

# Reading

- CS 420/594: Flake, ch. 18 (Natural & Artificial Computation)
- CS 594: Bar-Yam, ch. 2 (Neural Networks I), sections 2.1-2.2 (pp. 295-322)

# 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  $\square$  generosity is best
  - greater noise  $\square$  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  $i$ :

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

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

# Updating Proportional Population

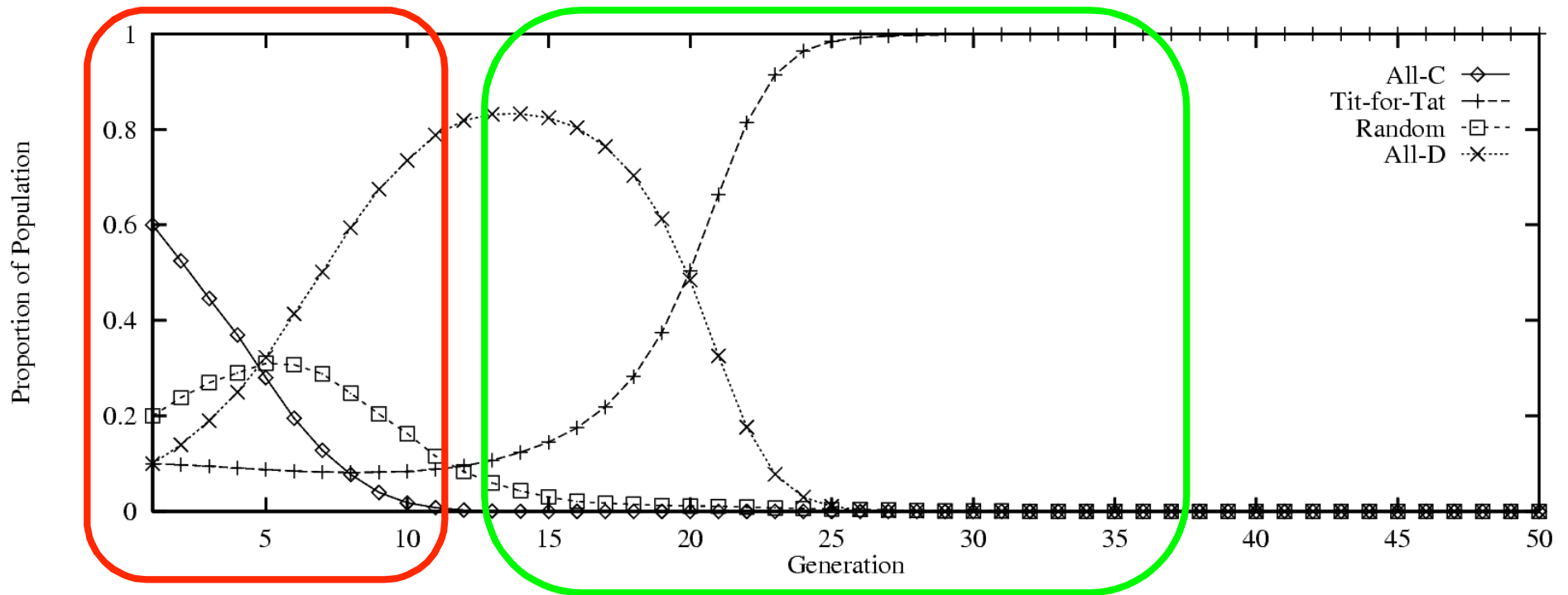
$$P_i(t + 1) = \frac{P_i(t)S_i(t)}{\sum_{j=1}^n P_j(t)S_j(t)}$$

# Some Simulations

- Usual Axelrod payoff matrix
- 200 rounds per step

# Demonstration Simulation

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



# Collectively Stable Strategy

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

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

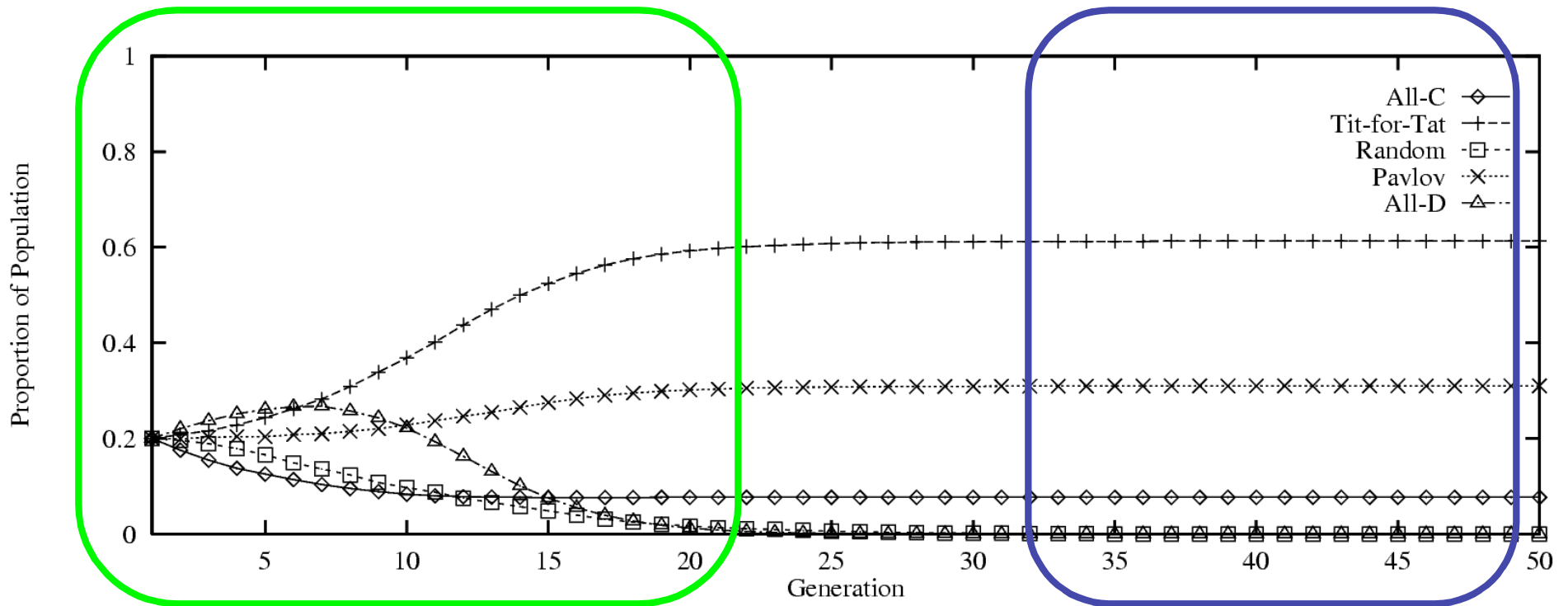
- The **TFT** users are in a Nash equilibrium

# “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



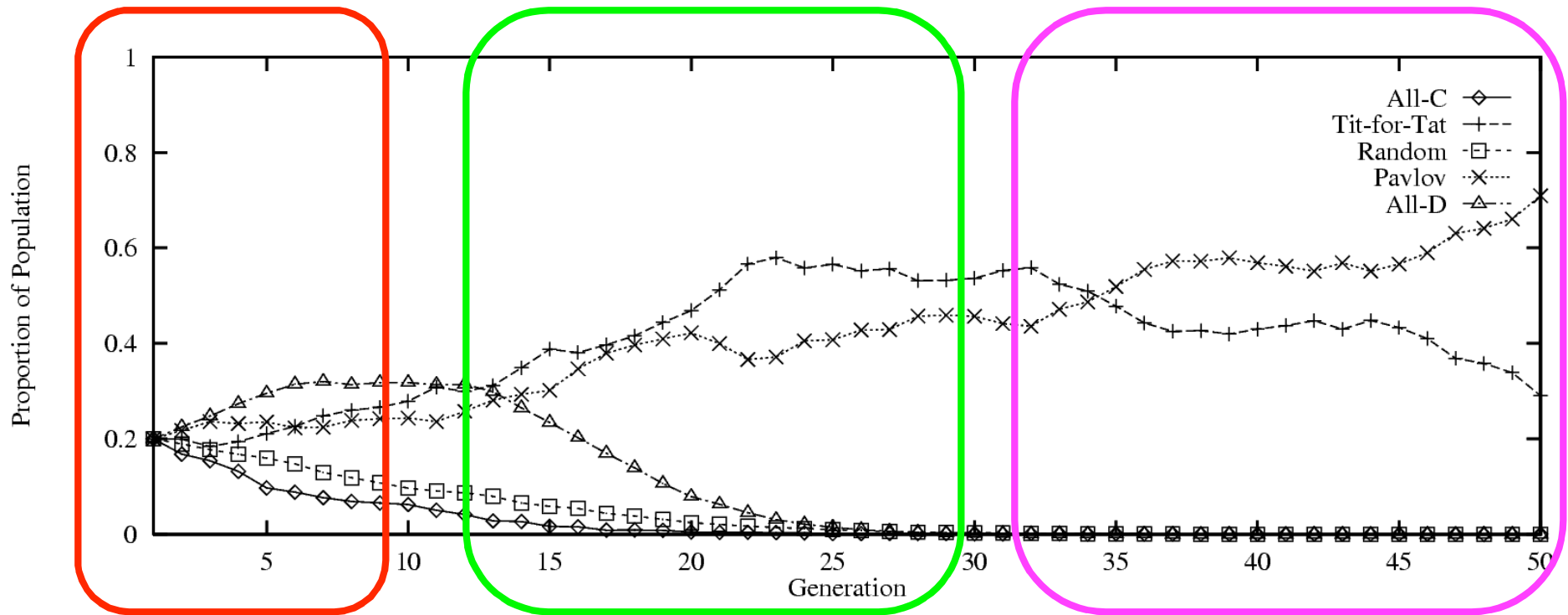
# Effects of Noise

- 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

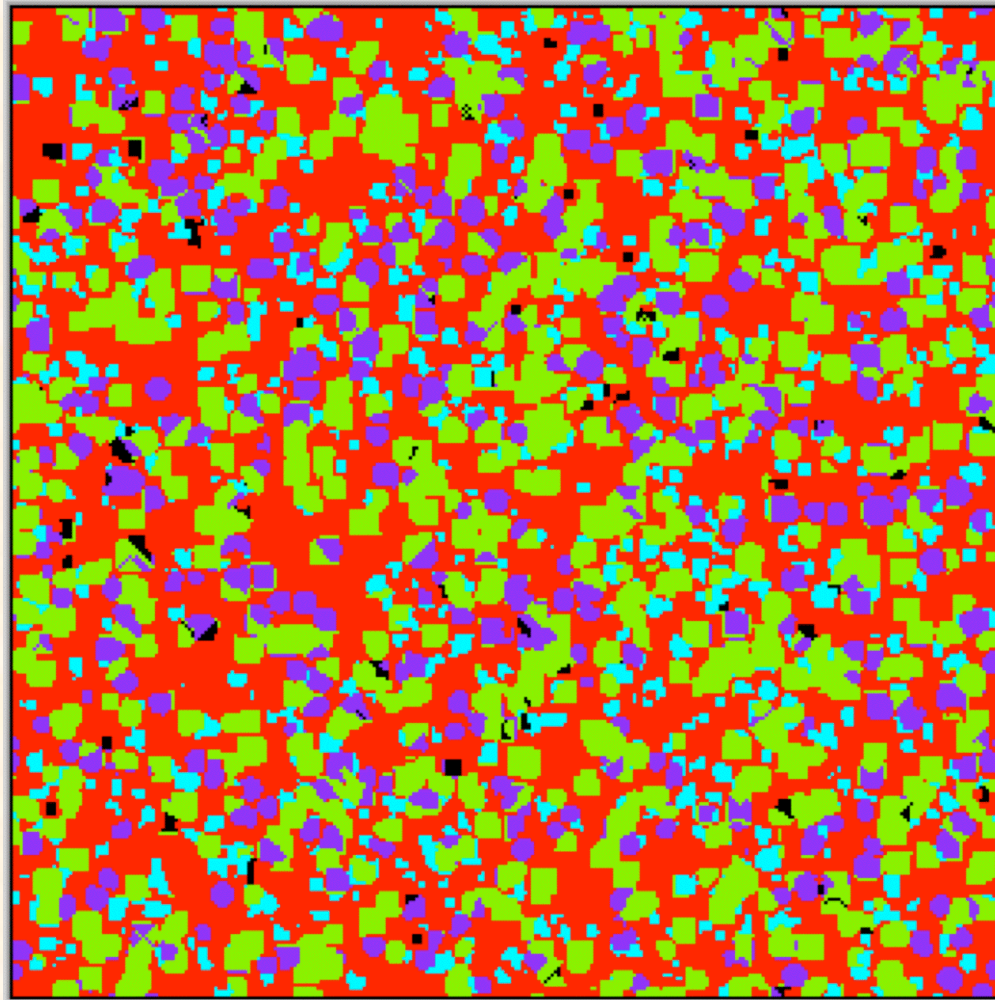
# Spatial Simulation

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

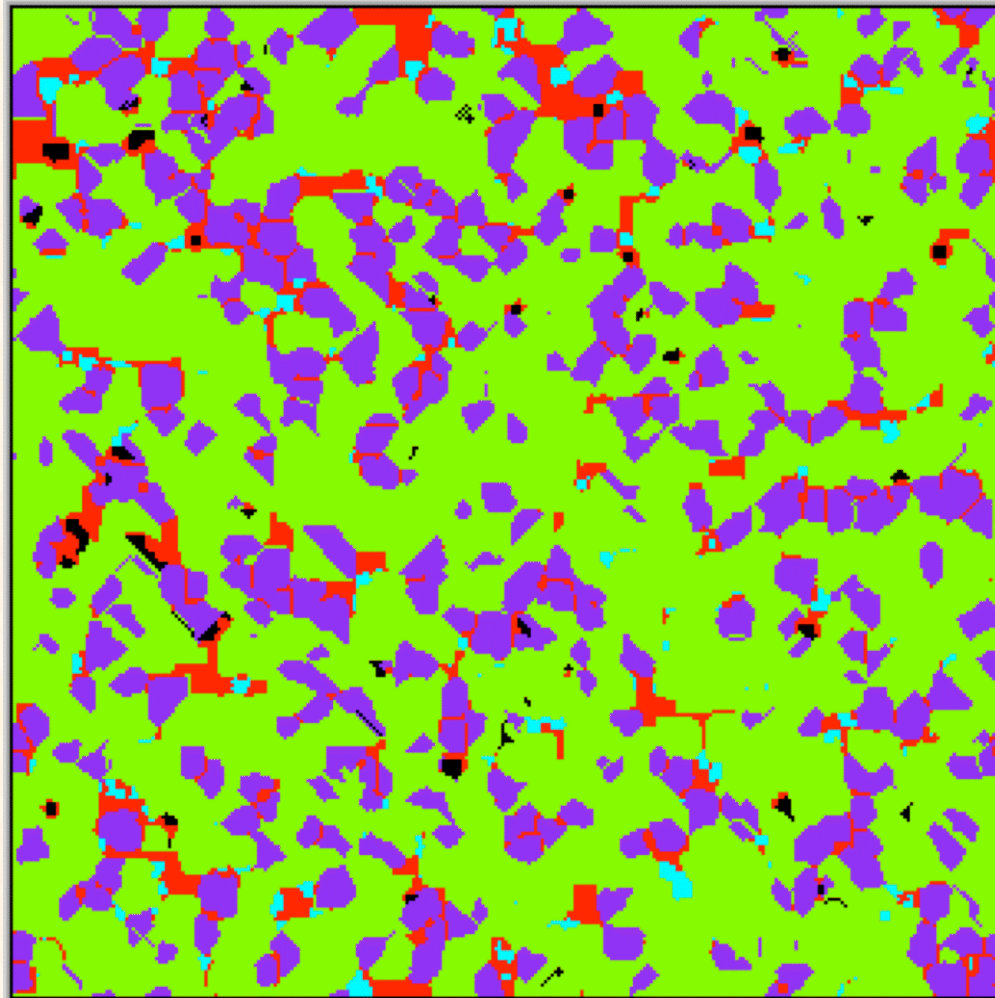
# Typical Simulation ( $t = 1$ )



# Typical Simulation ( $t = 5$ )

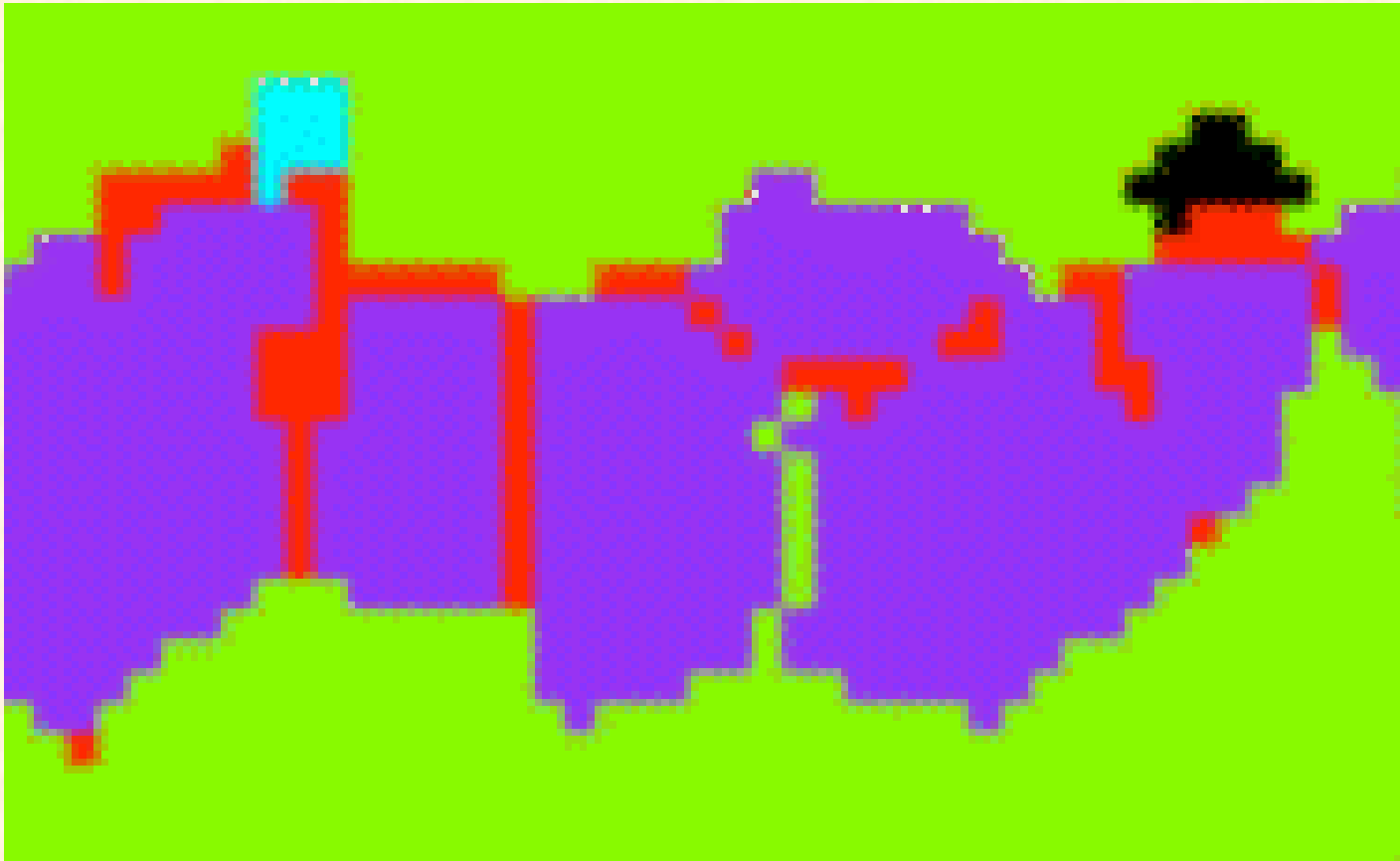


# Typical Simulation ( $t = 10$ )

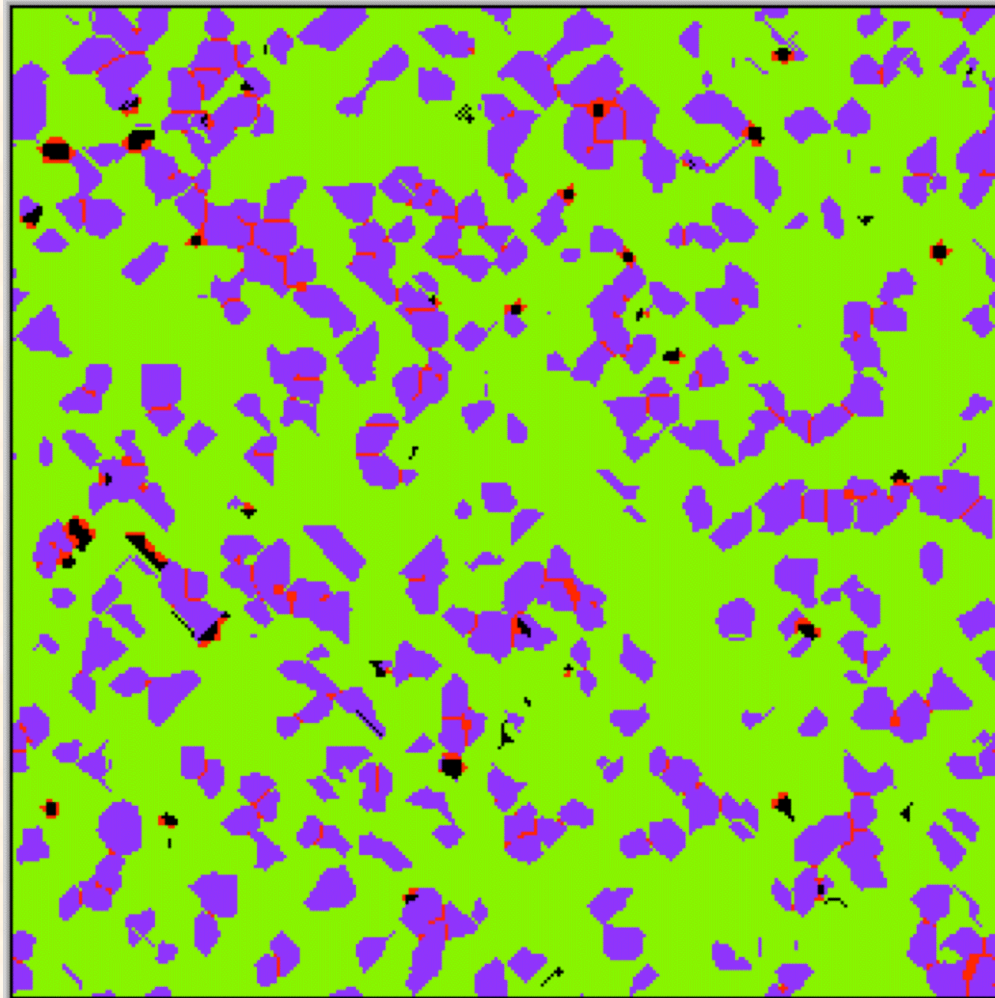


# Typical Simulation ( $t = 10$ )

## Zooming In

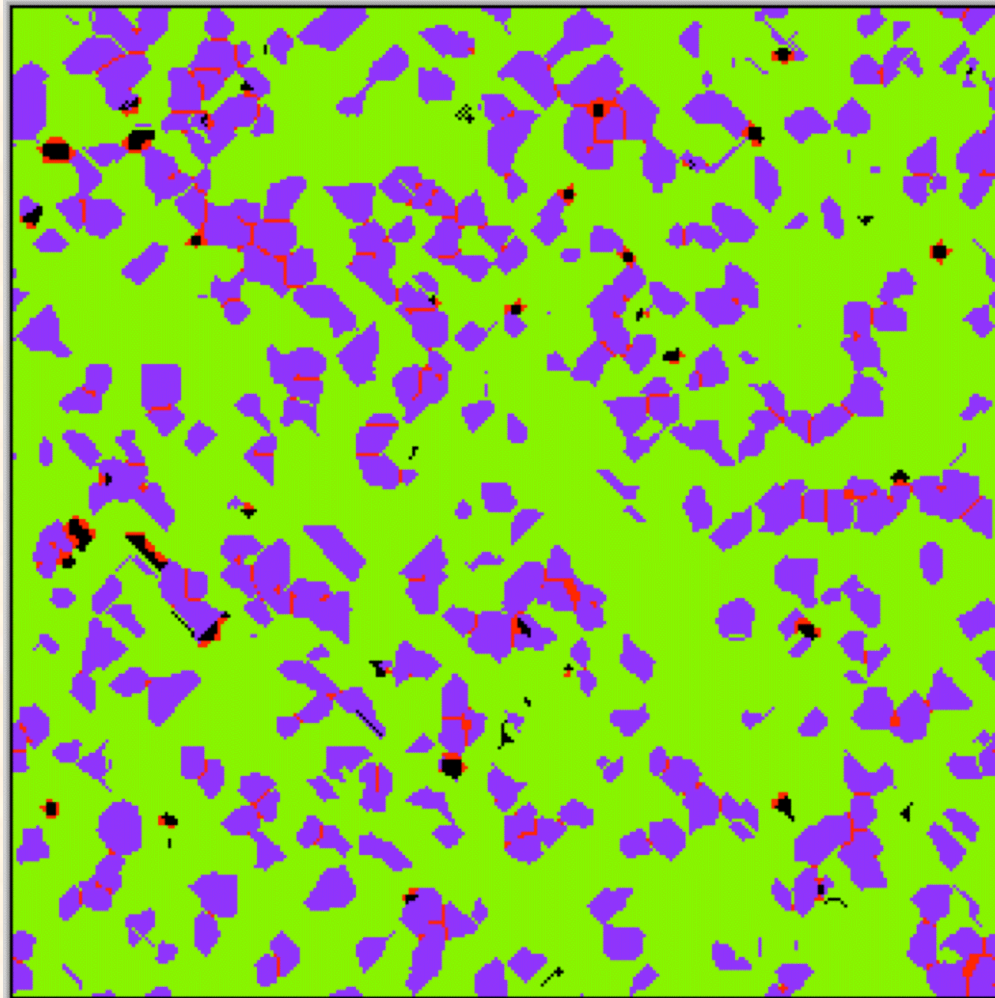


# Typical Simulation ( $t = 20$ )



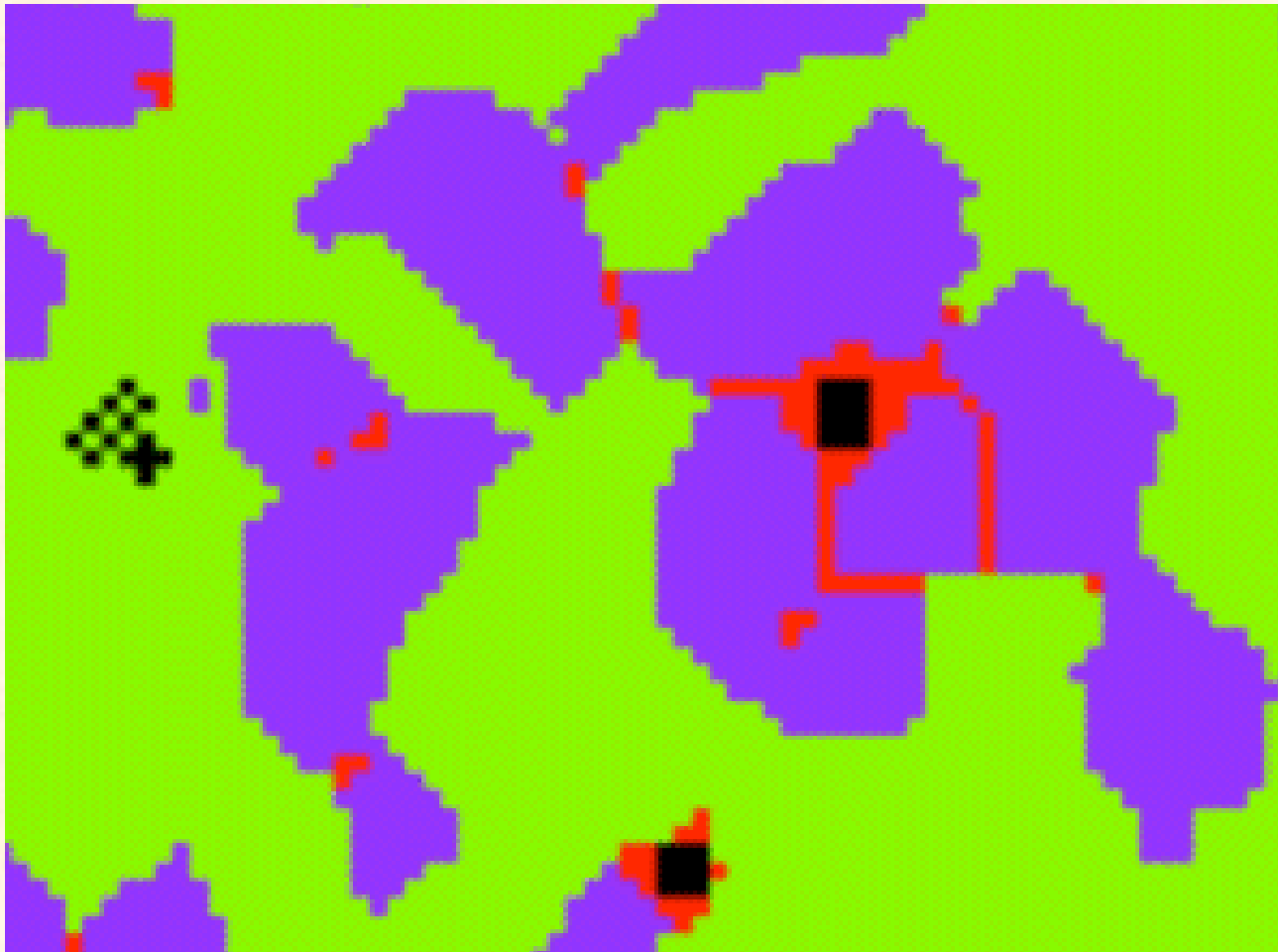


# Typical Simulation ( $t = 50$ )



# Typical Simulation ( $t = 50$ )

## Zoom In








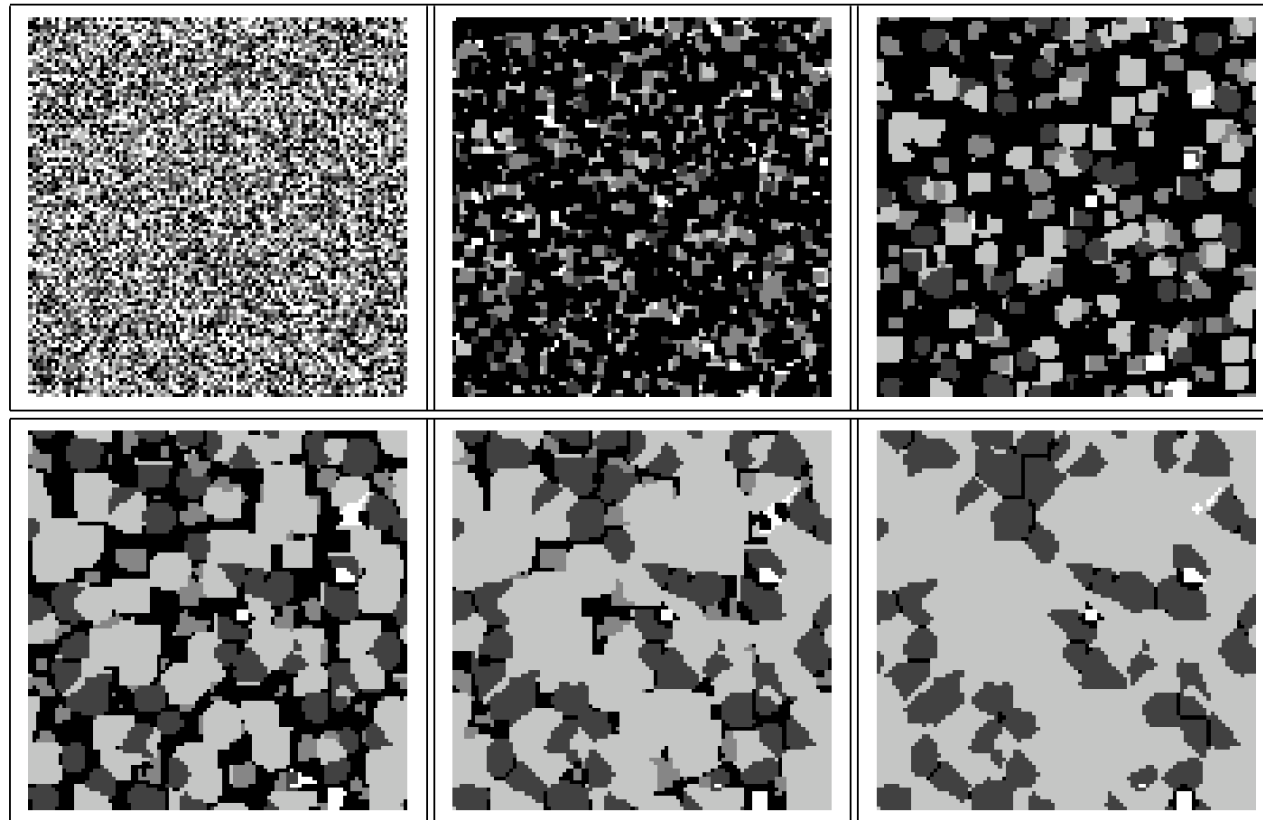
# Simulation of Spatial Iterated Prisoners Dilemma

[Run sipd simulator](#)

# 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.