

VII. Cooperation & Competition

Game Theory and the Iterated Prisoner's Dilemma

The Rudiments of Game Theory

Leibniz on Game Theory

- “Games combining chance and skill give the best representation of human life, particularly of military affairs and of the practice of medicine which necessarily depend partly on skill and partly on chance.” — Leibniz (1710)
- “... it would be desirable to have a complete study made of games, treated mathematically.”
— Leibniz (1715)



Origins of Modern Theory



- 1928: John von Neumann: optimal strategy for two-person zero-sum games
 - von Neumann: mathematician & pioneer computer scientist (CAs, “von Neumann machine”)
- 1944: von Neumann & Oskar Morgenstern: *Theory of Games and Economic Behavior*
 - Morgenstern: famous mathematical economist
- 1950: John Nash: *Non-cooperative Games*
 - his PhD dissertation (27 pages)
 - “genius,” Nobel laureate (1994), schizophrenic

Classification of Games

- **Games of Chance**
 - outcome is independent of players' actions
 - “uninteresting” (apply probability theory)
- **Games of Strategy**
 - outcome is at least partially dependent on players' actions
 - completely in chess
 - partially in poker

Classification of Strategy Games

- Number of players ($1, 2, 3, \dots, n$)
- Zero-sum or non zero-sum
- Essential or inessential
- Perfect or imperfect information

Zero-sum vs. Non Zero-sum

- **Zero-sum**: winnings of some is exactly compensated by losses of others
 - sum is zero for *every* set of strategies
- **Non zero-sum**:
 - positive sum (mutual gain)
 - negative sum (mutual loss)
 - constant sum
 - nonconstant sum (variable gain or loss)

Essential vs. Inessential

- **Essential**: there is an advantage in forming coalitions
 - may involve agreements for payoffs, cooperation, etc.
 - can happen in zero-sum games only if $n \geq 3$ (obviously!)
- **Inessential**: there is no such advantage
 - “everyone for themselves”

Perfect vs. Imperfect Information

- **Perfect information:** everyone has complete information about all previous moves
- **Imperfect information:** some or all have only partial information
 - players need not have complete information even about themselves (e.g. bridge)

Strategies

- **Strategy**: a complete sequence of actions for a player
- **Pure** strategy: the plan of action is completely determined
 - for each situation, a specific action is prescribed
 - disclosing the strategy might or might not be disadvantageous
- **Mixed** strategy: a probability is assigned to each plan of action

Von Neumann's Solution for Two-person Zero-sum Games

Maximin Criterion

- Choose the strategy that *maximizes* the *minimum* payoff
- Also called *minimax*: minimize the maximum loss
 - since it's zero-sum, your loss is the negative of your payoff
 - pessimistic?

Example

- Two mineral water companies competing for same market
- Each has fixed cost of \$5 000 (regardless of sales)
- Each company can charge \$1 or \$2 per bottle
 - at price of \$2 can sell 5 000 bottles, earning \$10 000
 - at price of \$1 can sell 10 000 bottles, earning \$10 000
 - if they charge same price, they split market
 - otherwise all sales are of lower priced water
 - payoff = revenue – \$5 000

Payoff Matrix

		Perrier	
		price = \$1	price = \$2
Apollinaris	price = \$1	0, 0	5000, -5000
	price = \$2	-5000, 5000	0, 0

Maximin for A.

<p>minimum at \$1</p> <p>Maximin</p> <p>minimum at \$2</p>		Perrier	
		price = \$1	price = \$2
Apollinaris	price = \$1	0, 0	5000, -5000
	price = \$2	-5000, 5000	0, 0

Maximin for P.

		Perrier	
		price = \$1	price = \$2
Apollinaris	price = \$1	0, 0	5000, -5000
	price = \$2	-5000, 5000	0, 0

Maximin Equilibrium

		Perrier	
		price = \$1	price = \$2
Apollinaris	price = \$1	0, 0	5000, -5000
	price = \$2	-5000, 5000	0, 0

Implications of the Equilibrium

- If both companies act “rationally,” they will pick the equilibrium prices
- If either behaves “irrationally,” the other will benefit (if it acts “rationally”)

Reading

- CS 420/594: Flake, ch. 18 (Natural & Analog Computation)

Matching Pennies

- Al and Barb each independently picks either heads or tails
- If they are both heads or both tails, Al wins
- If they are different, Barb wins

Payoff Matrix

Minimum of each pure strategy is the same		Barb	
		head	tail
Al	head	+1, -1	-1, +1
	tail	-1, +1	+1, -1

Mixed Strategy

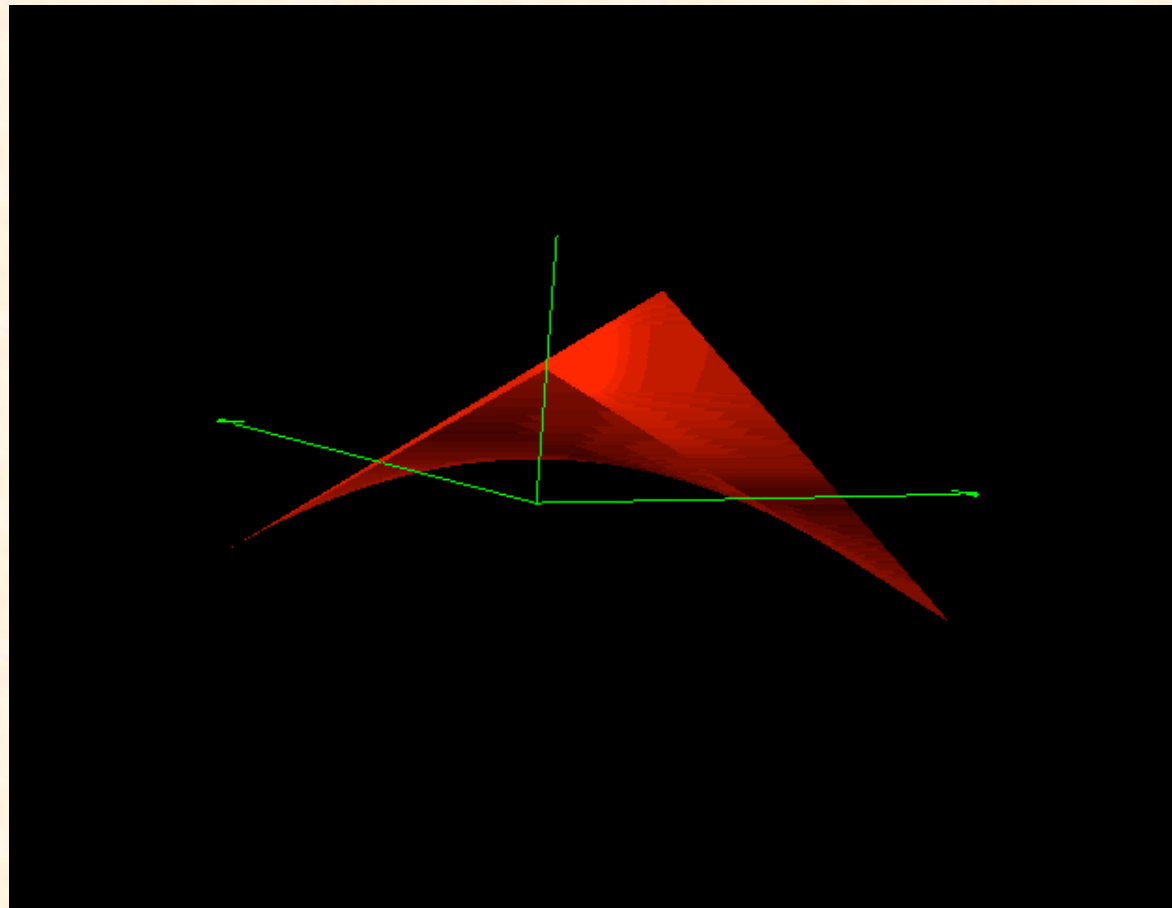
- Although we cannot use maximin to select a pure strategy, we can use it to select a mixed strategy
- Take the maximum of the minimum payoffs over all assignments of probabilities
- von Neumann proved you can always find an equilibrium if mixed strategies are permitted

Analysis

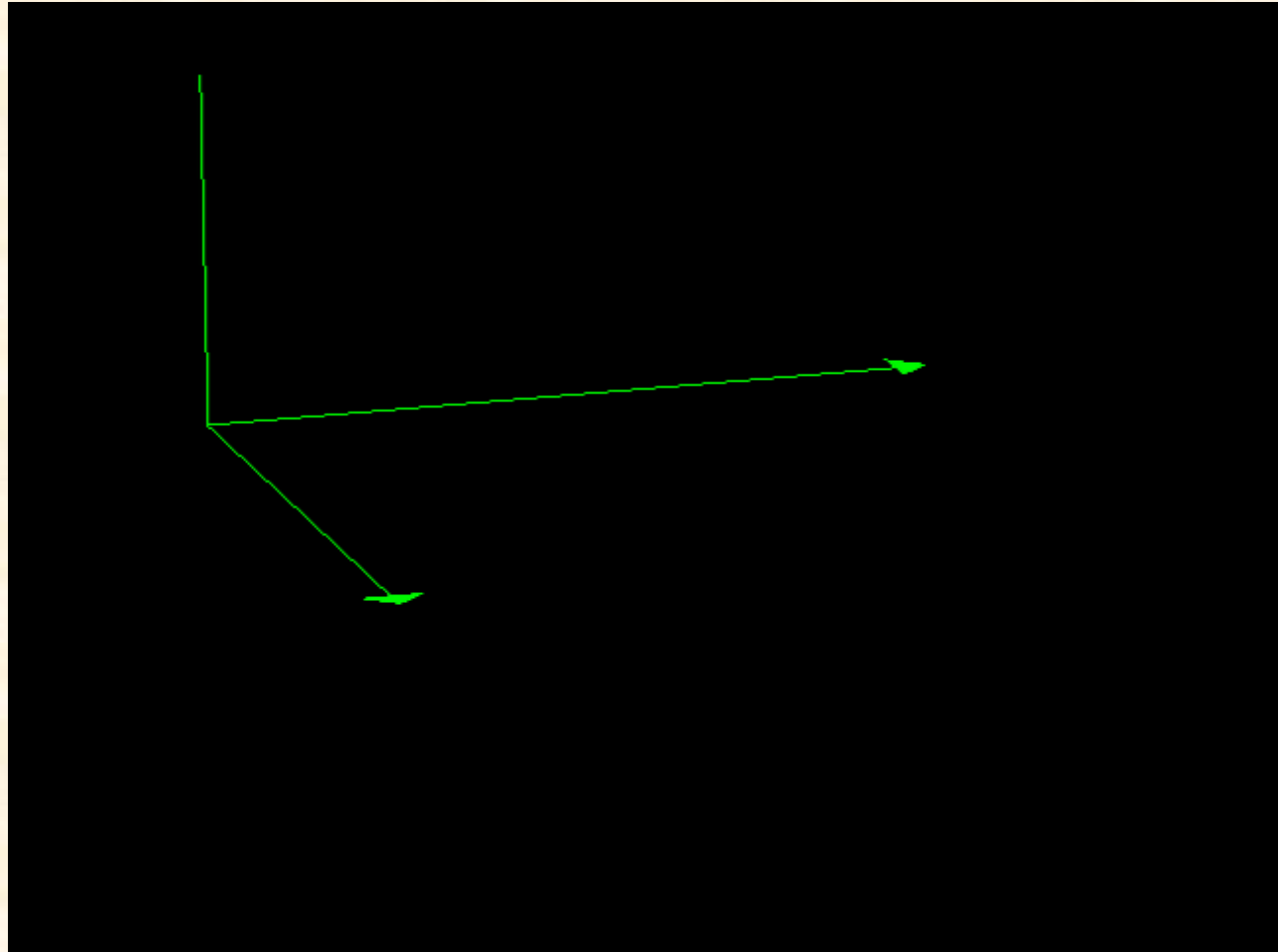
- Let P_A = probability Al picks head
- and P_B = probability Barb picks head
- Al's expected payoff:

$$\begin{aligned} E\{A\} &= P_A P_B - P_A (1 - P_B) - (1 - P_A) P_B \\ &\quad + (1 - P_A) (1 - P_B) \\ &= (2 P_A - 1) (2 P_B - 1) \end{aligned}$$

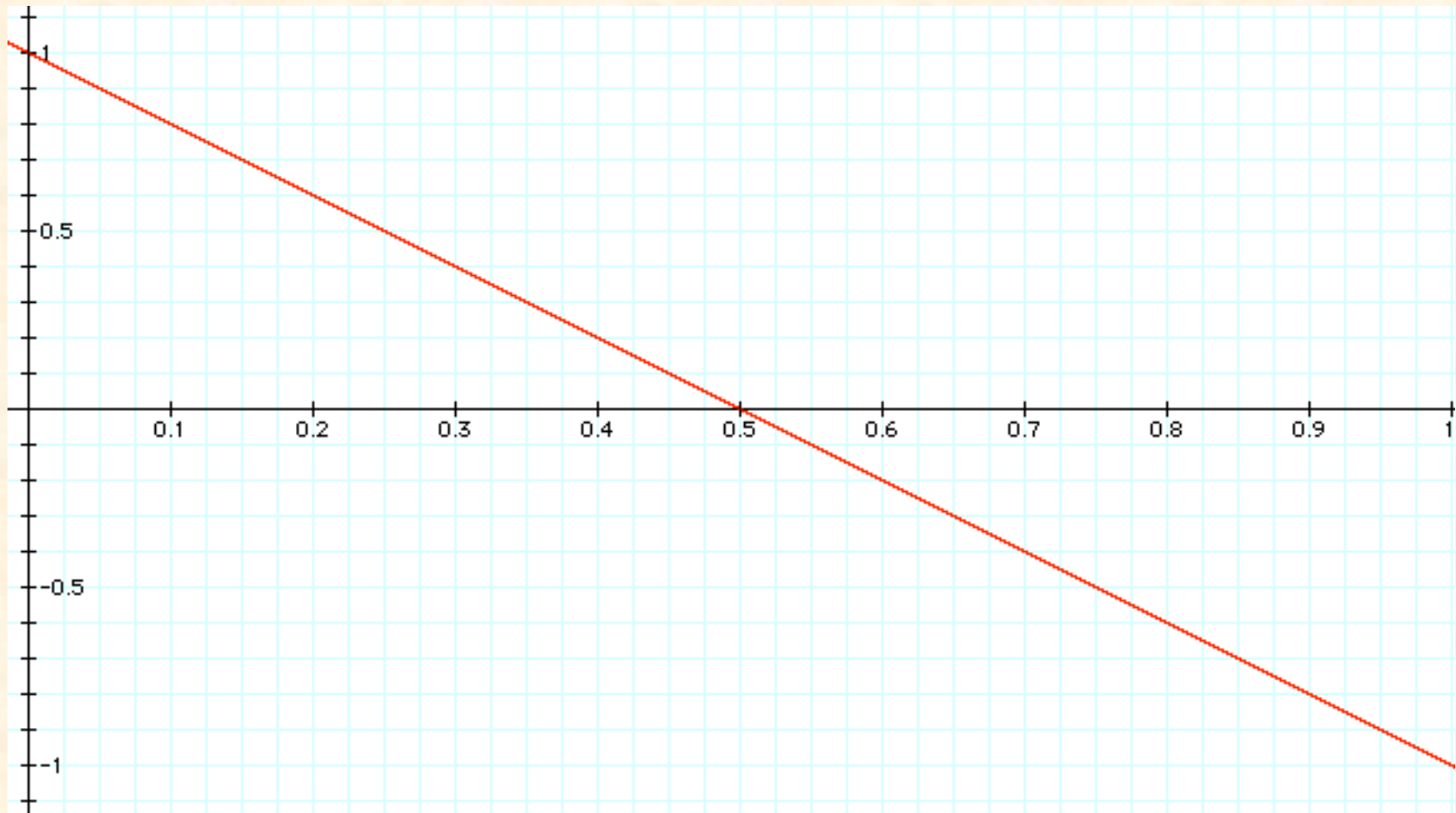
AI's Expected Payoff from Penny Game



How Barb's Behavior Affects Al's Expected Payoff



How Barb's Behavior Affects Al's Expected Payoff



More General Analysis (Differing Payoffs)

- Let A's payoffs be:

$$H = HH, h = HT, t = TH, T = TT$$

- $E\{A\} = P_A P_B H + P_A(1 - P_B)h + (1 - P_A)P_B t + (1 - P_A)(1 - P_B)T$
 $= (H + T - h - t)P_A P_B + (h - T)P_A + (t - T)P_B + T$
- To find saddle point set $\partial E\{A\}/\partial P_A = 0$ and $\partial E\{A\}/\partial P_B = 0$ to get:

$$P_A = \frac{T - t}{H + T - h - t}, \quad P_B = \frac{T - h}{H + T - h - t}$$

Random Rationality

“It seems difficult, at first, to accept the idea that ‘rationality’ — which appears to demand a clear, definite plan, a deterministic resolution — should be achieved by the use of probabilistic devices. Yet precisely such is the case.”

—Morgenstern

Probability in Games of Chance and Strategy

- “In games of chance the task is to determine and then to evaluate probabilities inherent in the game;
- in games of strategy we *introduce* probability in order to obtain the optimal choice of strategy.”

— Morgenstern

Review of von Neumann's Solution

- Every two-person zero-sum game has a maximin solution, provided we allow mixed strategies
- But— it applies only to two-person zero-sum games
- Arguably, few “games” in real life are zero-sum, except literal games (i.e., invented games for amusement)

Nonconstant Sum Games

- There is no agreed upon definition of rationality for nonconstant sum games
- Two common criteria:
 - dominant strategy equilibrium
 - Nash equilibrium

Dominant Strategy Equilibrium

- **Dominant strategy:**
 - consider each of opponents' strategies, and what your best strategy is in each situation
 - if the same strategy is best in all situations, it is the dominant strategy
- **Dominant strategy equilibrium:** occurs if each player has a dominant strategy and plays it

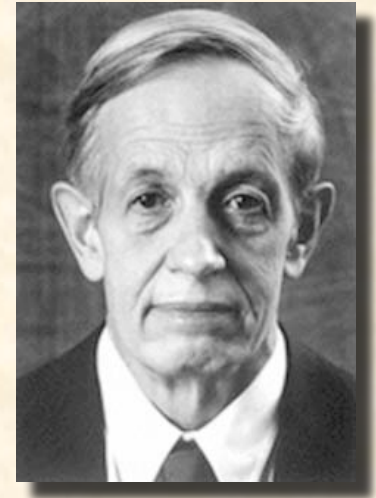
Another Example

Price Competition		Beta		
		$p = 1$	$p = 2$	$p = 3$
Alpha	$p = 1$	0, 0	50, -10	40, -20
	$p = 2$	-10, 50	20, 20	90, 10
	$p = 3$	-20, 40	10, 90	50, 50

There is no dominant strategy



Nash Equilibrium



- Developed by John Nash in 1950
- His 27-page PhD dissertation:
Non-Cooperative Games
- Received Nobel Prize in Economics for it in 1994
- Subject of *A Beautiful Mind*

Definition of Nash Equilibrium

- A set of strategies with the property:
No player can benefit by changing actions while others keep strategies unchanged
- Players are in equilibrium if any change of strategy would lead to lower reward for that player
- For mixed strategies, we consider expected reward

Another Example (Reconsidered)

Price Competition		Beta		
		$p = 1$	$p = 2$	$p = 3$
Alpha	$p = 1$	0, 0	50, -10	40, -20
	$p = 2$	-10, 50	20, 20	90, 10
	$p = 3$	-20, 40	10, 90	50, 50

better for Beta

better for Alpha

Not a Nash equilibrium

The Nash Equilibrium

Price Competition		Beta		
		$p = 1$	$p = 2$	$p = 3$
Alpha	$p = 1$	0, 0	50, -10	40, -20
	$p = 2$	-10, 50	20, 20	90, 10
	$p = 3$	-20, 40	10, 90	50, 50

Nash equilibrium

Extensions of the Concept of a Rational Solution

- Every maximin solution is a dominant strategy equilibrium
- Every dominant strategy equilibrium is a Nash equilibrium

Cooperation Better for Both: A Dilemma

Price Competition		Beta		
		$p = 1$	$p = 2$	$p = 3$
Alpha	$p = 1$	0, 0	50, -10	40, -20
	$p = 2$	-10, 50	20, 20	90, 10
	$p = 3$	-20, 40	10, 90	50, 50

Cooperation

Dilemmas

- Dilemma: “A situation that requires choice between options that are or seem equally unfavorable or mutually exclusive”

– *Am. Her. Dict.*

- In game theory: each player acts rationally, but the result is undesirable (less reward)

The Prisoners' Dilemma

- Devised by Melvin Dresher & Merrill Flood in 1950 at RAND Corporation
- Further developed by mathematician Albert W. Tucker in 1950 presentation to psychologists
- It “has given rise to a vast body of literature in subjects as diverse as philosophy, ethics, biology, sociology, political science, economics, and, of course, game theory.” — S.J. Hagenmayer
- “This example, which can be set out in one page, could be the most influential one page in the social sciences in the latter half of the twentieth century.” — R.A. McCain

Prisoners' Dilemma: The Story

- Two criminals have been caught
- They cannot communicate with each other
- If both confess, they will each get 10 years
- If one confesses and accuses other:
 - confessor goes free
 - accused gets 20 years
- If neither confesses, they will both get 1 year on a lesser charge

Prisoners' Dilemma Payoff Matrix

		Bob	
		cooperate	defect
Ann	cooperate	-1, -1	-20, 0
	defect	0, -20	-10, -10

- defect = confess, cooperate = don't
- payoffs < 0 because punishments (losses)

Ann's "Rational" Analysis (Dominant Strategy)

		Bob	
		cooperate	defect
Ann	cooperate	-1, -1	-20, 0
	defect	0, -20	-10, -10

- if cooperates, may get 20 years
- if defects, may get 10 years
- \therefore , best to defect

Bob's "Rational" Analysis (Dominant Strategy)

		Bob	
		cooperate	defect
Ann	cooperate	-1, -1	-20, 0
	defect	0, -20	-10, -10

- if he cooperates, may get 20 years
- if he defects, may get 10 years
- \therefore , best to defect

Suboptimal Result of “Rational” Analysis

		Bob	
		cooperate	defect
Ann	cooperate	-1, -1	-20, 0
	defect	0, -20	-10, -10

- each acts individually rationally \Rightarrow get 10 years (dominant strategy equilibrium)
- “irrationally” decide to cooperate \Rightarrow only 1 year

Summary

- Individually rational actions lead to a result that all agree is less desirable
- In such a situation you cannot act unilaterally in your own best interest
- Just one example of a (game-theoretic) *dilemma*
- Can there be a situation in which it would make sense to cooperate unilaterally?
 - **Yes**, if the players can expect to interact again in the future

Classification of Dilemmas

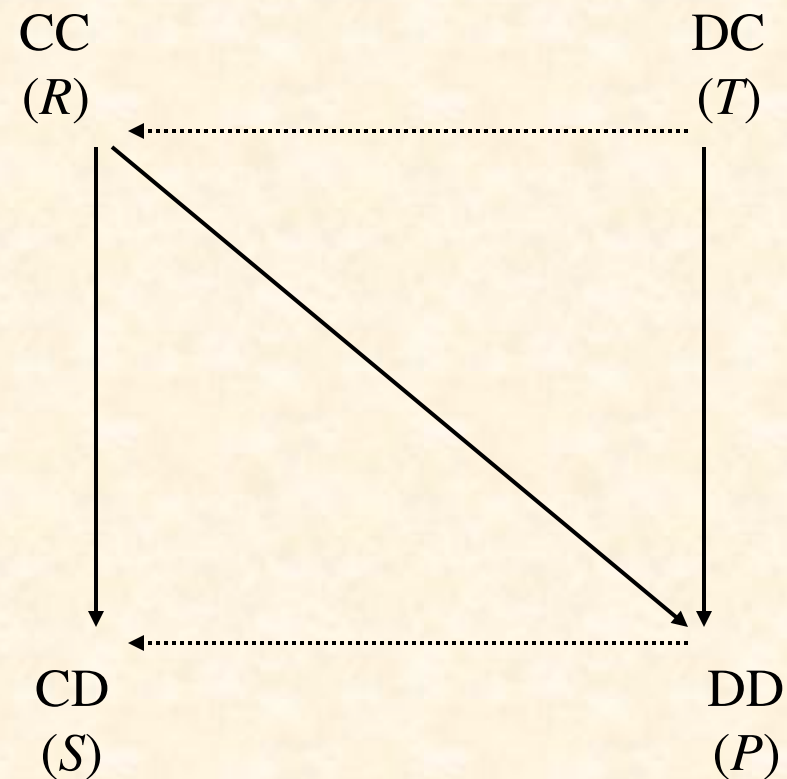
General Payoff Matrix

		Bob	
		cooperate	defect
Ann	cooperate	CC (<i>R</i>) Reward	CD (<i>S</i>) Sucker
	defect	DC (<i>T</i>) Temptation	DD (<i>P</i>) Punishment

General Conditions for a Dilemma

- You always benefit if the other cooperates:
 - $CC > CD$ *and* $DC > DD$
- You sometimes benefit from defecting:
 - $DC > CC$ *or* $DD > CD$
- Mutual coop. is preferable to mut. def.
 - $CC > DD$
- Consider relative size of CC , CD , DC , DD
 - think of as permutations of R, S, T, P
 - only three result in dilemmas

Three Possible Orders



The three dilemmas: *TRSP*, *RTPS*, *TRPS*

The Three Dilemmas

- Chicken (*TRSP*)
 - $DC > CC > CD > DD$
 - characterized by mutual defection being worst
 - two Nash equilibria (DC, CD)
- Stag Hunt (*RTPS*)
 - $CC > DC > DD > CD$
 - better to cooperate with cooperator
 - Nash equilibrium is CC
- Prisoners' Dilemma (*TRPS*)
 - $DC > CC > DD > CD$
 - better to defect on cooperator
 - Nash equilibrium is DD

The Iterated Prisoners' Dilemma

and Robert Axelrod's Experiments

Assumptions

- No mechanism for enforceable threats or commitments
- No way to foresee a player's move
- No way to eliminate other player or avoid interaction
- No way to change other player's payoffs
- Communication only through direct interaction

Axelrod's Experiments

- Intuitively, expectation of future encounters may affect rationality of defection
- Various programs compete for 200 rounds
 - encounters each other and self
- Each program can remember:
 - its own past actions
 - its competitors' past actions
- 14 programs submitted for first experiment

IPD Payoff Matrix

		B	
		cooperate	defect
A	cooperate	3, 3	0, 5
	defect	5, 0	1, 1

N.B. Unless $DC + CD < 2 CC$ (i.e. $T + S < 2 R$),
can win by alternating defection/cooperation

Indefinite Number of Future Encounters

- Cooperation depends on expectation of **indefinite** number of future encounters
- Suppose a known finite number of encounters:
 - No reason to C on last encounter
 - Since expect D on last, no reason to C on next to last
 - And so forth: there is no reason to C at all

Analysis of Some Simple Strategies

- Three simple strategies:
 - **ALL-D**: always defect
 - **ALL-C**: always cooperate
 - **RAND**: randomly cooperate/defect
- Effectiveness depends on environment
 - **ALL-D** optimizes local (individual) fitness
 - **ALL-C** optimizes global (population) fitness
 - **RAND** compromises

Expected Scores

↓ playing ⇒	ALL-C	RAND	ALL-D	Average
ALL-C	3.0	1.5	0.0	1.5
RAND	4.0	2.25	0.5	2.25
ALL-D	5.0	3.0	1.0	3.0

Result of Axelrod's Experiments

- Winner is Rapoport's **TFT** (Tit-for-Tat)
 - cooperate on first encounter
 - reply in kind on succeeding encounters
- Second experiment:
 - 62 programs
 - all know **TFT** was previous winner
 - **TFT** wins again

Demonstration of Iterated Prisoners' Dilemma

Run NetLogo demonstration
PD N-Person Iterated.nlogo

Characteristics of Successful Strategies

- *Don't be envious*
 - at best TFT ties other strategies
- *Be nice*
 - i.e. don't be first to defect
- *Reciprocate*
 - reward cooperation, punish defection
- *Don't be too clever*
 - sophisticated strategies may be unpredictable & look random; be clear

Tit-for-Two-Tats

- More forgiving than **TFT**
- Wait for two successive defections before punishing
- Beats **TFT** in a noisy environment
- E.g., an unintentional defection will lead **TFTs** into endless cycle of retaliation
- May be exploited by feigning accidental defection

Effects of Many Kinds of Noise Have Been Studied

- Misimplementation noise
- Misperception noise
 - noisy channels
- Stochastic effects on payoffs
- General conclusions:
 - sufficiently little noise \Rightarrow generosity is best
 - greater noise \Rightarrow generosity avoids unnecessary conflict but invites exploitation

More Characteristics of Successful Strategies

- Should be a generalist (robust)
 - i.e. do sufficiently well in wide variety of environments
- Should do well with its own kind
 - since successful strategies will propagate
- Should be cognitively simple
- Should be evolutionary stable strategy
 - i.e. resistant to invasion by other strategies

Kant's Categorical Imperative

“Act on maxims that can at the same time have for their object themselves as universal laws of nature.”

Ecological & Spatial Models

Ecological Model

- What if more successful strategies spread in population at expense of less successful?
- Models success of programs as fraction of total population
- Fraction of strategy = probability random program obeys this strategy

Variables

- $P_i(t)$ = probability = proportional population of strategy i at time t
- $S_i(t)$ = score achieved by strategy i
- $R_{ij}(t)$ = relative score achieved by strategy i playing against strategy j over many rounds
 - fixed (not time-varying) for now

Computing Score of a Strategy

- Let n = number of strategies in ecosystem
- Compute score achieved by strategy i :

$$S_i(t) = \sum_{k=1}^n R_{ik}(t)P_k(t)$$

$$\mathbf{S}(t) = \mathbf{R}(t)\mathbf{P}(t)$$

Updating Proportional Population

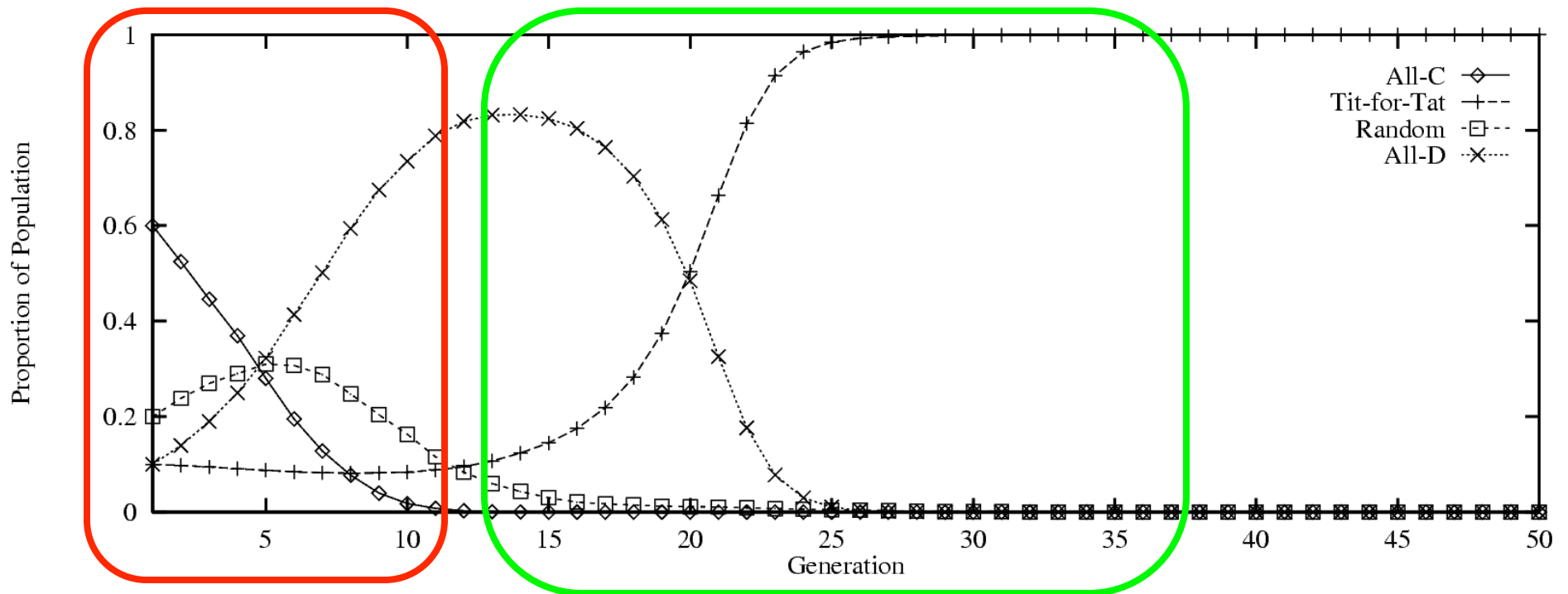
$$P_i(t + 1) = \frac{P_i(t)S_i(t)}{\sum_{j=1}^n P_j(t)S_j(t)}$$

Some Simulations

- Usual Axelrod payoff matrix
- 200 rounds per step

Demonstration Simulation

- 60% ALL-C
- 20% RAND
- 10% ALL-D, TFT



NetLogo Demonstration of Ecological IPD

Run EIPD.nlogo

Collectively Stable Strategy

- Let w = probability of future interactions
- Suppose cooperation based on reciprocity has been established
- Then no one can do better than **TFT** provided:

$$w \geq \max\left(\frac{T - R}{R - S}, \frac{T - R}{T - P}\right)$$

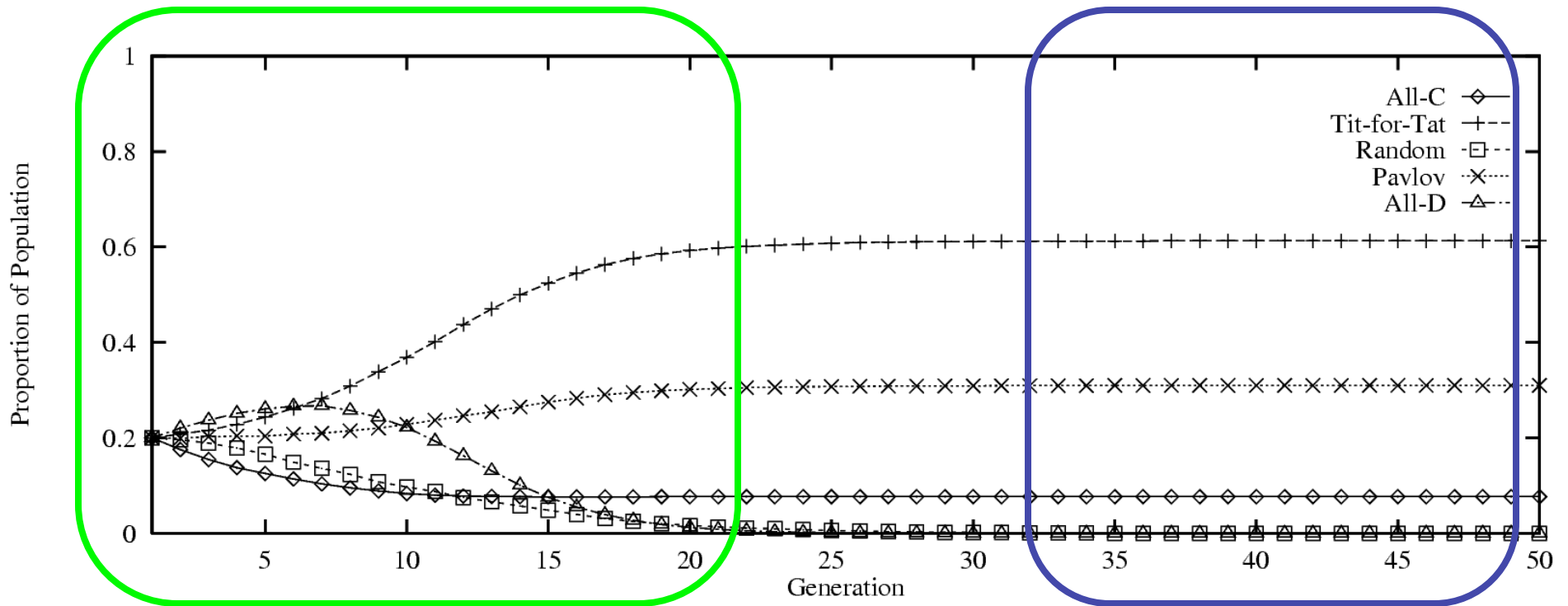
- The **TFT** users are in a Nash equilibrium

“Win-Stay, Lose-Shift” Strategy

- Win-stay, lose-shift strategy:
 - begin cooperating
 - if other cooperates, continue current behavior
 - if other defects, switch to opposite behavior
- Called **PAV** (because suggests Pavlovian learning)

Simulation without Noise

- 20% each
- no noise

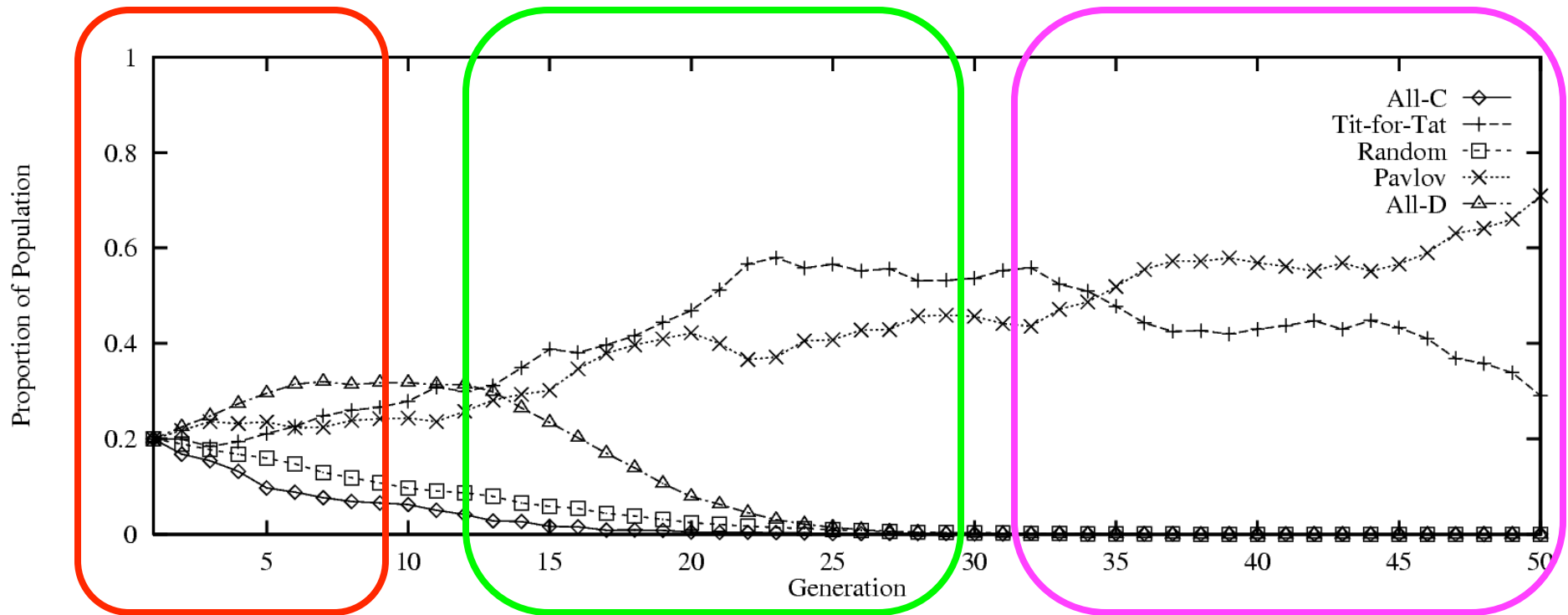


Effects of Noise

- Consider effects of noise or other sources of error in response
- **TFT:**
 - cycle of alternating defections (CD, DC)
 - broken only by another error
- **PAV:**
 - eventually self-corrects (CD, DC, DD, CC)
 - can exploit **ALL-C** in noisy environment
- Noise added into computation of $R_{ij}(t)$

Simulation with Noise

- 20% each
- 0.5% noise



Spatial Effects

- Previous simulation assumes that each agent is equally likely to interact with each other
- So strategy interactions are proportional to fractions in population
- More realistically, interactions with “neighbors” are more likely
 - “Neighbor” can be defined in many ways
- Neighbors are more likely to use the same strategy

Spatial Simulation

- Toroidal grid
- Agent interacts only with eight neighbors
- Agent adopts strategy of most successful neighbor
- Ties favor current strategy

NetLogo Simulation of Spatial IPD

Run SIPD.nlogo

Typical Simulation ($t = 1$)



Colors:

ALL-C

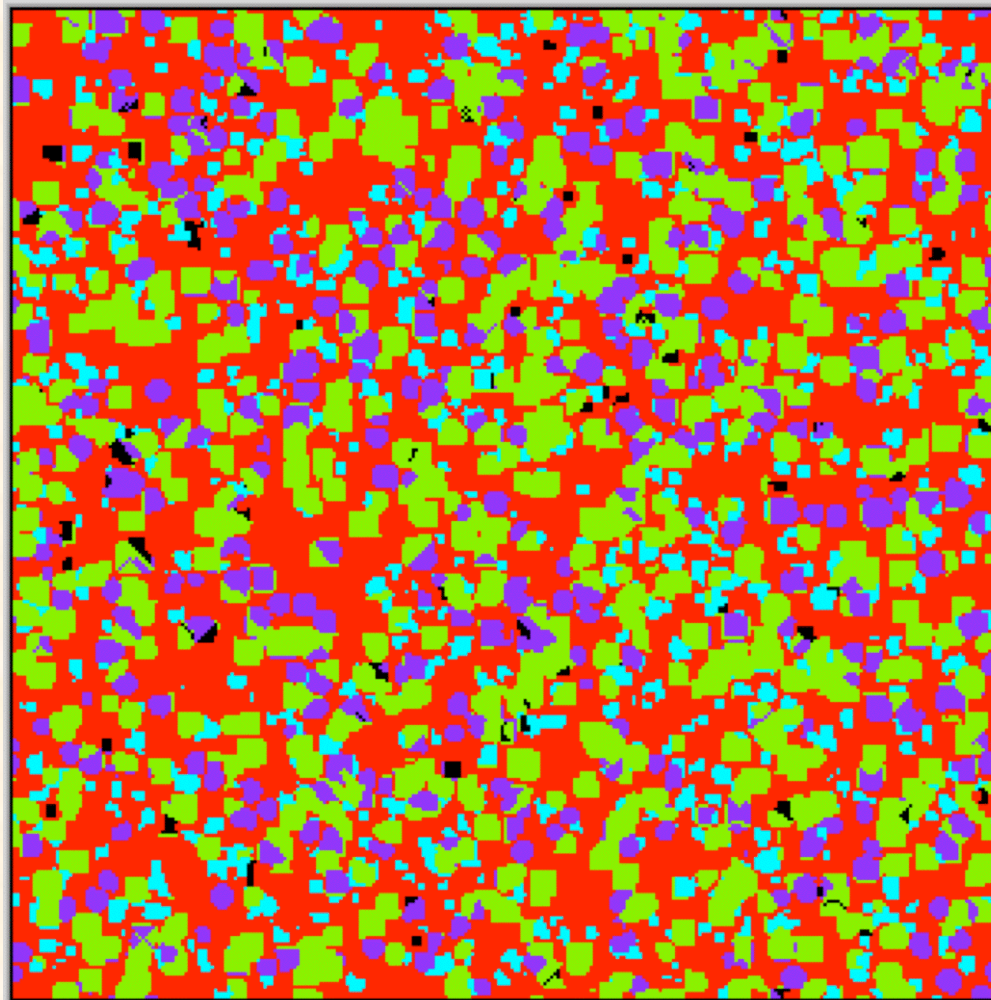
TFT

RAND

PAV

ALL-D

Typical Simulation ($t = 5$)



Colors:

ALL-C

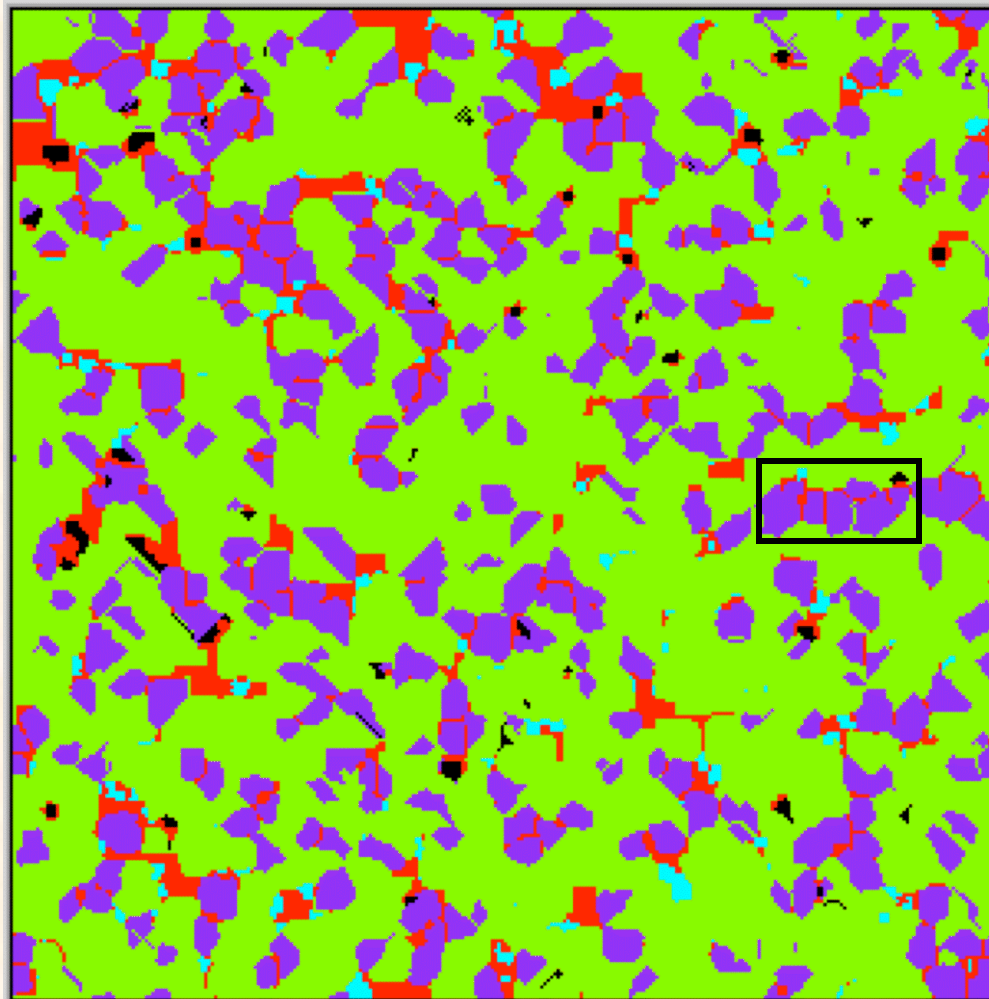
TFT

RAND

PAV

ALL-D

Typical Simulation ($t = 10$)



Colors:

ALL-C

TFT

RAND

PAV

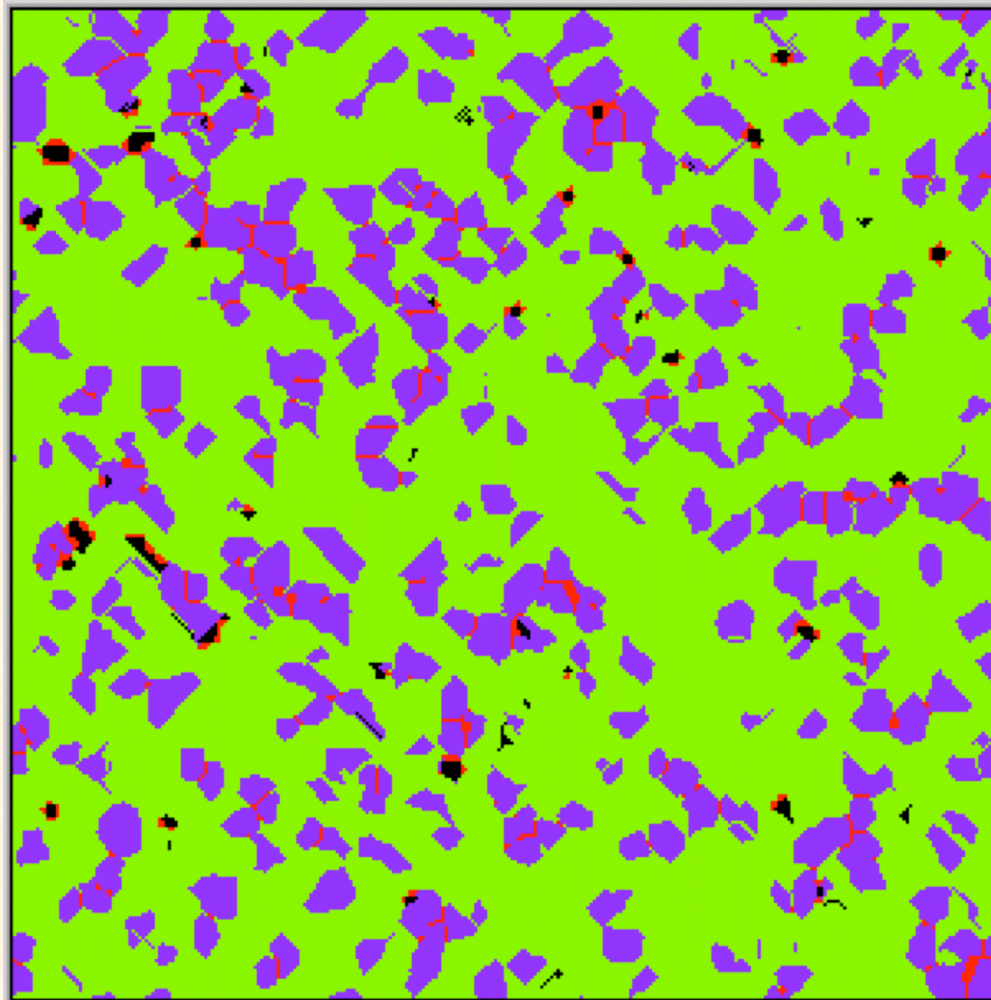
ALL-D

Typical Simulation ($t = 10$)

Zooming In



Typical Simulation ($t = 20$)



Colors:

ALL-C

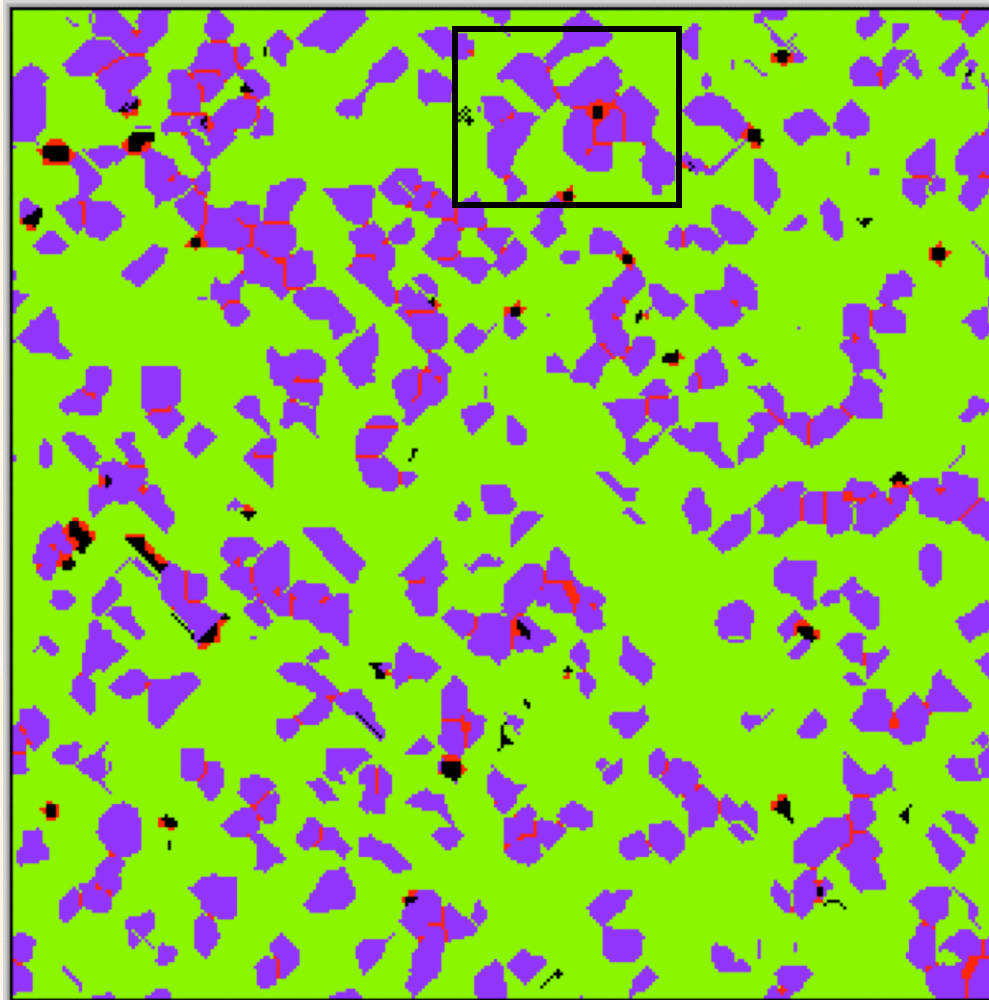
TFT

RAND

PAV

ALL-D

Typical Simulation ($t = 50$)



Colors:

ALL-C

TFT

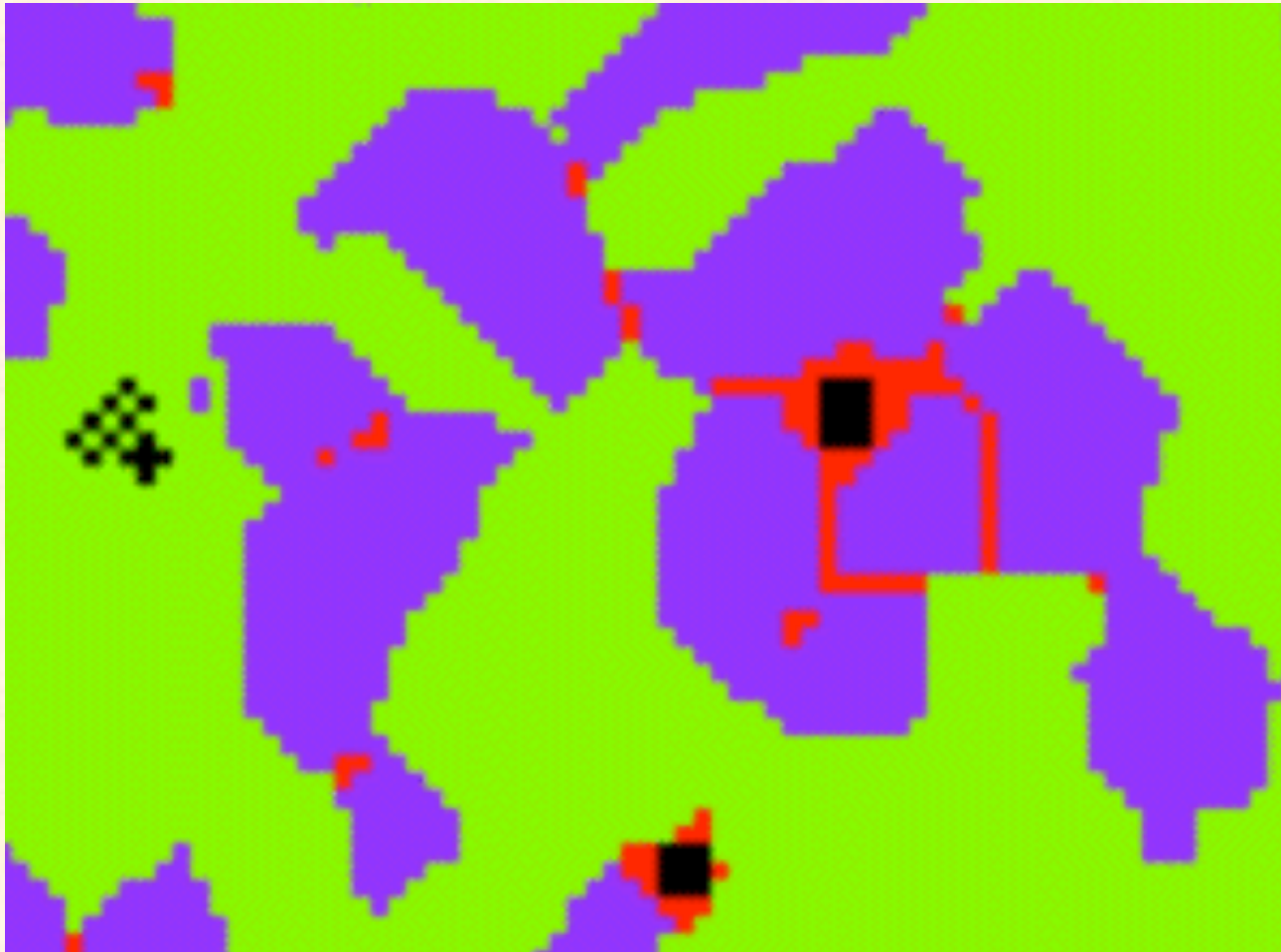
RAND

PAV

ALL-D






Typical Simulation ($t = 50$)

Zoom In



SIPD Without Noise

Legend

-  — All-C
-  — Tit-for-Tat
-  — Random
-  — Pavlov
-  — All-D

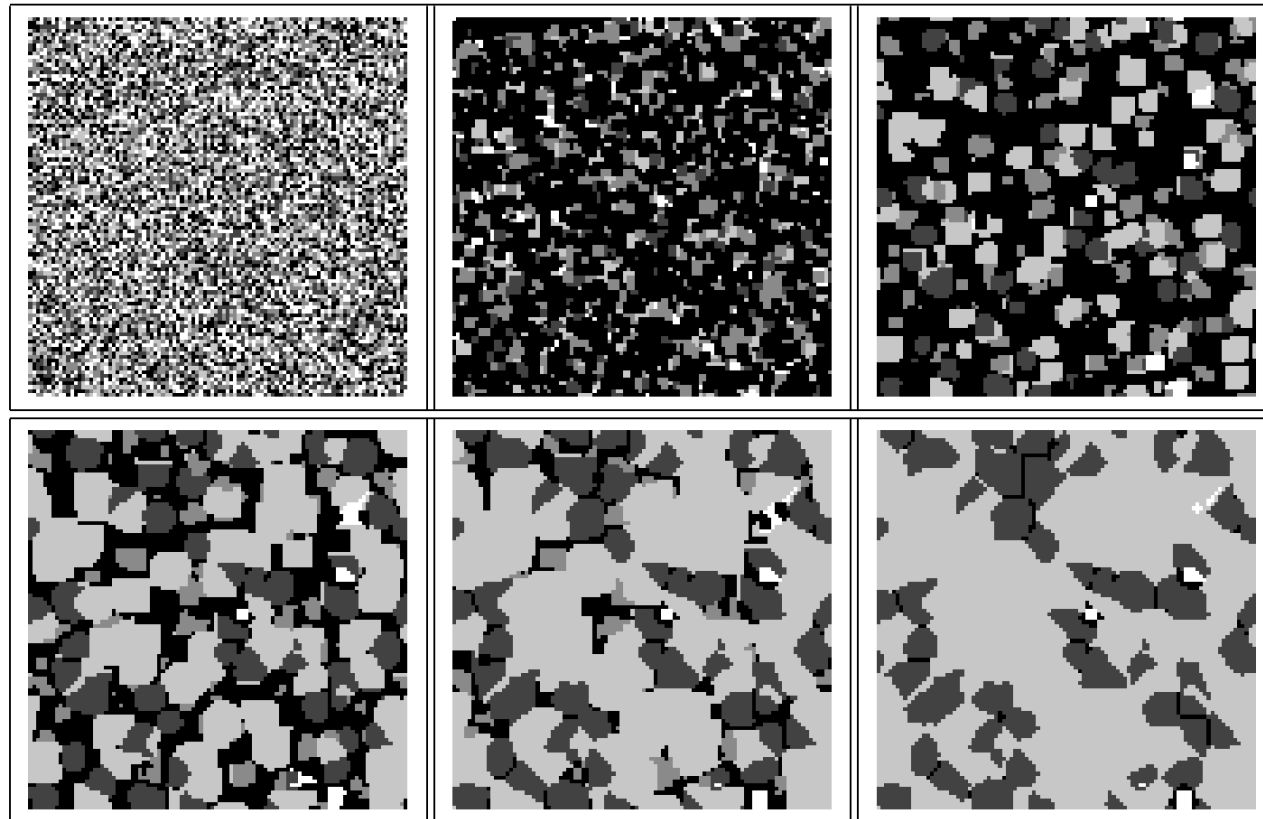


Figure 17.4 Competition in the spatial iterated Prisoner's Dilemma without noise

Figure from *The Computational Beauty of Nature: Computer Explorations of Fractals, Chaos, Complex Systems, and Adaptation*. Copyright © 1998–2000 by Gary William Flake. All rights reserved. Permission granted for educational, scholarly, and personal use provided that this notice remains intact and unaltered. No part of this work may be reproduced for commercial purposes without prior written permission from the MIT Press.

Conclusions: Spatial IPD

- Small clusters of cooperators can exist in hostile environment
- Parasitic agents can exist only in limited numbers
- Stability of cooperation depends on expectation of future interaction
- Adaptive cooperation/defection beats unilateral cooperation or defection

Additional Bibliography

1. von Neumann, J., & Morgenstern, O. *Theory of Games and Economic Behavior*, Princeton, 1944.
2. Morgenstern, O. "Game Theory," in *Dictionary of the History of Ideas*, Charles Scribners, 1973, vol. 2, pp. 263-75.
3. Axelrod, R. *The Evolution of Cooperation*. Basic Books, 1984.
4. Axelrod, R., & Dion, D. "The Further Evolution of Cooperation," *Science* **242** (1988): 1385-90.
5. Poundstone, W. *Prisoner's Dilemma*. Doubleday, 1992.