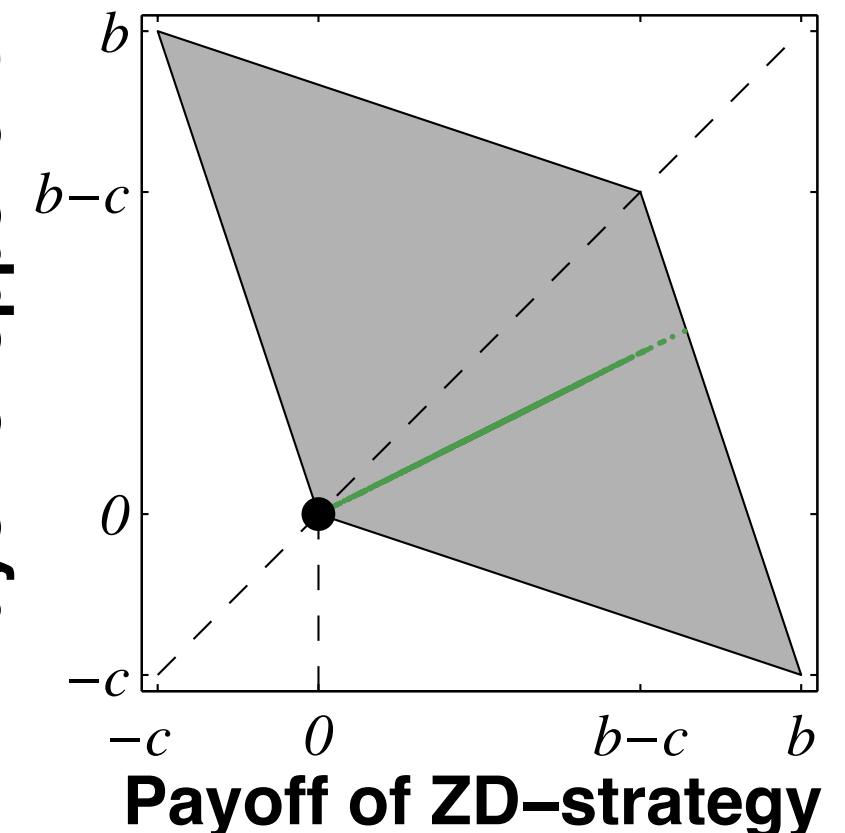


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Extortionate  
ZD strategy



# Direct reciprocity: Evolution of extortion

Evolution of extortion in Iterated Prisoner's Dilemma games

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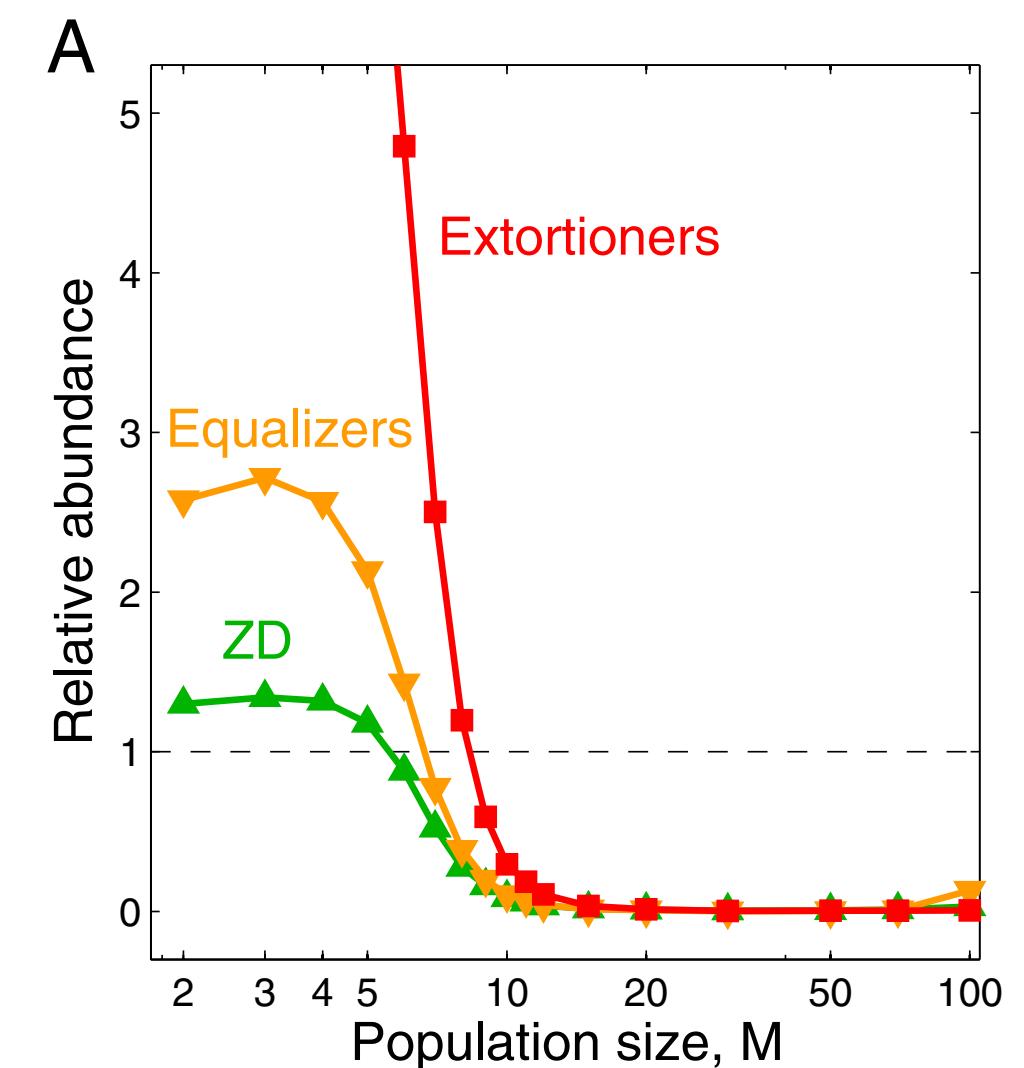
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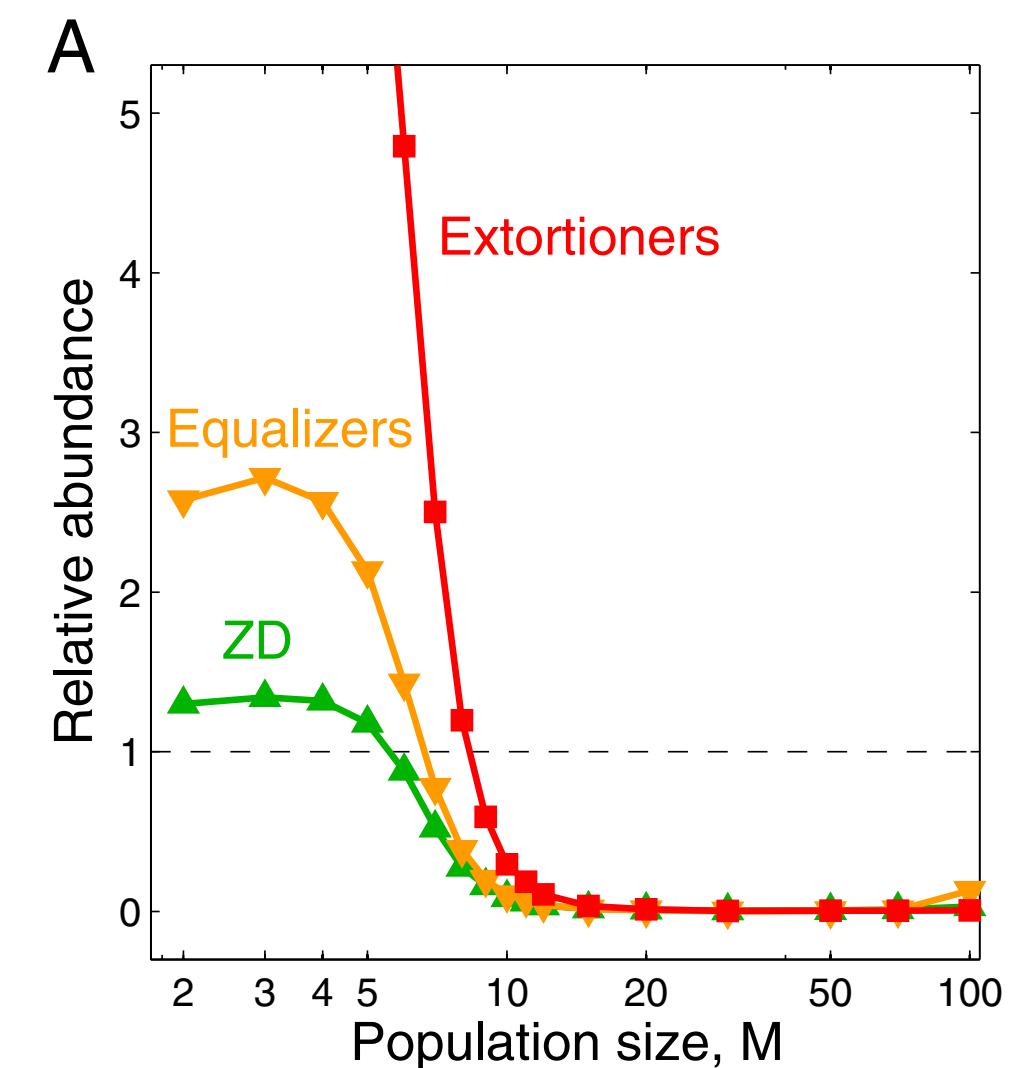
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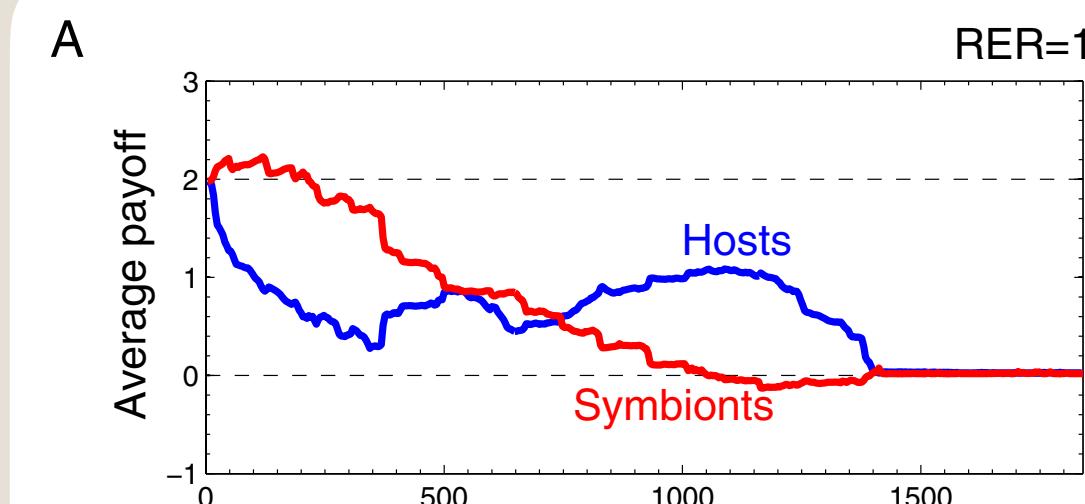
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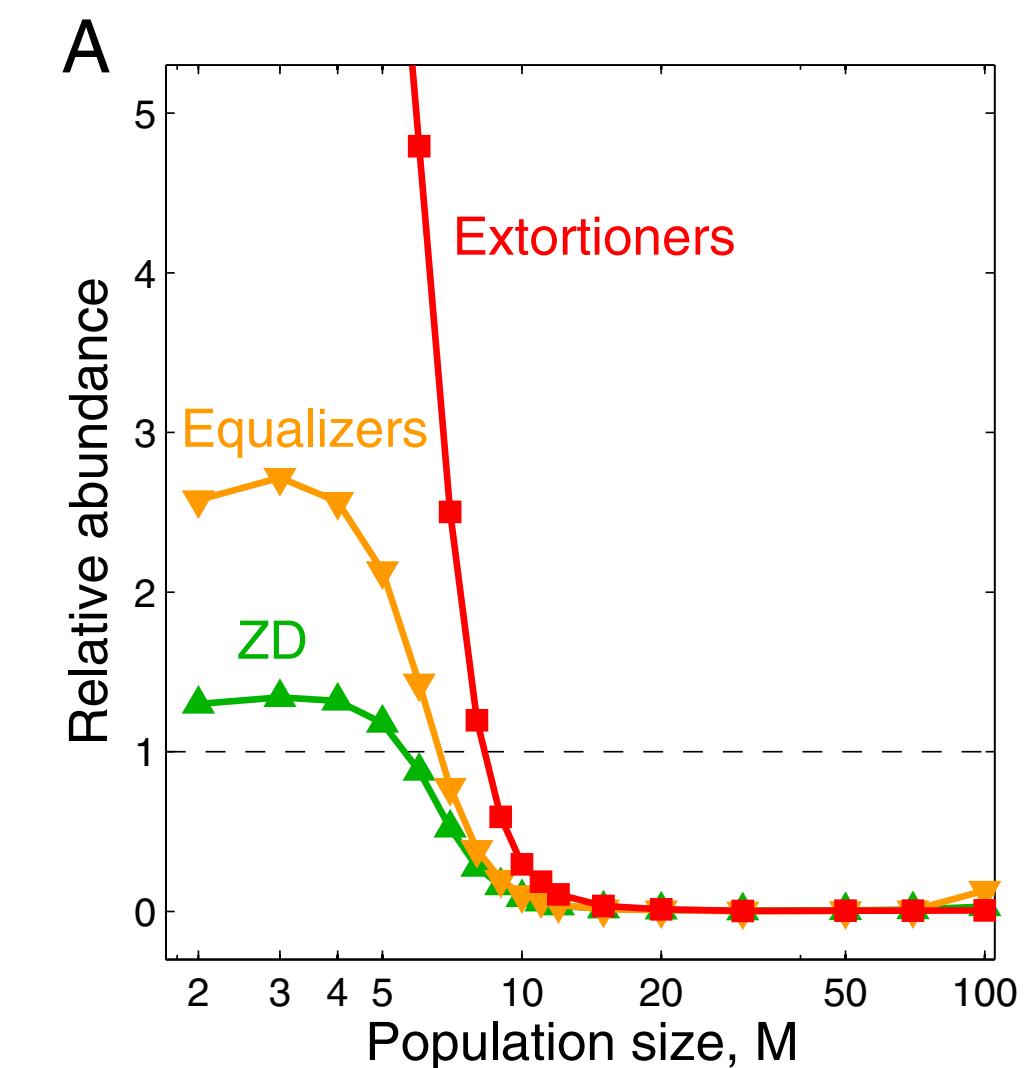
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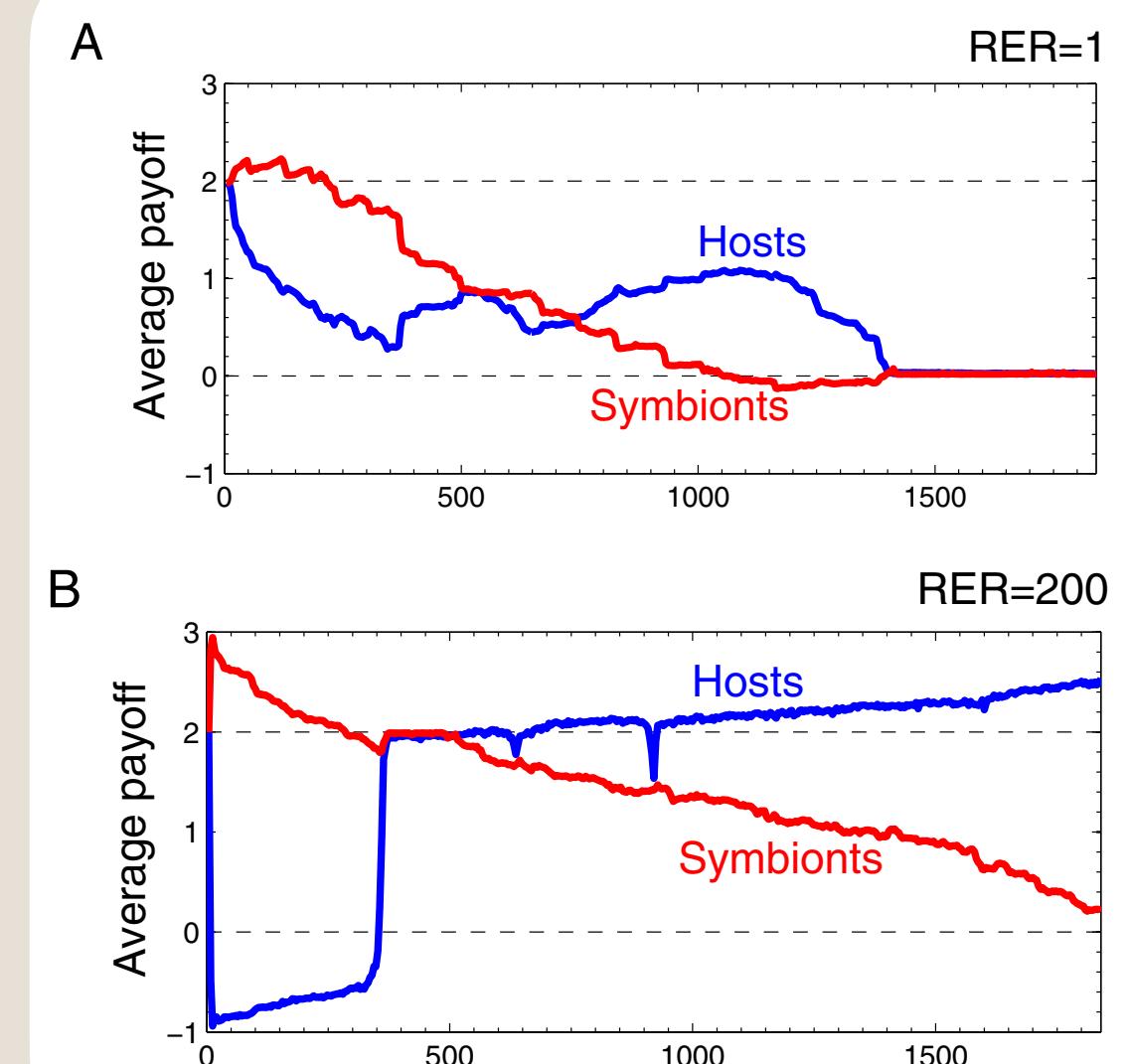
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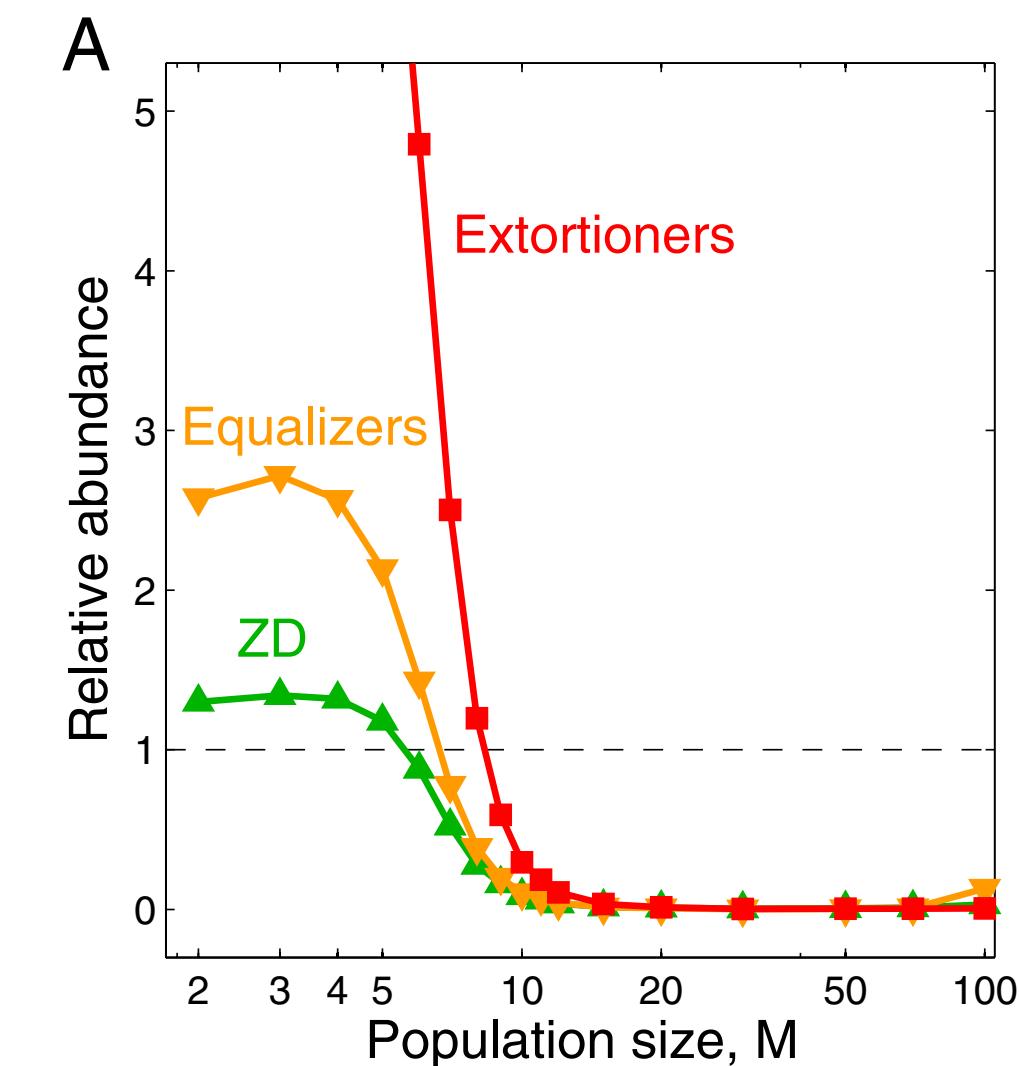


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- Result: Extortion only evolves when populations are small, or when two populations interact and one population evolves at a much slower rate ("Red king effect")

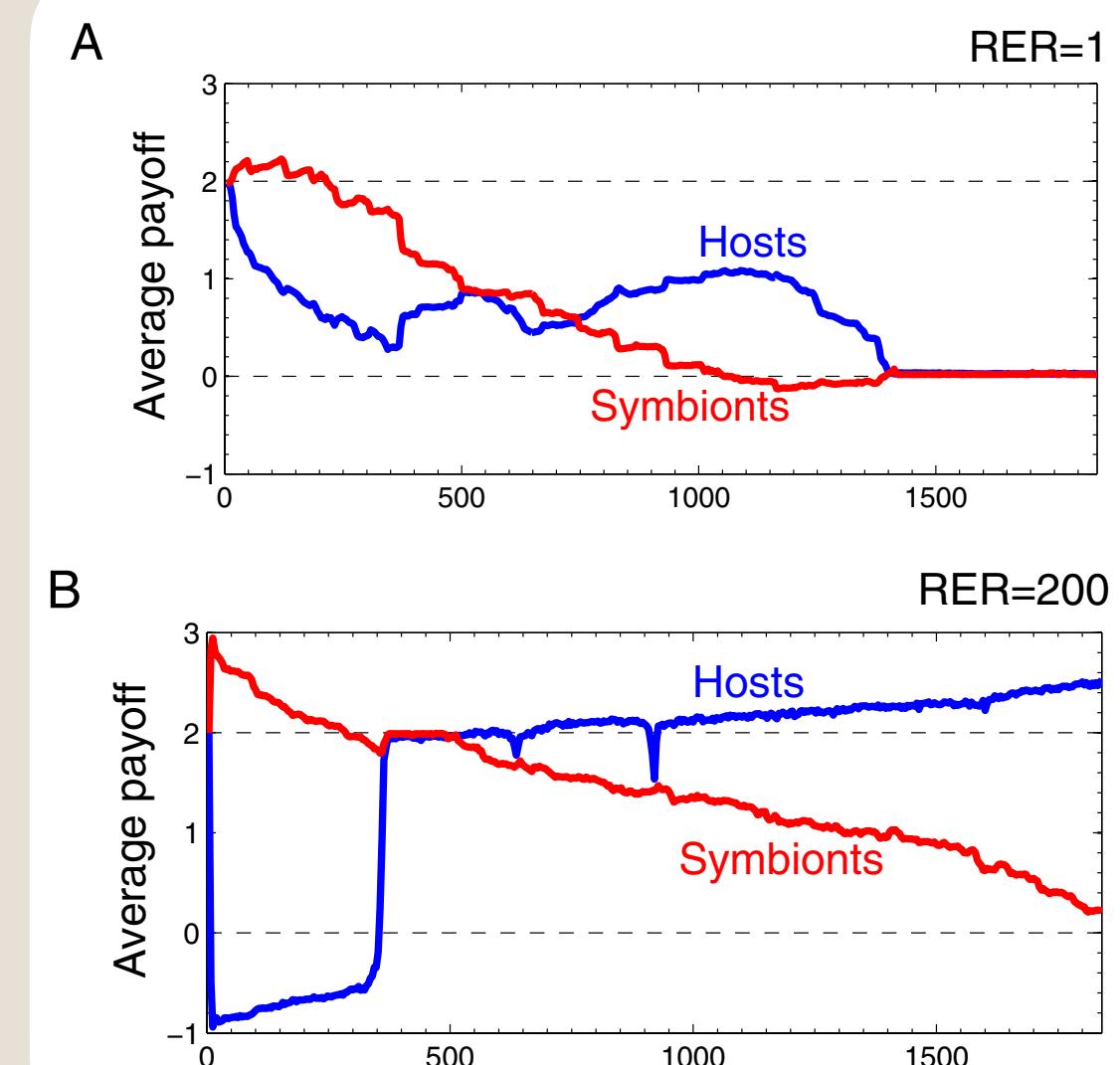
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Extortion subdues human players but is finally punished in the prisoner's dilemma

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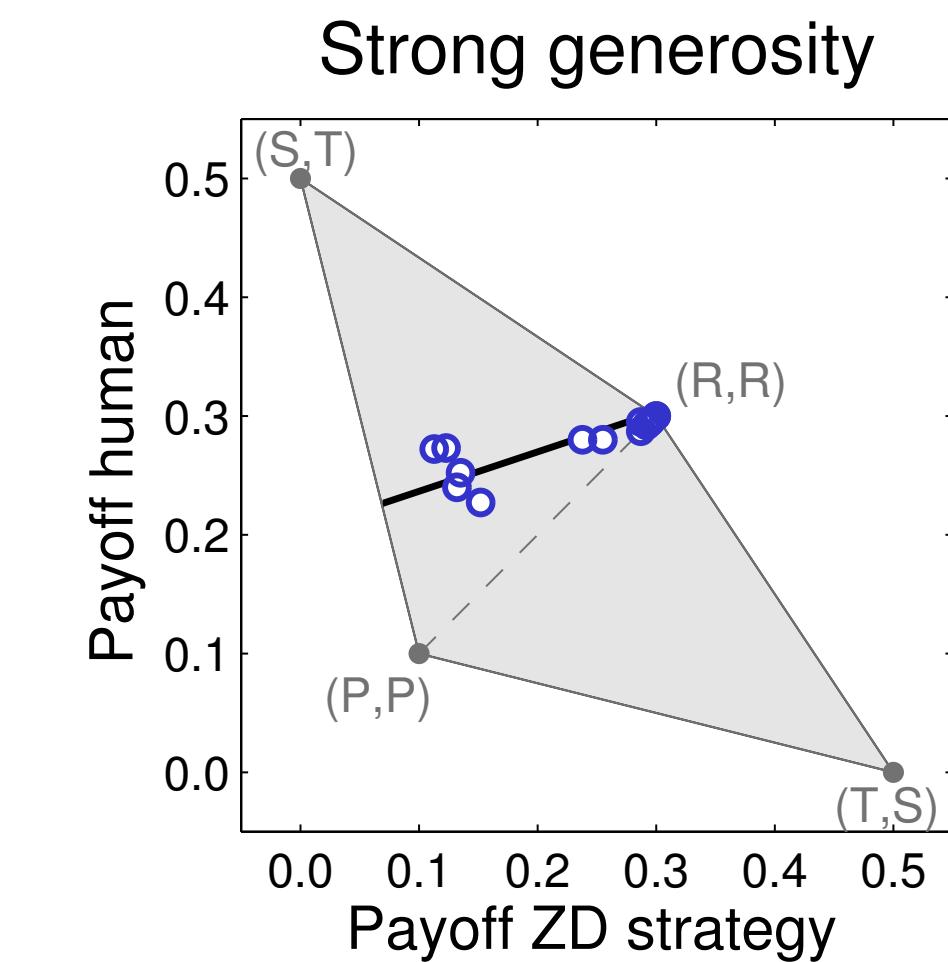
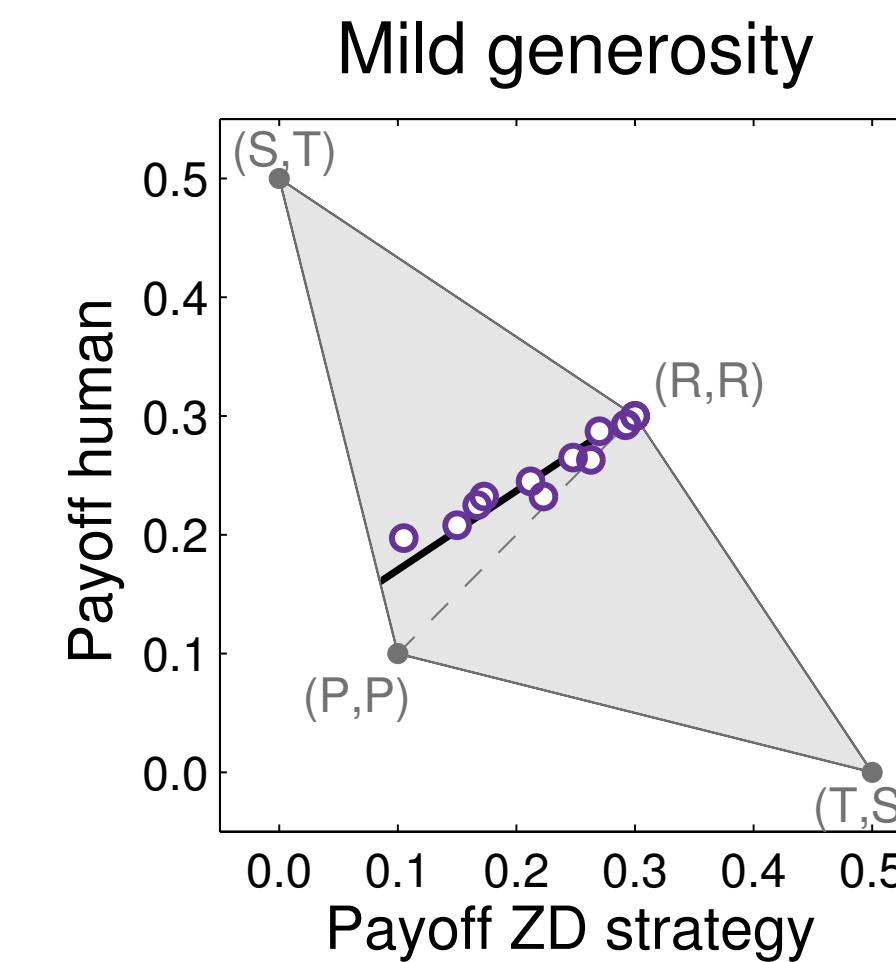
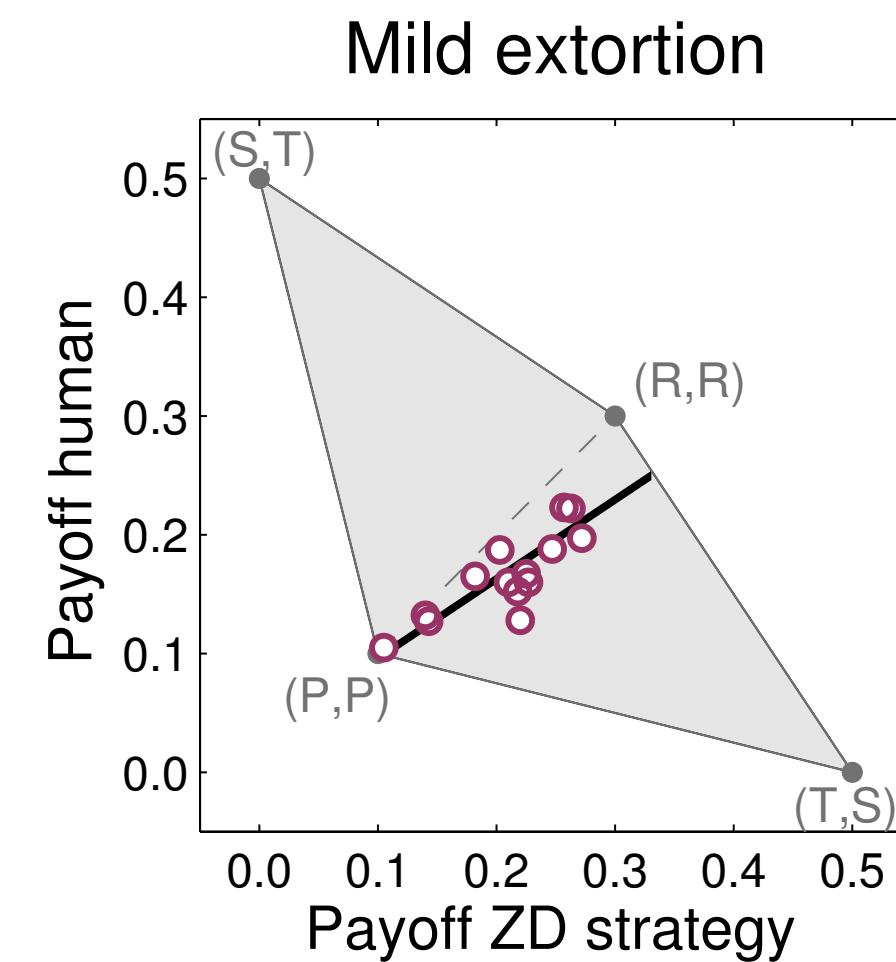
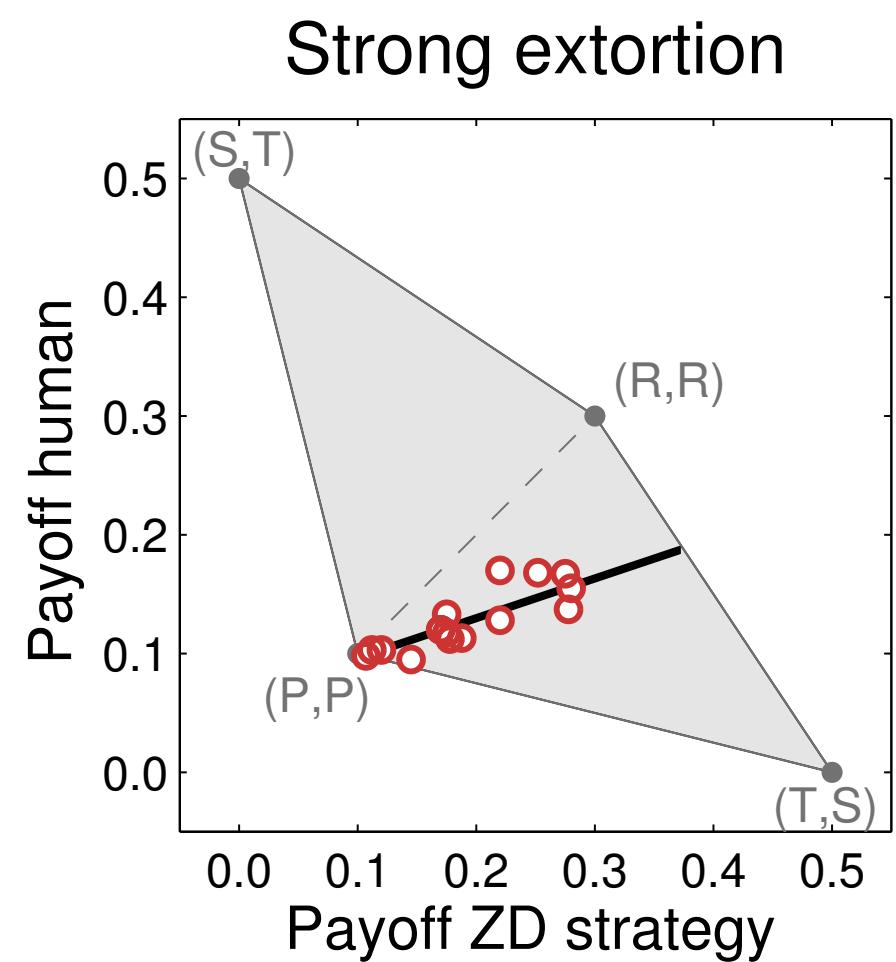
Extortion subdues human players but is finally punished in the prisoner's dilemma

Christian Hilbe<sup>1,2</sup>, Torsten Röhl<sup>1</sup> & Manfred Milinski<sup>3</sup>

## Remark 2.16. Human reactions to extortion (continued)

Prediction 1 (*Computer vs Humans*)

In the treatment with extortion, the computer should get a higher payoff than the participants. In the treatment with generosity, the human participants should get the higher payoff.



# Direct reciprocity: Extortion in the lab

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## Remark 2.16. Human reactions to extortion (continued)

Prediction 2 (*Dynamics of cooperation*)

In all treatments, the best response for humans is to cooperate in every round. Hence we would expect a general trend towards cooperation in all treatments.

# Direct reciprocity: Extortion in the lab

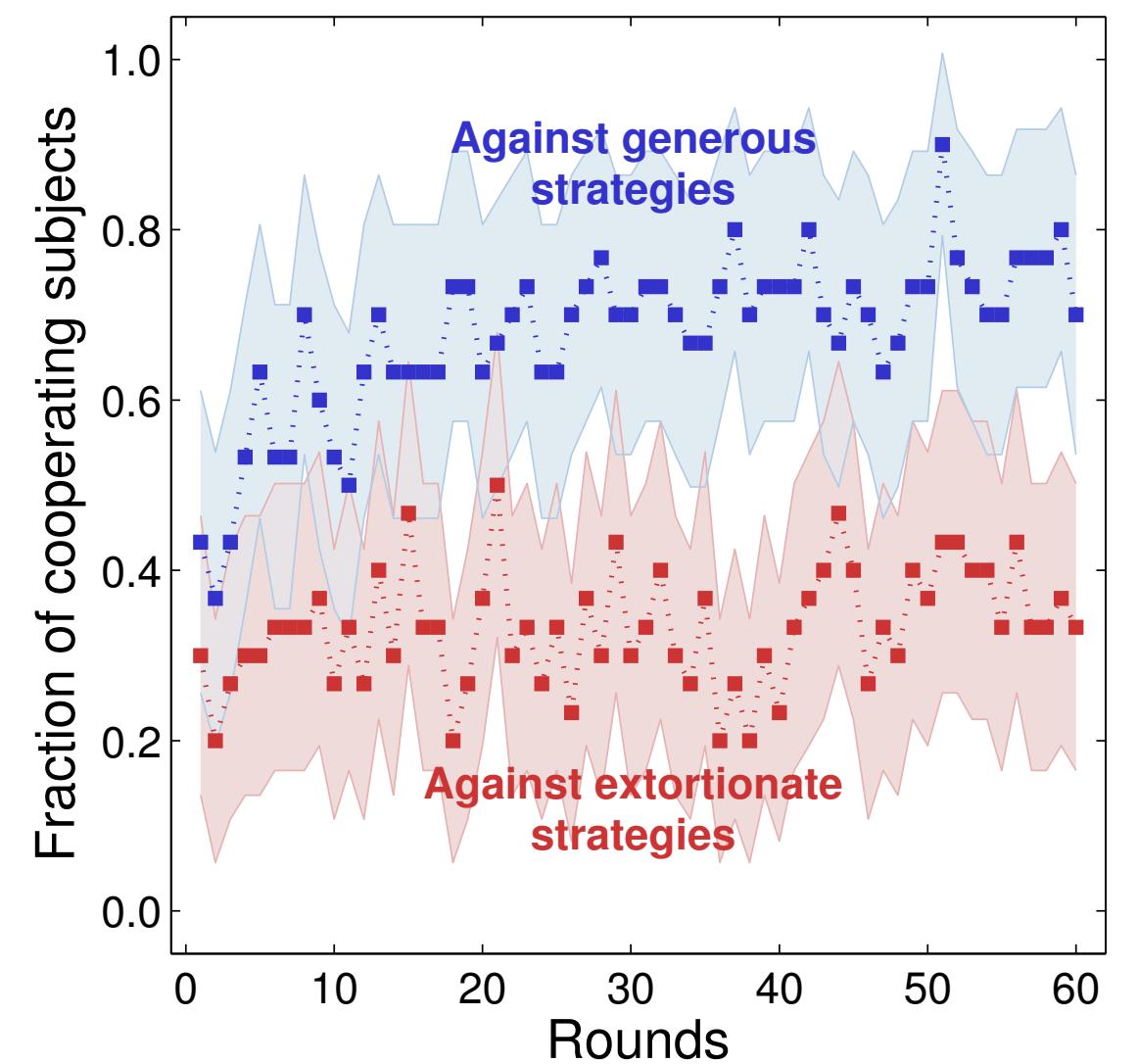
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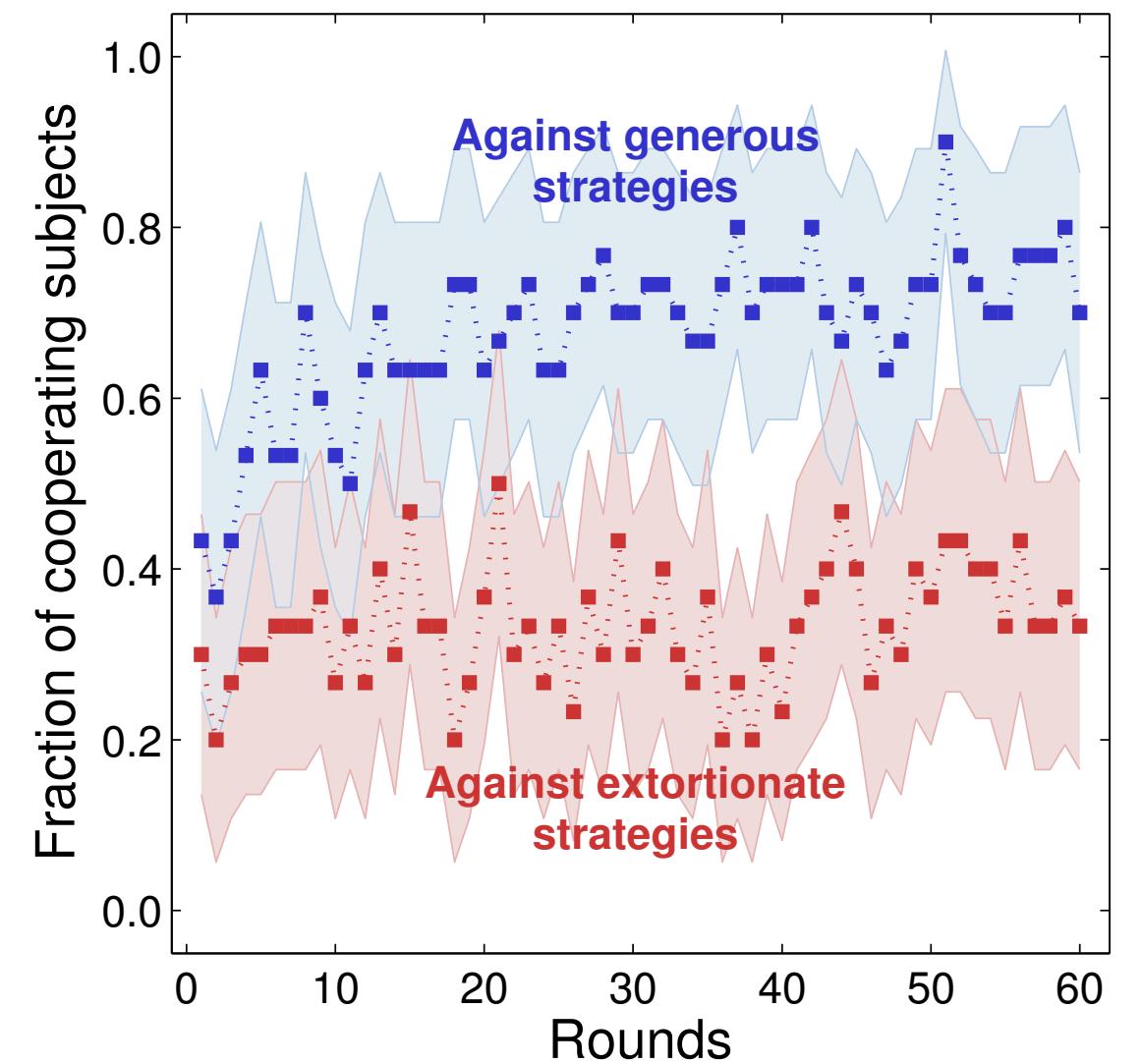
## Remark 2.16. Human reactions to extortion (continued)

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# Direct reciprocity: Extortion in the lab

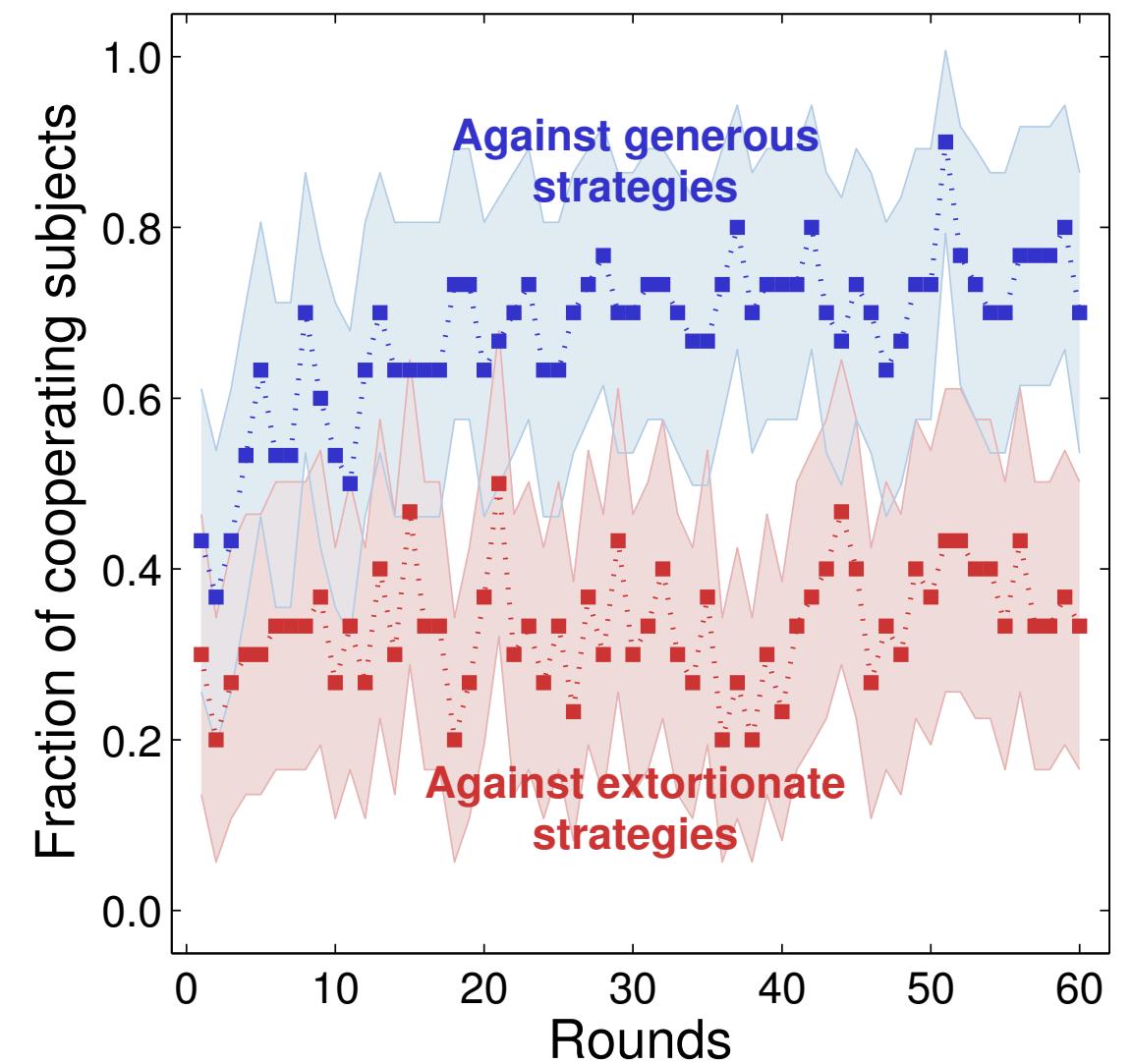
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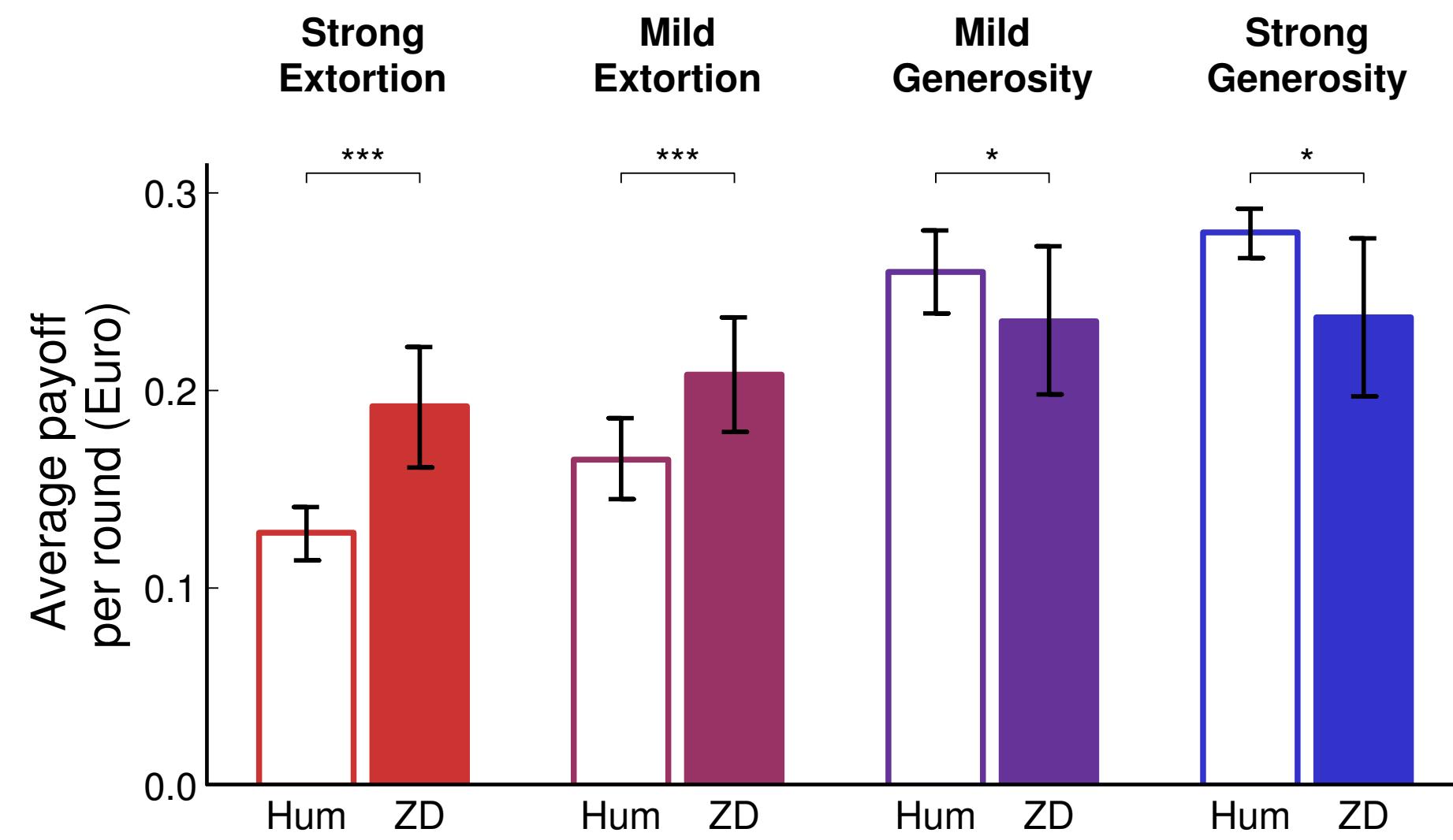
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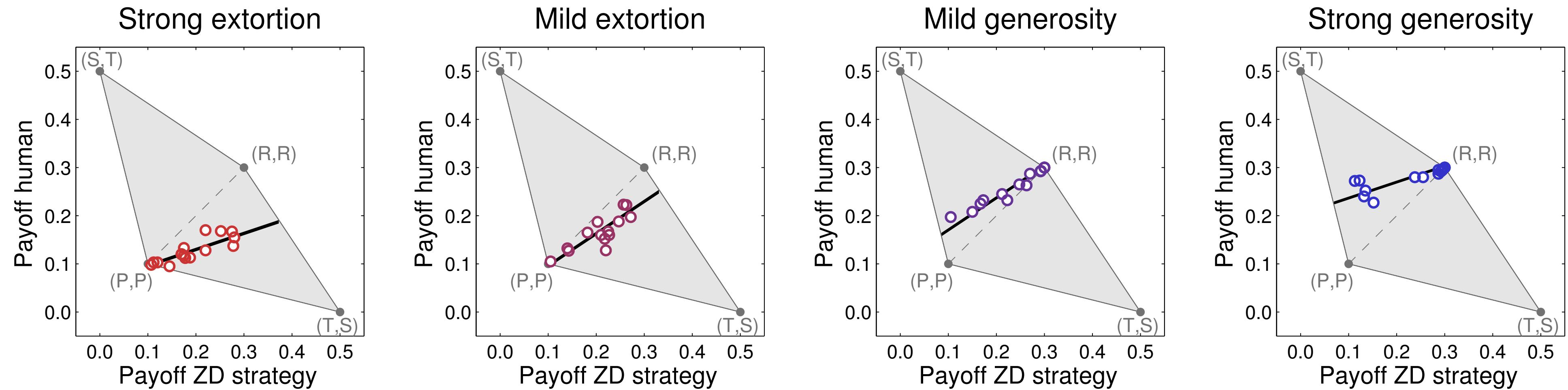
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## Remark 2.16. Human reactions to extortion (continued)

Interpretation:



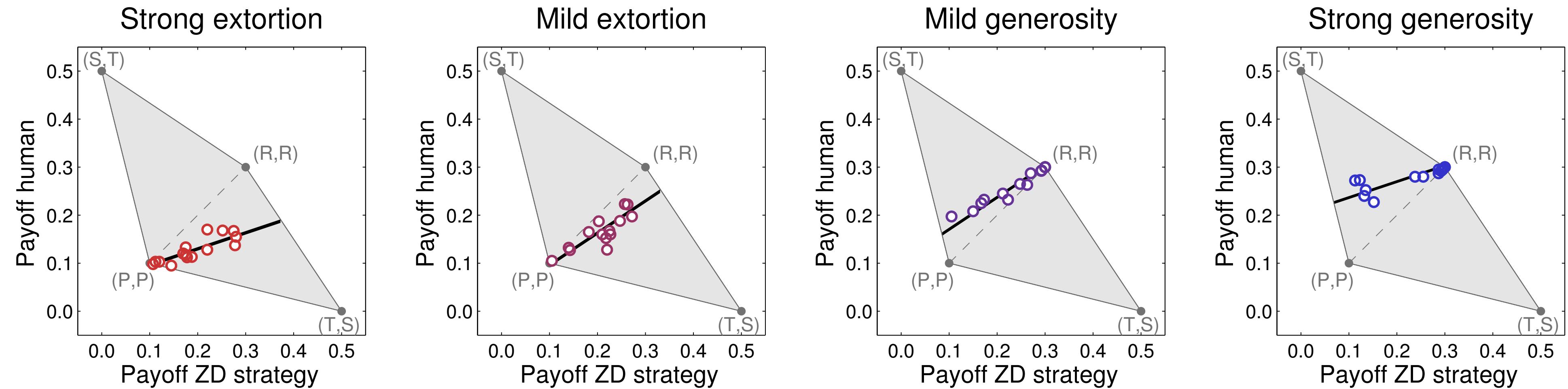
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## Remark 2.16. Human reactions to extortion (continued)

Interpretation:



- A possible explanation for these patterns is that human participants had a strong preference for fair outcomes. In the generosity treatments, payoff-maximisation and fairness are aligned. In the extortion treatments, they are mis-aligned.

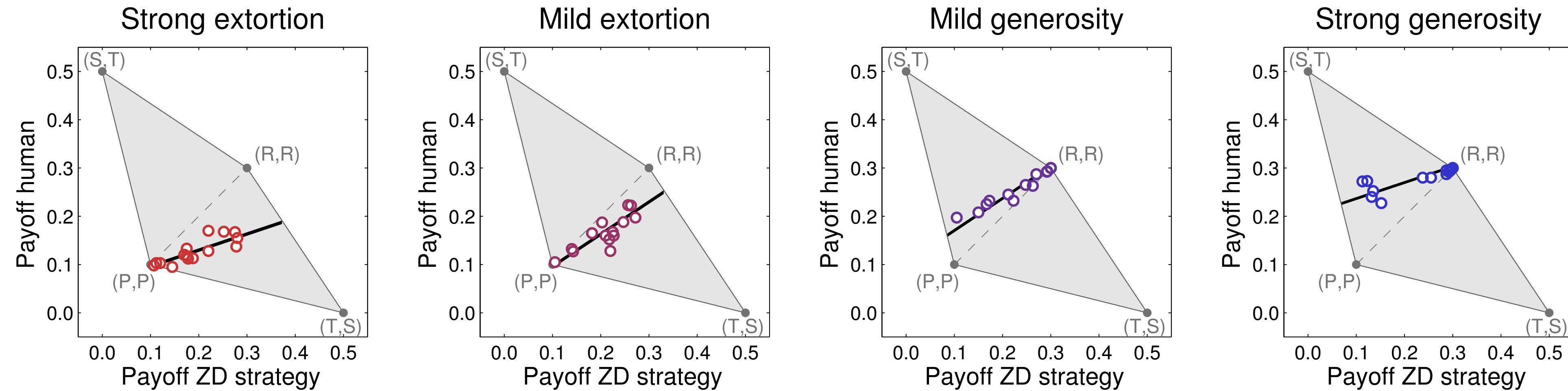
# Direct reciprocity: Extortion in the lab

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## Remark 2.16. Human reactions to extortion (continued)

Interpretation:

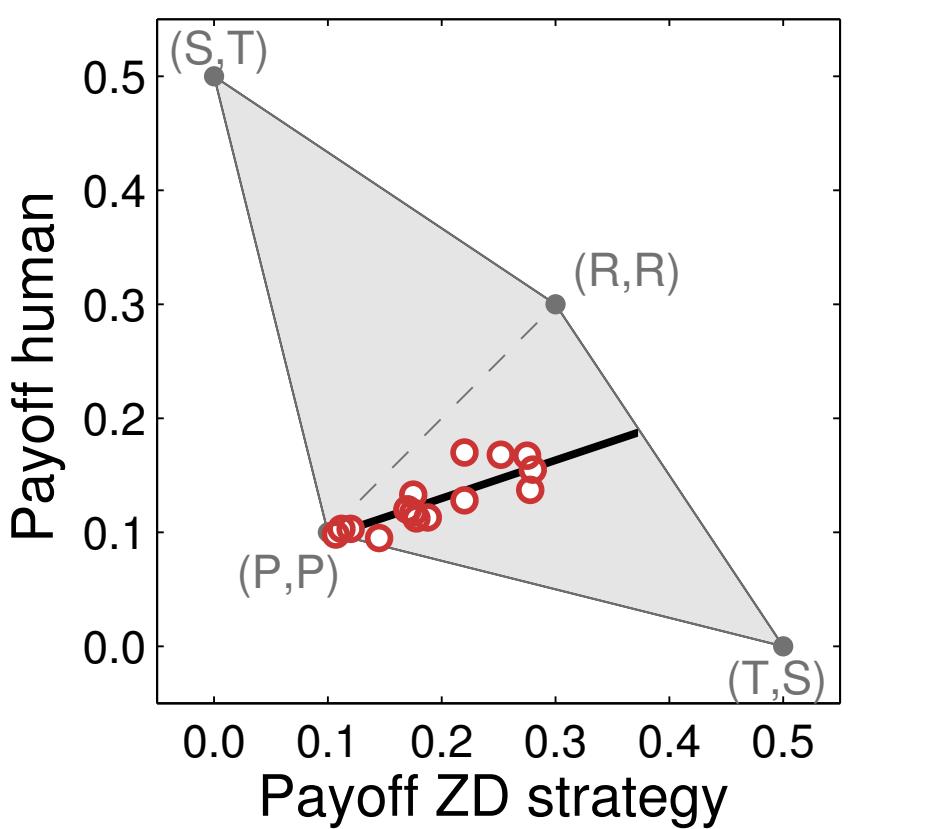
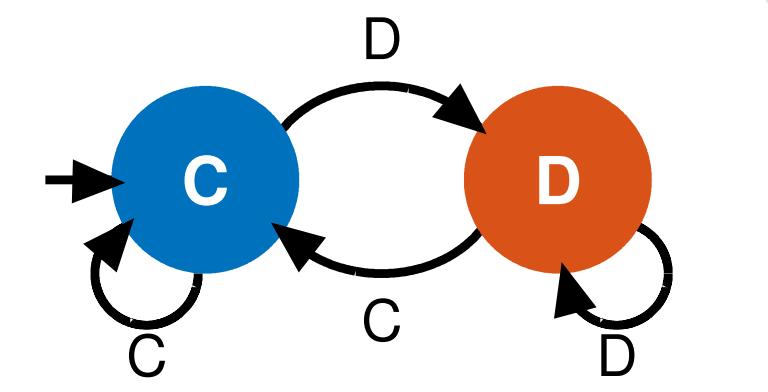


- A possible explanation for these patterns is that human participants had a strong preference for fair outcomes. In the generosity treatments, payoff-maximisation and fairness are aligned. In the extortion treatments, they are mis-aligned.
- In line with this interpretation, the effect vanishes if human participants receive information about the nature of their opponent before the experiment.

## Summary

Some things you should have learned today:

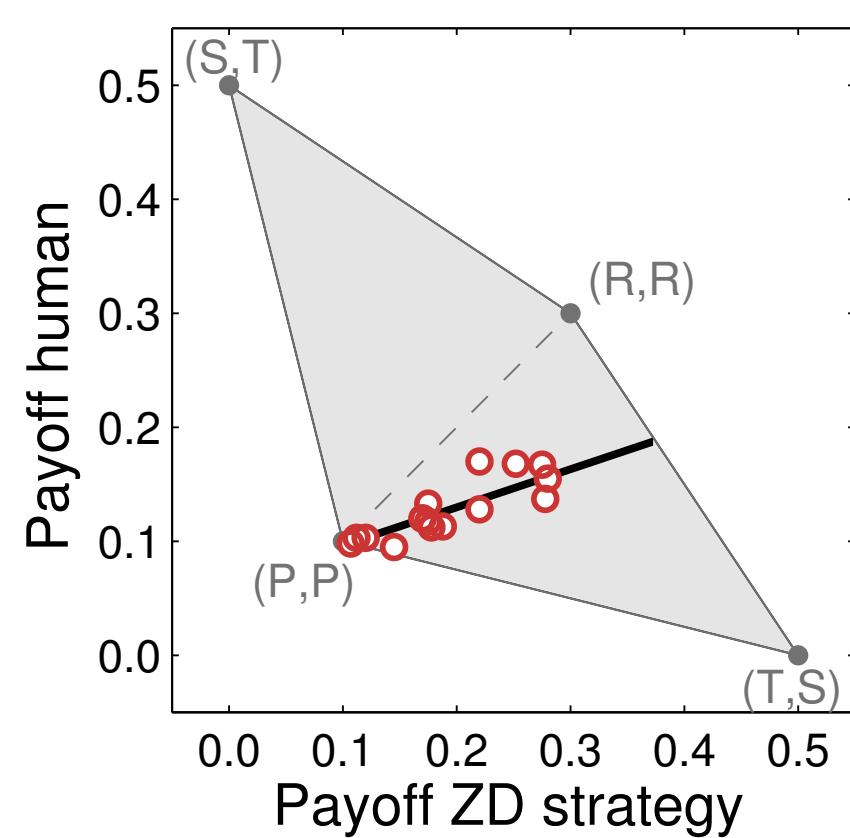
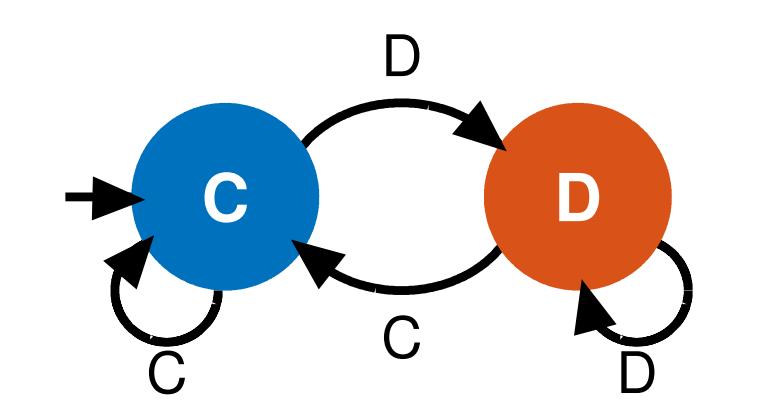
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## Summary

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1. We have used our techniques (Defining a game, characterizing Nash equilibria, evolutionary dynamics) to explore why people may cooperate.
2. One mechanism for cooperation is direct reciprocity. It is based on the intuition that cooperation can in fact be a profitable strategy when people interact repeatedly (no conscious decision-making required!)



## Summary

Some things you should have learned today:

1. We have used our techniques (Defining a game, characterizing Nash equilibria, evolutionary dynamics) to explore why people may cooperate.
2. One mechanism for cooperation is direct reciprocity. It is based on the intuition that cooperation can in fact be a profitable strategy when people interact repeatedly (no conscious decision-making required!)
3. Extortionate (zero-determinant) strategies have interesting mathematical properties, but against humans they don't pay.

