







B.
Pattern Formation

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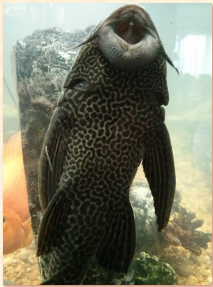
**Differentiation
& Pattern Formation**



- A central problem in development: How do cells differentiate to fulfill different purposes?
- How do complex systems generate spatial & temporal structure?
- CAs are natural models of intercellular communication

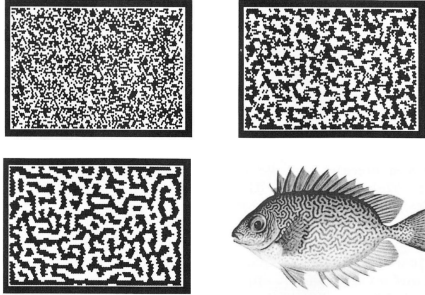
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Plecostomus



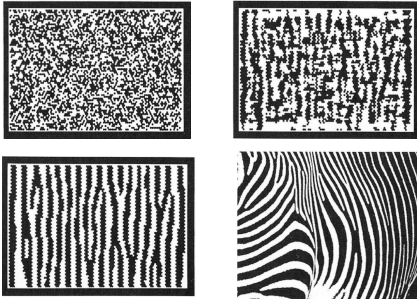
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Vermiculated Rabbit Fish



2015/1/28 figs. from Camazine & al.: *Self-Org. Biol. Sys.* 4

Zebra



2015/1/28 figs. from Camazine & al.: *Self-Org. Biol. Sys.* 5

Activation & Inhibition in Pattern Formation

- Color patterns typically have a characteristic length scale
- Independent of cell size and animal size
- Achieved by:
 - short-range activation \Rightarrow local uniformity
 - long-range inhibition \Rightarrow separation

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Interaction Parameters

- R_1 and R_2 are the interaction ranges
- J_1 and J_2 are the interaction strengths

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CA Activation/Inhibition Model

- Let states $s_i \in \{-1, +1\}$
- and h be a bias parameter
- and r_{ij} be the distance between cells i and j
- Then the state update rule is:

$$s_i(t+1) = \text{sign} \left[h + J_1 \sum_{r_{ij} < R_1} s_j(t) + J_2 \sum_{R_1 \leq r_{ij} < R_2} s_j(t) \right]$$

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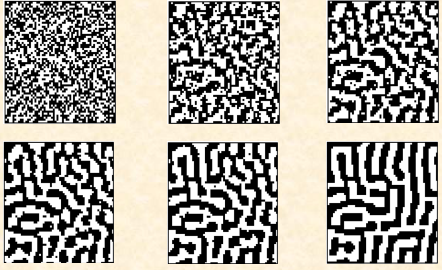
Demonstration of NetLogo Program for Activation/Inhibition Pattern Formation

[RunAICA.nlogo](#)

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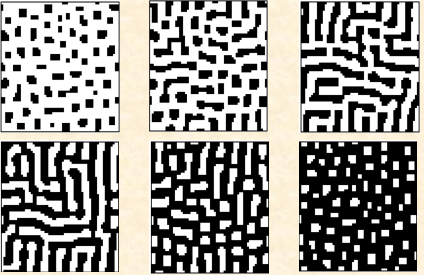
Example
 $(R_1=1, R_2=6, J_1=1, J_2=-0.1, h=0)$



2015/1/28 figs. from Bar-Yam 10






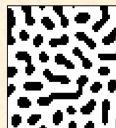
Effect of Bias
 $(h = -6, -3, -1; 1, 3, 6)$



2015/1/28 figs. from Bar-Yam 11



Effect of Interaction Ranges

$R_2 = 6$ $R_1 = 1$ $h = 0$			$R_2 = 8$ $R_1 = 1$ $h = 0$
$R_2 = 6$ $R_1 = 1.5$ $h = 0$			$R_2 = 6$ $R_1 = 1.5$ $h = -3$

2015/1/28 figs. from Bar-Yam 12



Differential Interaction Ranges

- How can a system using strictly local interactions discriminate between states at long and short range?
- E.g. cells in developing organism
- Can use two different *morphogens* diffusing at two different rates
 - activator diffuses slowly (short range)
 - inhibitor diffuses rapidly (long range)

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Digression on Diffusion

- Simple 2-D diffusion equation:

$$\dot{A}(x, y) = D\nabla^2 A(x, y)$$
- Recall the 2-D Laplacian:

$$\nabla^2 A(x, y) = \frac{\partial^2 A(x, y)}{\partial x^2} + \frac{\partial^2 A(x, y)}{\partial y^2}$$
- The Laplacian (like 2nd derivative) is:
 - positive in a local minimum
 - negative in a local maximum

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Reaction-Diffusion System

diffusion

$$\frac{\partial A}{\partial t} = D_A \nabla^2 A + f_A(A, I)$$

$$\frac{\partial I}{\partial t} = D_I \nabla^2 I + f_I(A, I)$$

reaction

$$\frac{\partial}{\partial t} \begin{pmatrix} A \\ I \end{pmatrix} = \begin{pmatrix} D_A & 0 \\ 0 & D_I \end{pmatrix} \begin{pmatrix} \nabla^2 A \\ \nabla^2 I \end{pmatrix} + \begin{pmatrix} f_A(A, I) \\ f_I(A, I) \end{pmatrix}$$

$$\dot{\mathbf{c}} = \mathbf{D}\nabla^2 \mathbf{c} + \mathbf{f}(\mathbf{c}), \text{ where } \mathbf{c} = \begin{pmatrix} A \\ I \end{pmatrix}$$

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General Reaction-Diffusion System

$$\frac{\partial c_i}{\partial t} = \sum_{\alpha} k_{\alpha} v_{i\alpha} \left(\prod_{k=1}^n c_k^{m_{k\alpha}} \right) - \nabla \cdot \mathbf{j}_i$$

where $\mathbf{j}_i = \bar{\mu}_i c_i - \mathbf{div} \mathbf{D}_i c_i$ (flux)

where k_{α} = rate constant for reaction α
 and $v_{i\alpha}$ = stoichiometric coefficient
 and $m_{k\alpha}$ = a non-negative integer
 and $\bar{\mu}_i$ = drift vector
 and \mathbf{D}_i = diffusivity matrix

where $\mathbf{div} \mathbf{D} c = \sum_j \mathbf{e}_j \sum_k D_{jk} \frac{\partial c}{\partial x_k}$

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Framework for Complexity

- change = source terms + transport terms
- source terms = local coupling
= interactions local to a small region
- transport terms = spatial coupling
= interactions with contiguous regions
= advection + diffusion
 - advection: non-dissipative, time-reversible
 - diffusion: dissipative, irreversible

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NetLogo Simulation of Reaction-Diffusion System

1. Diffuse activator in X and Y directions
2. Diffuse inhibitor in X and Y directions
3. Each patch performs:
 - stimulation = bias + activator - inhibitor + noise
 - if stimulation > 0 then
 - set activator and inhibitor to 100
 - else
 - set activator and inhibitor to 0

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**Demonstration of NetLogo
 Program for Activator/Inhibitor
 Pattern Formation**

[Run Pattern.nlogo](#)

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Continuous-time Activator-Inhibitor System

- Activator A and inhibitor I may diffuse at different rates in x and y directions
- Cell becomes more active if activator + bias exceeds inhibitor
- Otherwise, less active
- A and I are limited to $[0, 100]$ (depletion/saturation)

$$\frac{\partial A}{\partial t} = d_{Ax} \frac{\partial^2 A}{\partial x^2} + d_{Ay} \frac{\partial^2 A}{\partial y^2} + k_A(A + B - I)$$

$$\frac{\partial I}{\partial t} = d_{Ix} \frac{\partial^2 I}{\partial x^2} + d_{Iy} \frac{\partial^2 I}{\partial y^2} + k_I(A + B - I)$$

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**Demonstration of NetLogo
 Program for Activator/Inhibitor
 Pattern Formation
 with Continuous State Change**

[Run Activator-Inhibitor.nlogo](#)

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Turing Patterns

- Alan Turing studied the mathematics of reaction-diffusion systems
- Turing, A. (1952). The chemical basis of morphogenesis. *Philosophical Transactions of the Royal Society* **B 237**: 37–72.
- The resulting patterns are known as *Turing patterns*

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Observations

- With local activation and lateral inhibition
- And with a random initial state
- You can expect to get Turing patterns
- These are stationary states (dynamic equilibria)
- Macroscopically, Class I behavior
 - Microscopically, may be class III

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A Key Element of Self-Organization

- Activation vs. Inhibition
- Cooperation vs. Competition
- Amplification vs. Stabilization
- Growth vs. Limit
- Positive Feedback vs. Negative Feedback
 - Positive feedback creates
 - Negative feedback shapes

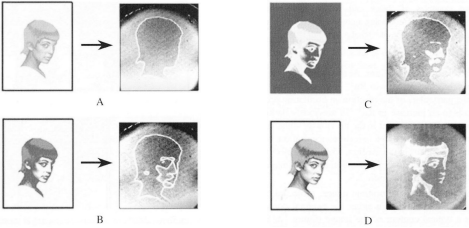
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Reaction-Diffusion Computing

- Has been used for image processing
 - diffusion \Rightarrow noise filtering
 - reaction \Rightarrow contrast enhancement
- Depending on parameters, RD computing can:
 - restore broken contours
 - detect edges
 - improve contrast

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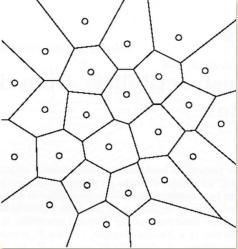
Image Processing in BZ Medium



- (A) boundary detection, (B) contour enhancement, (C) shape enhancement, (D) feature enhancement

2015/1/28 Image < Adamatzky, *Comp. in Nonlinear Media & Autom. Coll.* 26

Voronoi Diagrams



- Given a set of generating points:
- Construct a polygon around each generating point of set, so all points in a polygon are closer to its generating point than to any other generating points.

2015/1/28 Image < Adamatzky & al., *Reaction-Diffusion Computers* 27

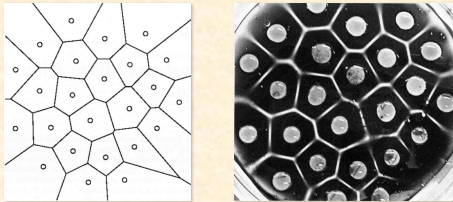
Some Uses of Voronoi Diagrams

- Collision-free path planning
- Determination of service areas for power substations
- Nearest-neighbor pattern classification
- Determination of largest empty figure

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Computation of Voronoi Diagram by Reaction-Diffusion Processor

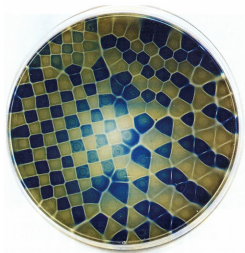


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Image < Adamatzky & al., *Reaction-Diffusion Computers*

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Mixed Cell Voronoi Diagram

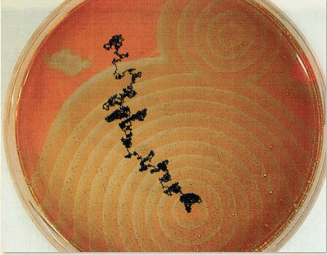


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Image < Adamatzky & al., *Reaction-Diffusion Computers*

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Path Planning via BZ medium:
No Obstacles



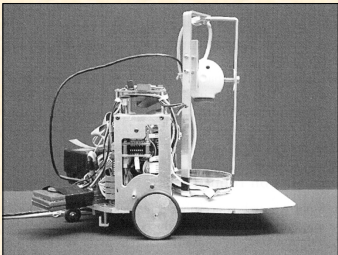
2015/1/28 Image < Adamatzky & al., *Reaction-Diffusion Computers* 31

Path Planning via BZ medium:
Circular Obstacles



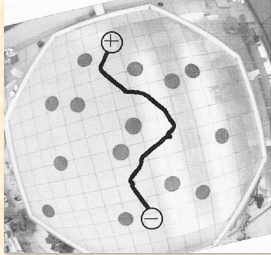
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Mobile Robot with Onboard
Chemical Reactor



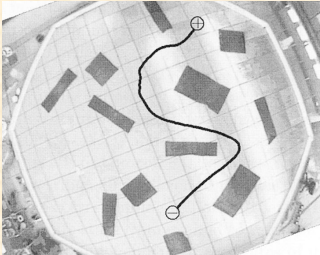
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Actual Path: Pd Processor



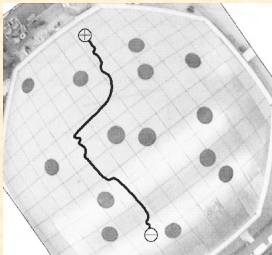
2015/1/28 Image < Adamatzky & al., *Reaction-Diffusion Computers* 34

Actual Path: Pd Processor



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Actual Path: BZ Processor



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**Bibliography for
Reaction-Diffusion Computing**

1. Adamatzky, Adam. *Computing in Nonlinear Media and Automata Collectives*. Bristol: Inst. of Physics Publ., 2001.
2. Adamatzky, Adam, De Lacy Costello, Ben, & Asai, Tetsuya. *Reaction Diffusion Computers*. Amsterdam: Elsevier, 2005.

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