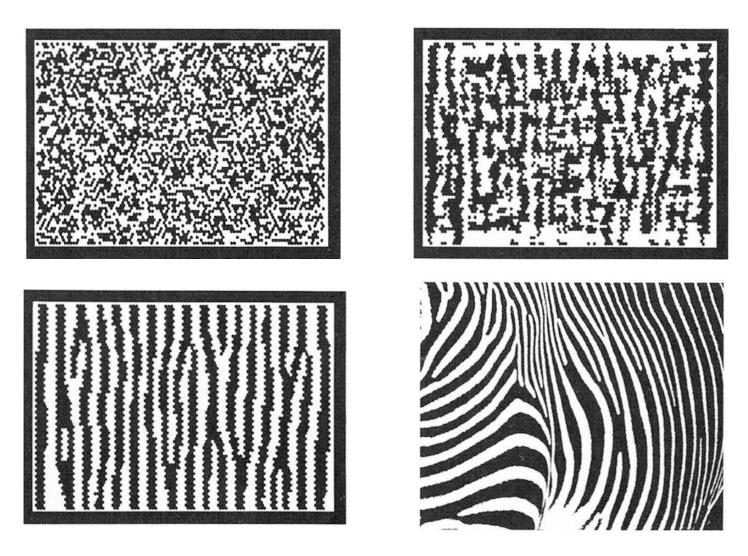
B. Pattern Formation

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

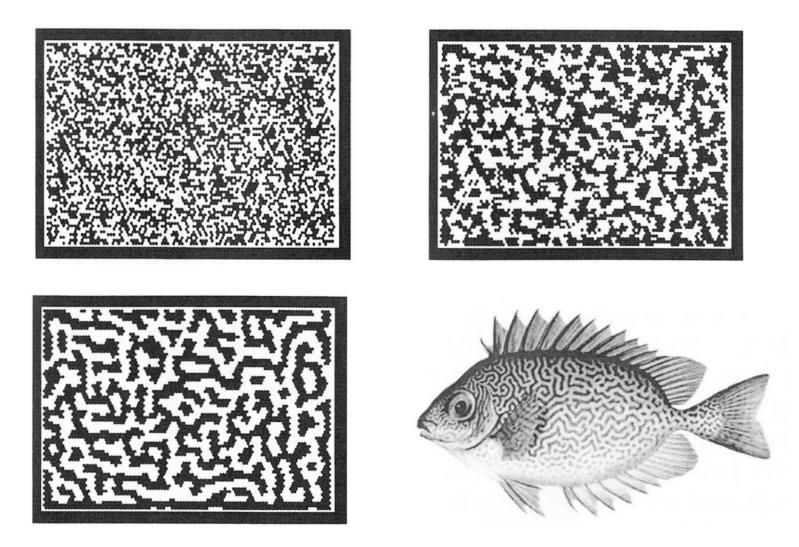
Zebra



9/8/09

figs. from Camazine & al.: Self-Org. Biol. Sys.

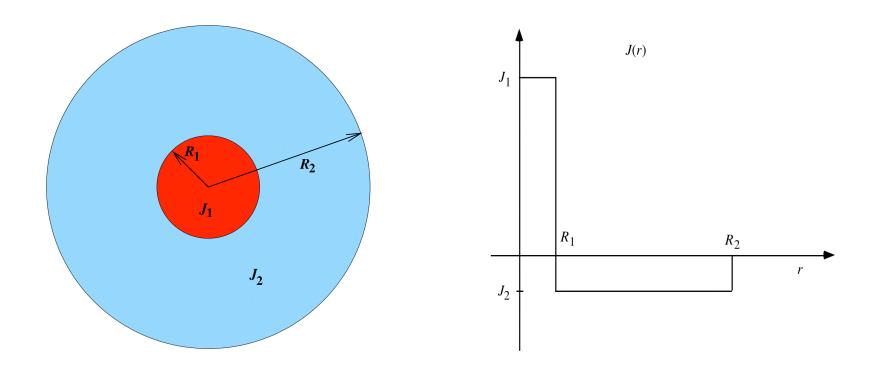
Vermiculated Rabbit Fish



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 ⇒ local uniformity
 - long-range inhibition \Rightarrow separation

Interaction Parameters



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

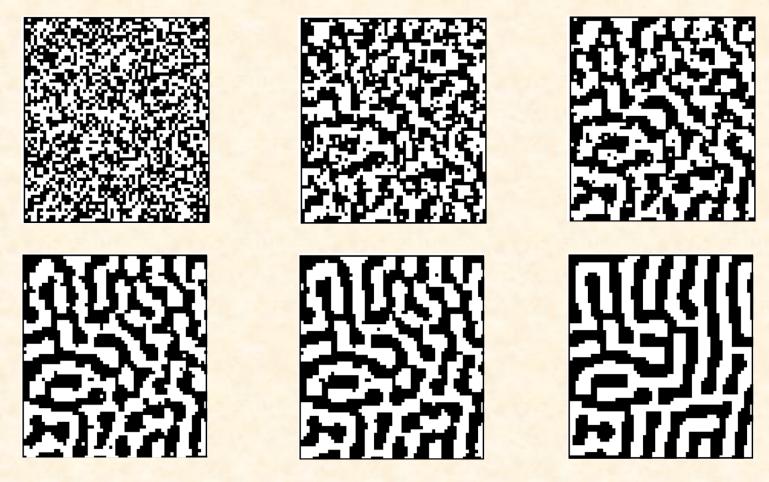
CA Activation/Inhibition Model

- 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:

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

Example

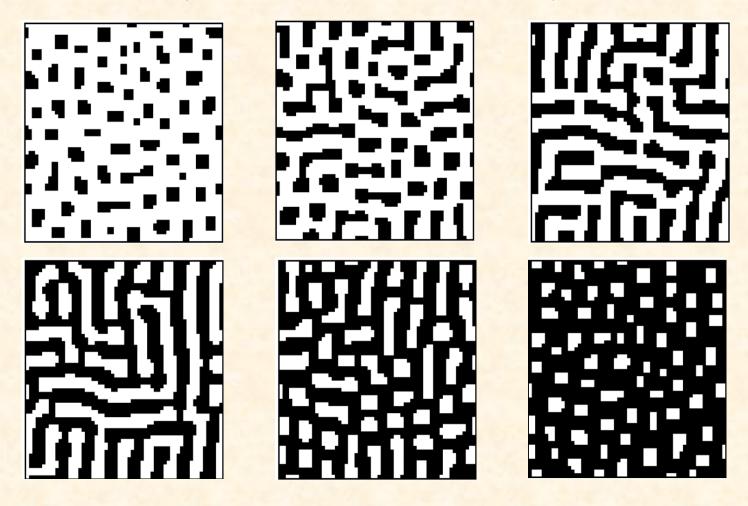
 $(R_1=1, R_2=6, J_1=1, J_2=-0.1, h=0)$



figs. from Bar-Yam

Effect of Bias

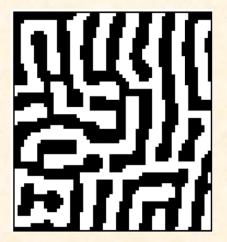
(h = -6, -3, -1; 1, 3, 6)

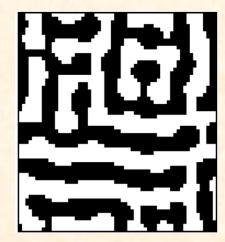


figs. from Bar-Yam

Effect of Interaction Ranges







$$R_2 = 8$$

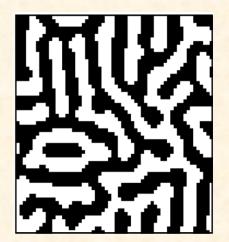
$$R_1 = 1$$

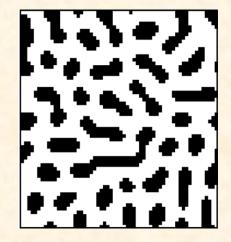
$$h = 0$$

$$R_2 = 6$$

$$R_1 = 1.5$$

$$h = 0$$





$$R_2 = 6$$

 $R_1 = 1.5$
 $h = -3$

Demonstration of NetLogo Program for Activation/Inhibition Pattern Formation: Fur

RunAICA.nlogo

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)

Digression on Diffusion

• Simple 2-D diffusion equation:

$$\dot{A}(x,y) = c\nabla^2 A(x,y)$$

• 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

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

$$\frac{\partial}{\partial t} \begin{pmatrix} A \\ I \end{pmatrix} = \begin{pmatrix} d_{A} & 0 \\ 0 & d_{I} \end{pmatrix} \begin{pmatrix} \nabla^{2} A \\ \nabla^{2} I \end{pmatrix} + \begin{pmatrix} f_{A}(A, I) \\ f_{I}(A, I) \end{pmatrix}$$

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

Example:

Activation-Inhibition System

- Let σ be some kind of threshold function
- Activator A and inhibitor I may diffuse at different rates in x and y directions
- Cell is "on" if activator + bias exceeds inhibitor

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

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

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

Demonstration of NetLogo Program for Activation/Inhibition Pattern Formation

Run Pattern.nlogo

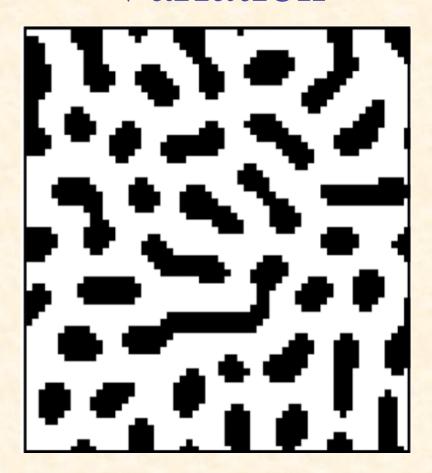
Turing Patterns

- 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

Abstract Activation/Inhibition Spaces

- Consider two axes of cultural preference
 - E.g. hair length & interpersonal distance
 - Fictitious example!
- Suppose there are no objective reasons for preferences
- Suppose people approve/encourage those with similar preferences
- Suppose people disapprove/discourage those with different preferences
- What is the result?

Emergent Regions of Acceptable Variation



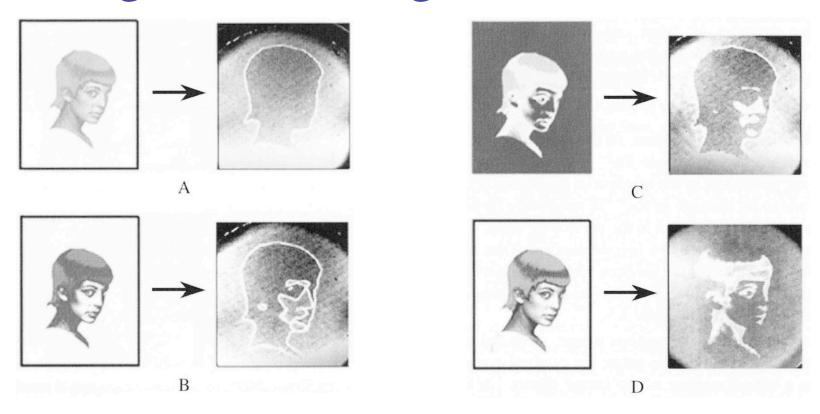
A Key Element of Self-Organization

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

Reaction-Diffusion Computing

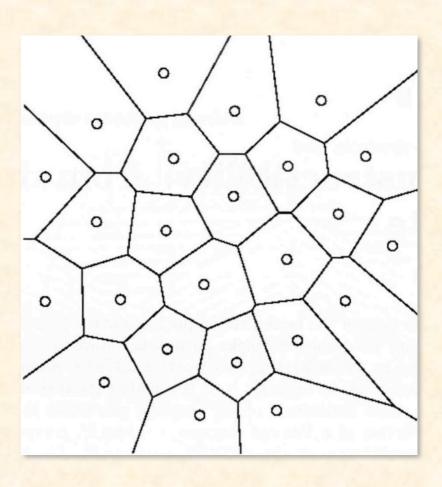
- Has been used for image processing
 - diffusion ⇒ noise filtering
 - reaction \Rightarrow contrast enhancement
- Depending on parameters, RD computing can:
 - restore broken contours
 - detect edges
 - improve contrast

Image Processing in BZ Medium



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

Voronoi Diagrams

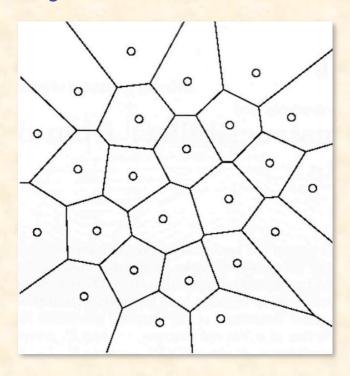


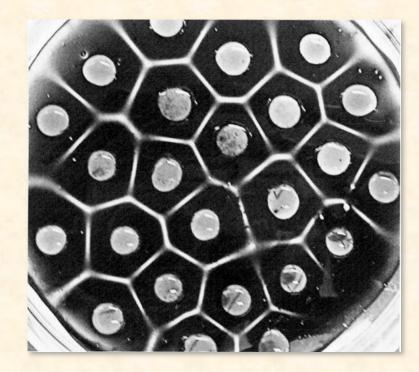
- Given a set of generating points:
- Construct polygon around each gen. point of set, so all points in poly. are closer to its generating point than to any other generating points.

Some Uses of Voronoi Diagrams

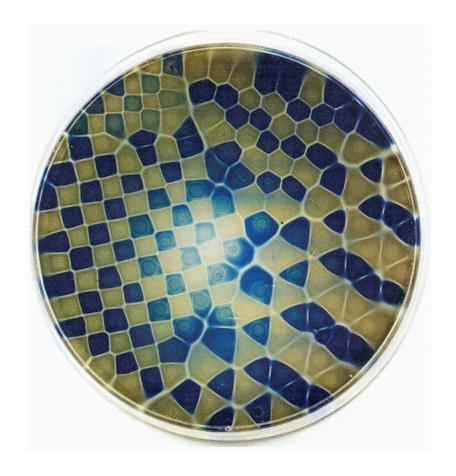
- Collision-free path planning
- Determination of service areas for power substations
- Nearest-neighbor pattern classification
- Determination of largest empty figure

Computation of Voronoi Diagram by Reaction-Diffusion Processor

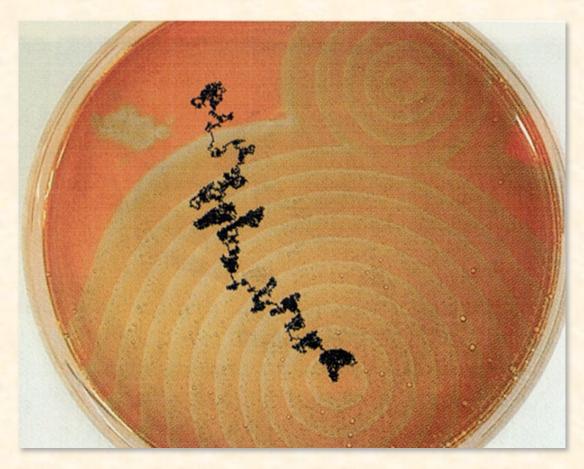




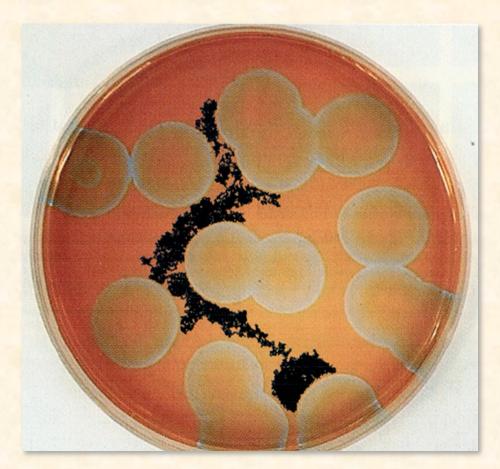
Mixed Cell Voronoi Diagram



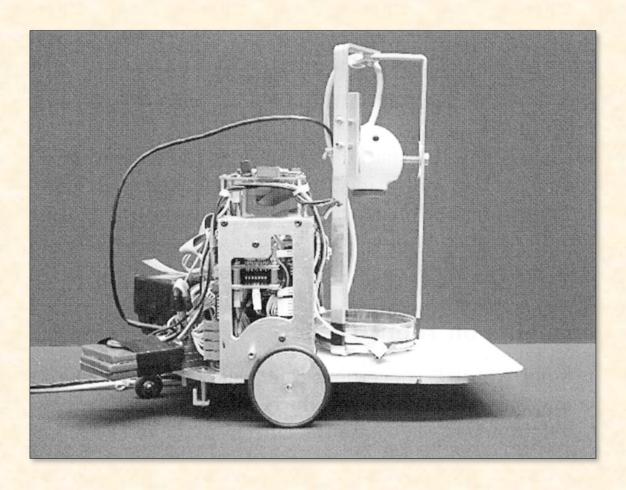
Path Planning via BZ medium: No Obstacles



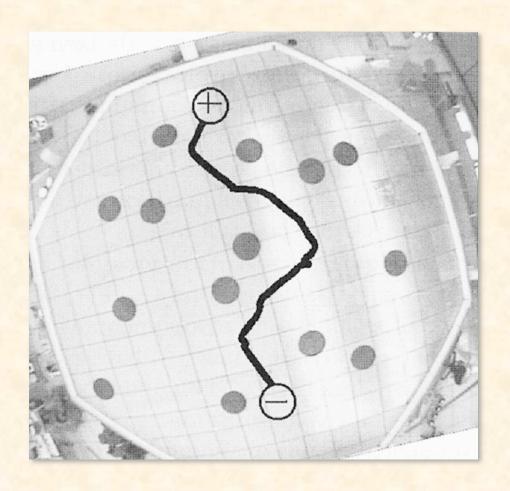
Path Planning via BZ medium: Circular Obstacles



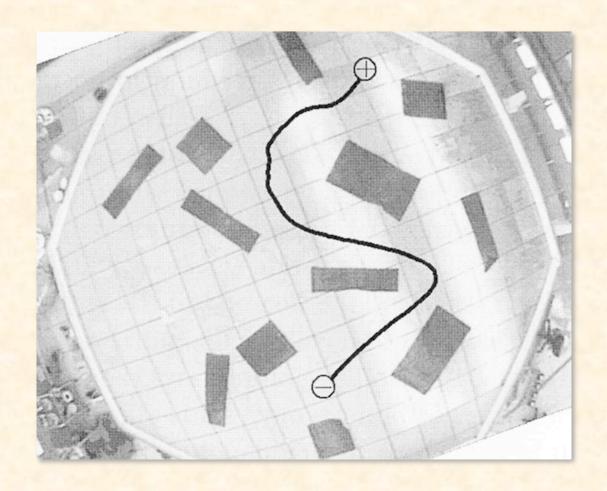
Mobile Robot with Onboard Chemical Reactor



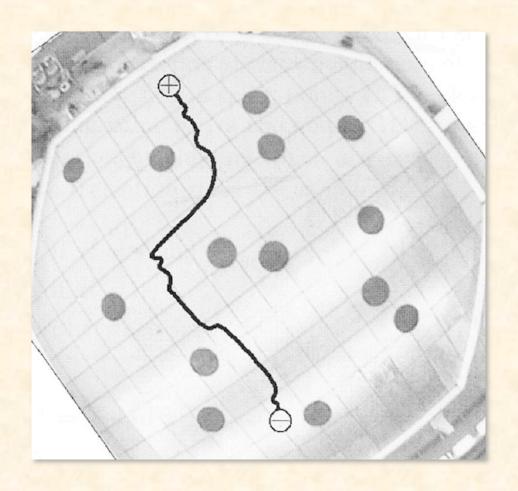
Actual Path: Pd Processor



Actual Path: Pd Processor



Actual Path: BZ Processor



Bibliography for Reaction-Diffusion Computing

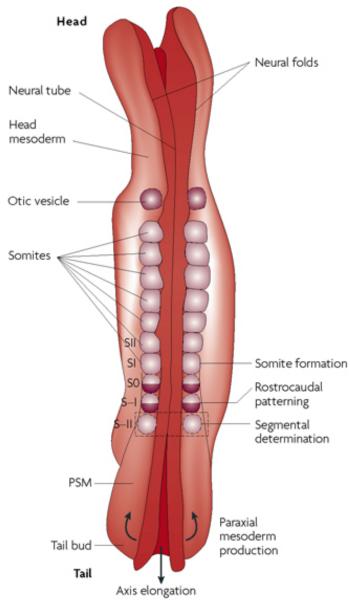
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- 2. Adamatzky, Adam, De Lacy Costello, Ben, & Asai, Tetsuya. *Reaction Diffusion Computers*. Amsterdam: Elsevier, 2005.

Segmentation

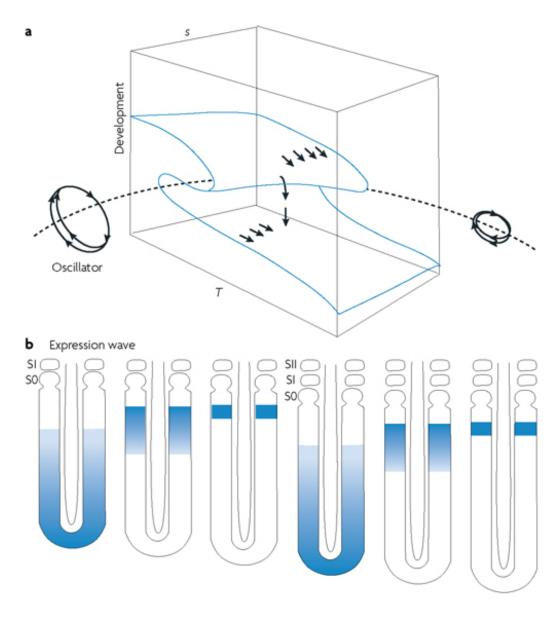
(in embryological development)

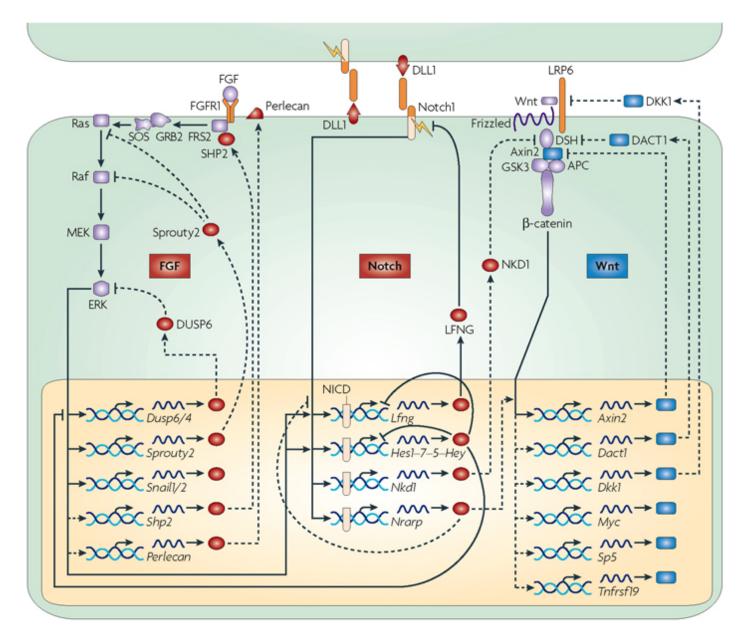
Vertebrae

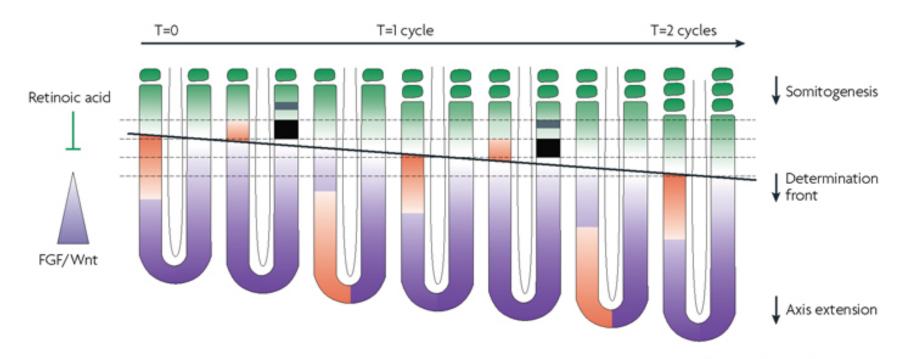
- Humans: 33, chickens: 55, mice: 65, corn snake: 315
- Characteristic of species
- How does an embryo "count" them?
- "Clock and wavefront model" of Cooke & Zeeman (1976).



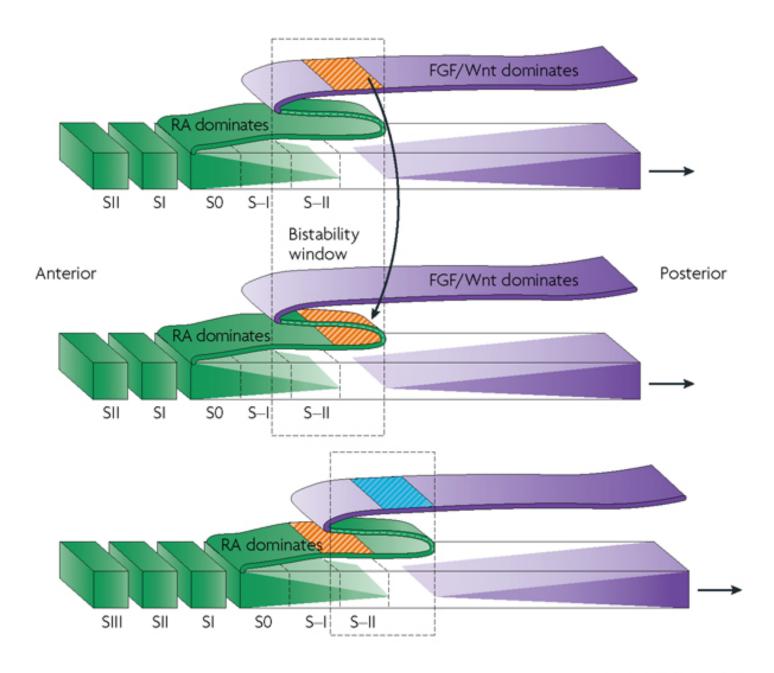
Nature Reviews | Genetics







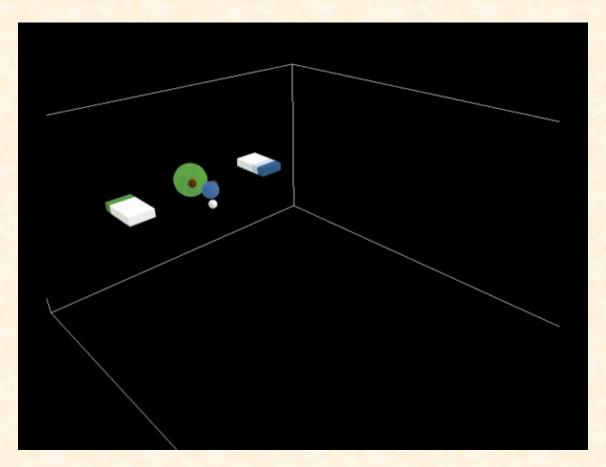
Nature Reviews | Genetics



NetLogo Simulation of Segmentation

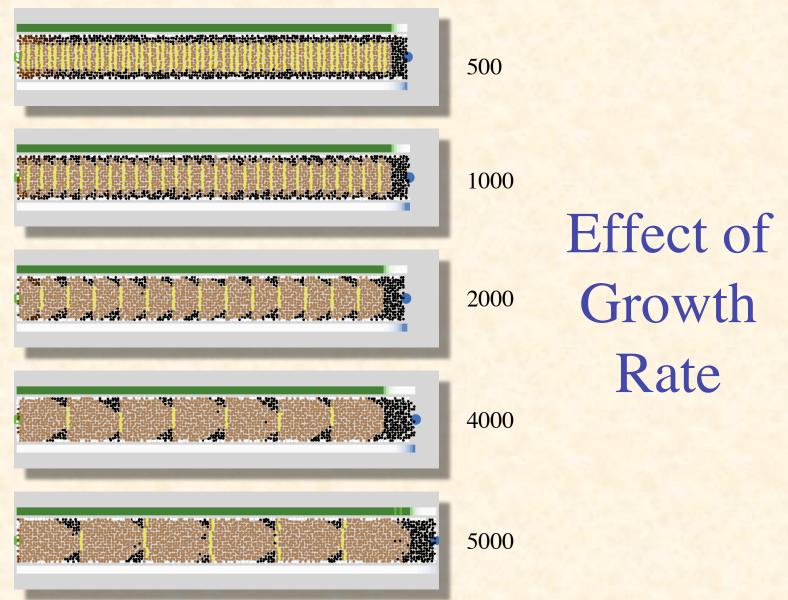
Run Segmentation.nlogo

Simulated Segmentation by Clock-and-Wavefront Process



2D Simulation of Clock-and-Wavefront Process





Segmentation References

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- 2. Dequéant, M.-L., & Pourquié, O. (2008). Segmental patterning of the vertebrate embryonic axis. *Nature Reviews Genetics* **9**: 370–82.
- 3. Gomez, C., Özbudak, E.M., Wunderlich, J., Baumann, D., Lewis, J., & Pourquié, O. (2008). Control of segment number in vertebrate embryos. *Nature* **454**: 335–9.

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- 2. Gerhardt, M., Schuster, H., & Tyson, J. J. "A Cellular Automaton Model of Excitable Media Including Curvature and Dispersion," *Science* **247** (1990): 1563-6.
- 3. Tyson, J. J., & Keener, J. P. "Singular Perturbation Theory of Traveling Waves in Excitable Media (A Review)," *Physica D* **32** (1988): 327-61.
- 4. Camazine, S., Deneubourg, J.-L., Franks, N. R., Sneyd, J., Theraulaz, G.,& Bonabeau, E. *Self-Organization in Biological Systems*. Princeton, 2001.
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- 6. Solé, R., & Goodwin, B. Signs of Life: How Complexity Pervades Biology. Basic Books, 2000.