



Dynamical systems and CA

- Outline

- Dynamical systems
- Cellular automata (CA) as models of nonlinear dynamical systems
- Complexity



Dynamical systems and CA

■ References

- Gilbert, N. and Troitzsch, K.G. (1999), *Simulation for the Social Scientist*, Buckingham, Philadelphia: Open University Press, Chapter 7.
(<http://www.uni-koblenz.de/~kgt/Learn/Textbook/Book.html>)
- Foley, D.K. (1998), "Introduction", in P.S. Albin, *Barriers and Bounds to Rationality*, Princeton, NJ: Princeton University Press, Chapter 1.
- Albin, B.S. (1998), *Barriers and Bounds to Rationality*, Princeton, NJ: Princeton University Press, Chapters 3, 4 and 5.
- Day, R.H. (1982), "Irregular Growth Cycles", *American Economic Review* (72): 406-414.



Dynamical systems

- A dynamical system can be defined by the relation

$$x_{t+1} = F_a(x_t)$$

where x_t is a vector in a state space X , F_a an operator on the state space, a the vector of potentially changeable parameters of the system.



Linear dynamical systems

- In a linear dynamical system, F_a is a matrix, A

$$x_{t+1} = A(x_t)$$

A matrix determines the trajectories of the system that can be decomposed into three types of independent motion.



Linear dynamical systems

- Geometric expansion or contraction
 - Eigenvalues of A are real, positive
- Oscillating geometric expansion or contraction
 - Eigenvalues of A are real, negative
- Inward or outward spiraling
 - Eigenvalues of A are complex



Linear dynamical systems

- Stability of linear dynamical systems
 - The magnitude of the eigenvalues determines the stability of the system.
 - If an eigenvalue (or pair of complex eigenvalues) has a magnitude smaller than 1, the corresponding component of the motion will be stable, moving to the origin (the “equilibrium” point).
 - If an eigenvalue (or pair of complex eigenvalues) has a magnitude greater than 1, the corresponding component of the motion will be unstable, moving away to the origin indefinitely.
 - Eigenvalues with magnitude just equal to 1 are neutral, neither stable nor unstable.



Linear dynamical systems

- Example

$$x_{t+1} = \lambda \cos\omega x_t - \lambda \sin\omega y_t$$

$$y_{t+1} = \lambda \sin\omega x_t + \lambda \cos\omega y_t$$

- The eigenvalues are
 $\lambda (\cos\omega \pm i \sin\omega)$

If $\lambda < 1$, stable; if $\lambda > 1$, unstable.

ω determines the speed and direction of spiraling.



Linear dynamical systems

Example

Rewrite the system in polar coordinates

$$r(x, y) = (x^2 + y^2)^{1/2}$$

$$\Theta = \arctan(y/x)$$

Then,

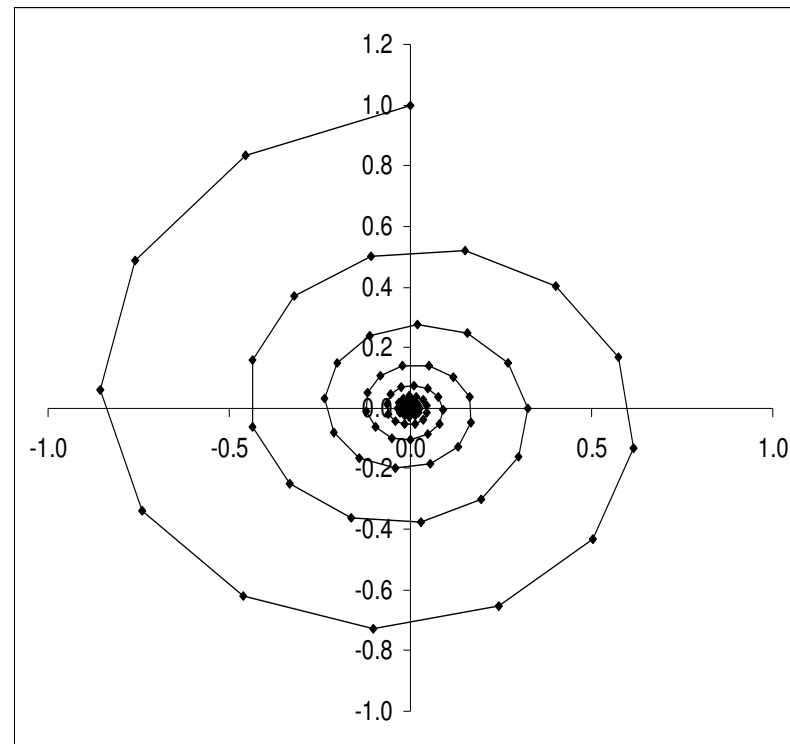
$$r_{t+1} = \lambda r_t$$

$$\Theta_{t+1} = \Theta_t + \omega$$



Linear dynamical systems

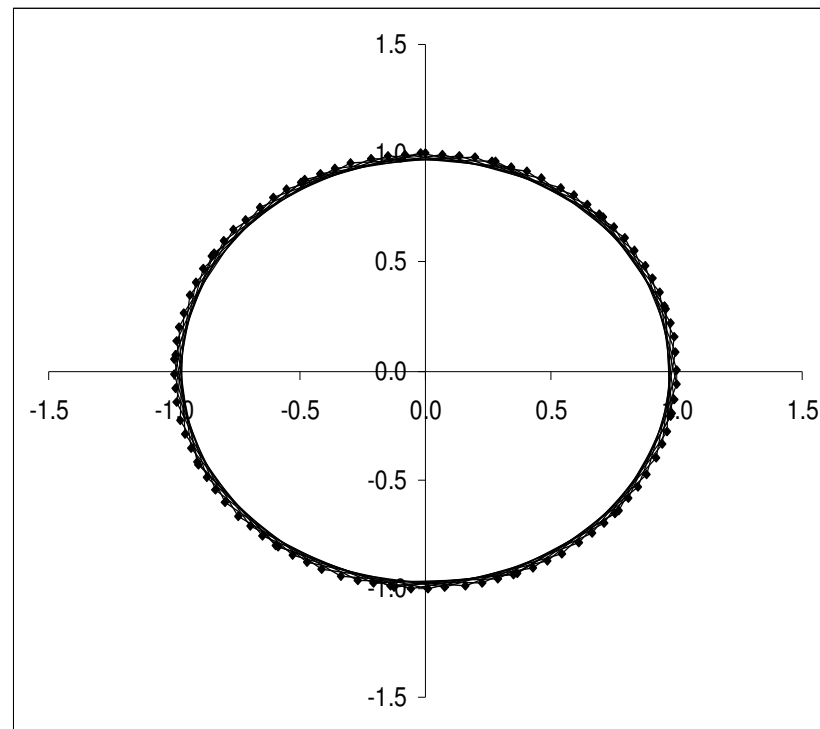
- $\lambda = 0.95, \omega = 0.5, x_0 = 0, y_0 = 1$





Linear dynamical systems

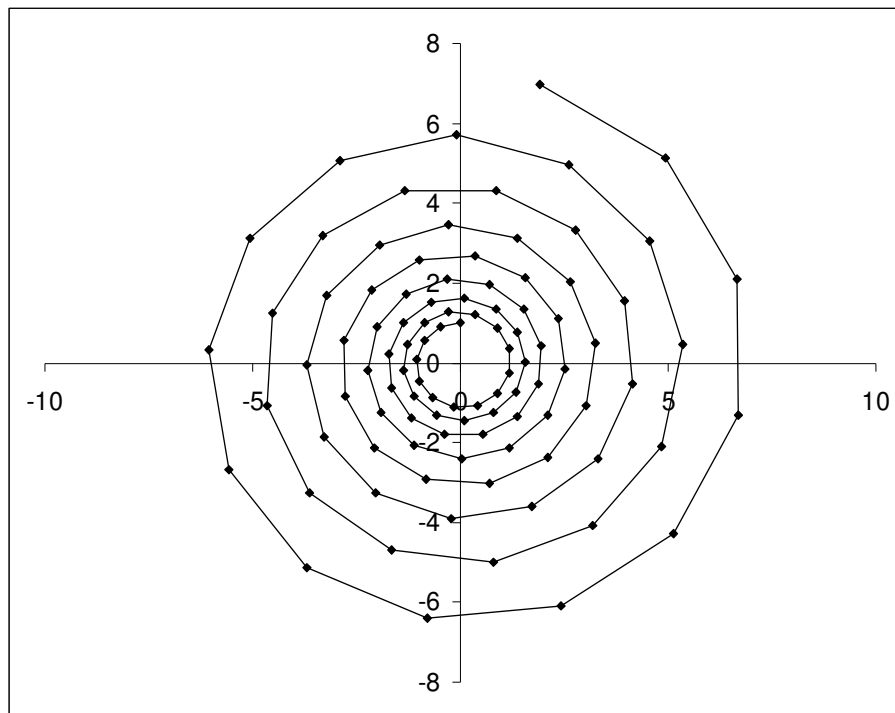
- $\lambda = 1, \omega = 0.5, x_0 = 0, y_0 = 1$





Linear dynamical systems

- $\lambda = 1.02, \omega = 0.5, x_0 = 0, y_0 = 1$





Linear dynamical systems

Characteristics of linear systems

- It is possible to predict the motion of linear systems by observing their past trajectories (for example, estimate the A matrix).
- Their behavior in all regions of the state space is proportional to their behavior in a small neighborhood of the origin (i.e., the observation of a linear system in a small part of state space is potentially sufficient to understand its behavior everywhere).
- Smooth changes in parameter values a (the elements of the matrix A) lead to smooth changes in the behavior of the trajectories (i.e., estimation errors in parameters/initial conditions do not create a major problem).



Nonlinear dynamical systems

- If the operator $F_a(\cdot)$ is not linear, then the system is nonlinear.
- Assume that the origin of the state space is an equilibrium of the system, i.e.,
$$F_a(0) = 0$$
- A nonlinear system can be approximated by linear systems (a power series expansion, for example, Taylor series).



Nonlinear dynamical systems

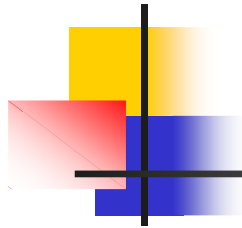
Types of motion in nonlinear systems

- Moves toward the origin
- Develops a stable limit cycles
- Multiple cycles
- Develop a “chaotic” behavior

Consider the following nonlinear system

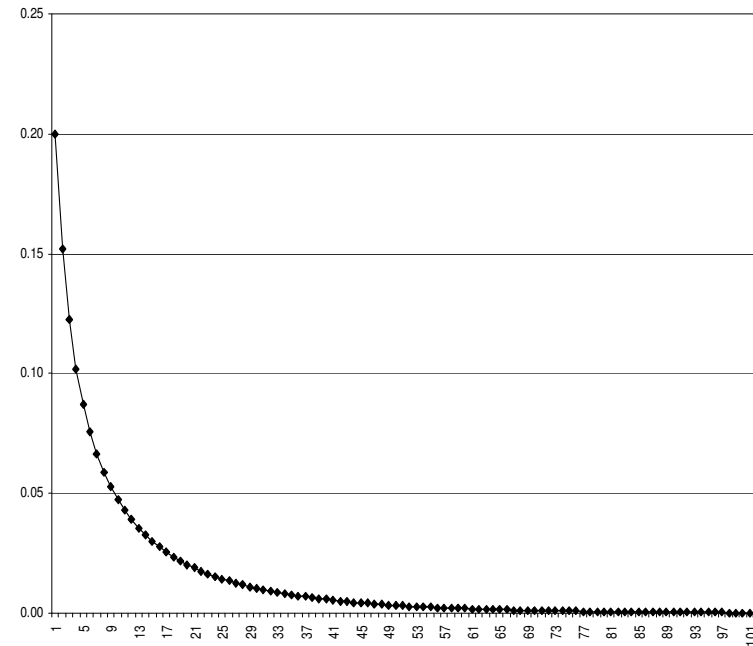
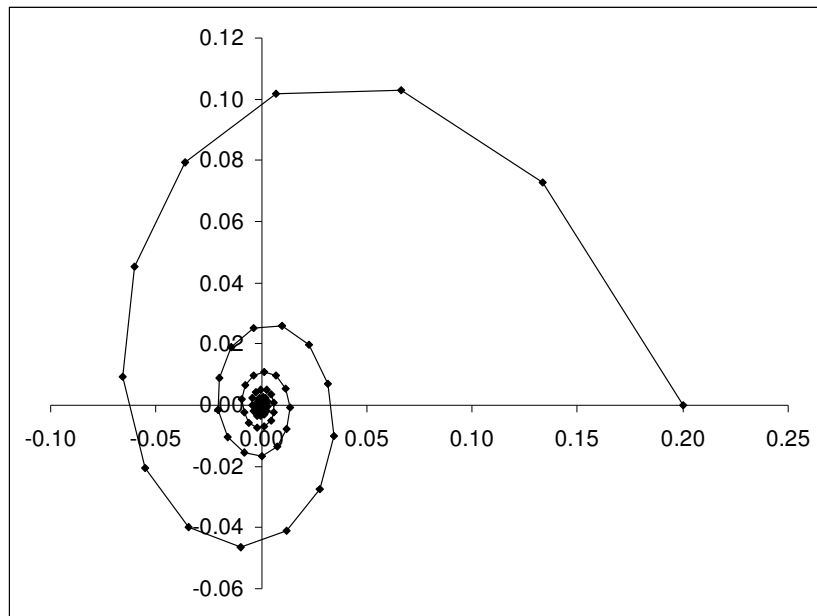
$$r_{t+1} = \lambda r_t - \lambda r_t^2 = \lambda r_t(1 - r_t)$$

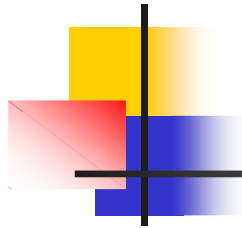
$$\Theta_{t+1} = \Theta_t + \omega$$



Nonlinear dynamical systems

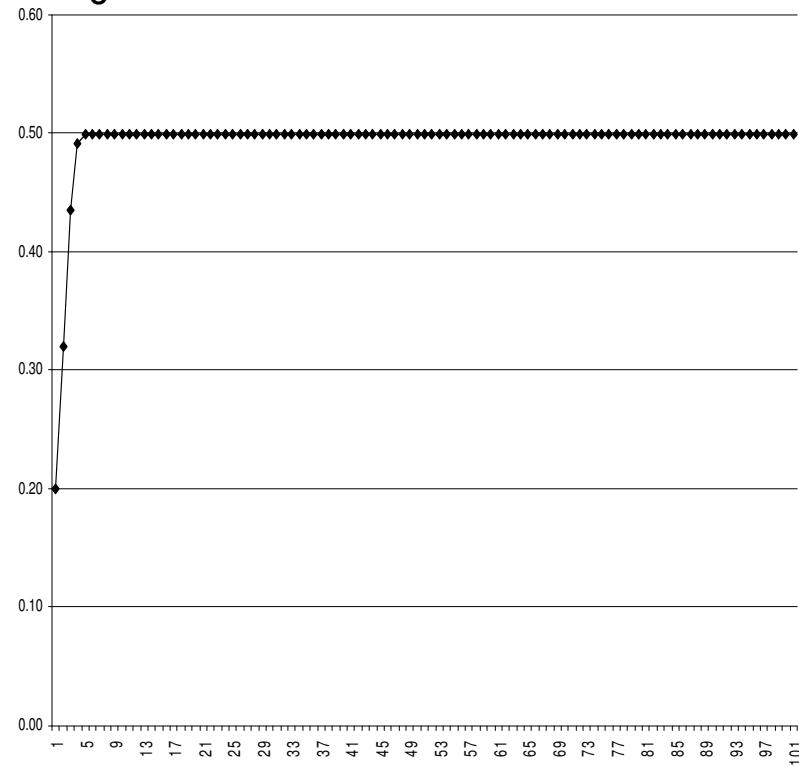
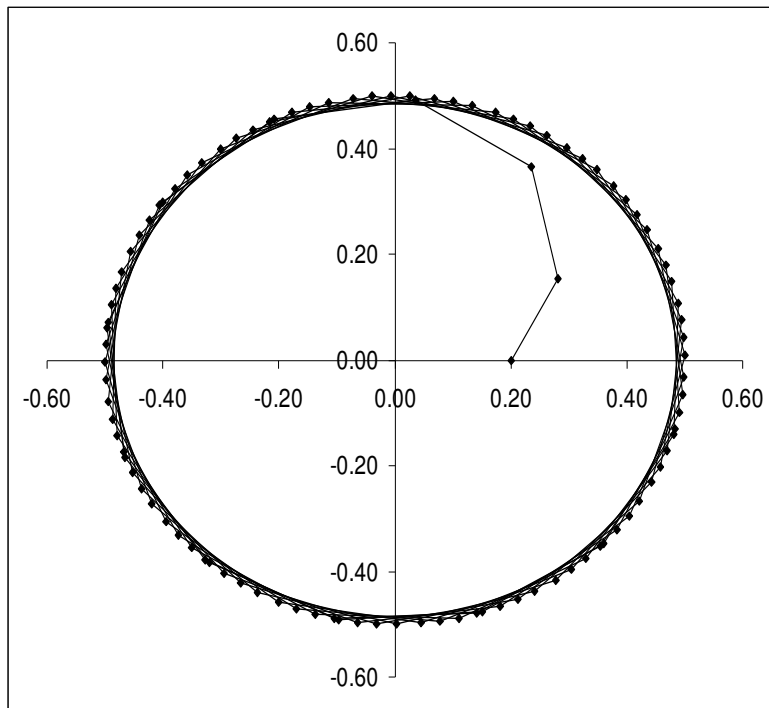
- $\lambda = 0.95, \omega = 0.5, x_0 = 0.2, y_0 = 0$





Nonlinear dynamical systems

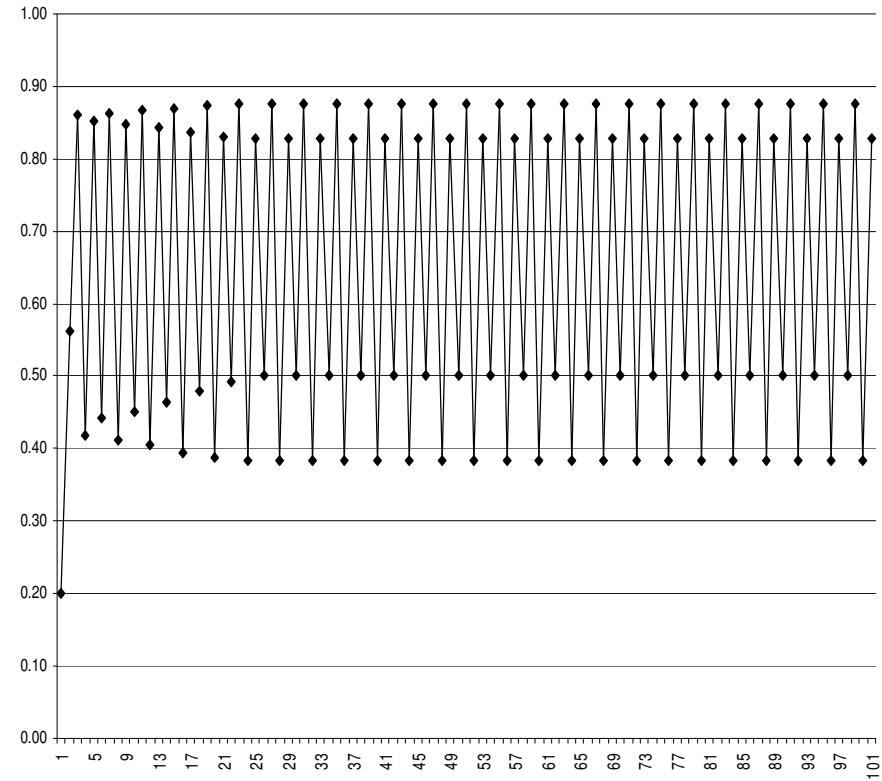
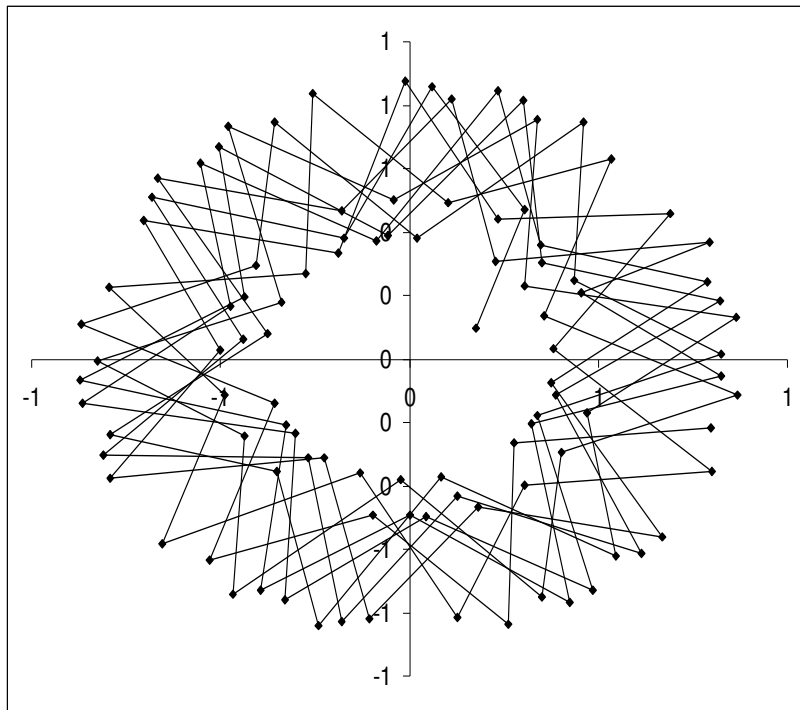
- $\lambda = 2, \omega = 0.5, x_0 = 0.2, y_0 = 0$





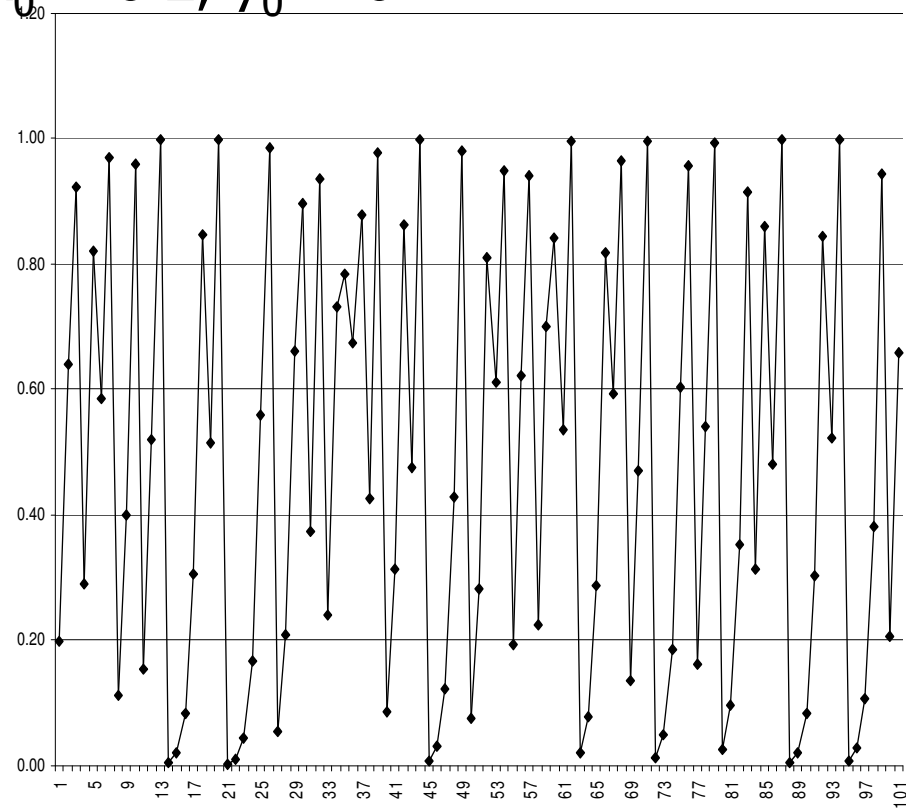
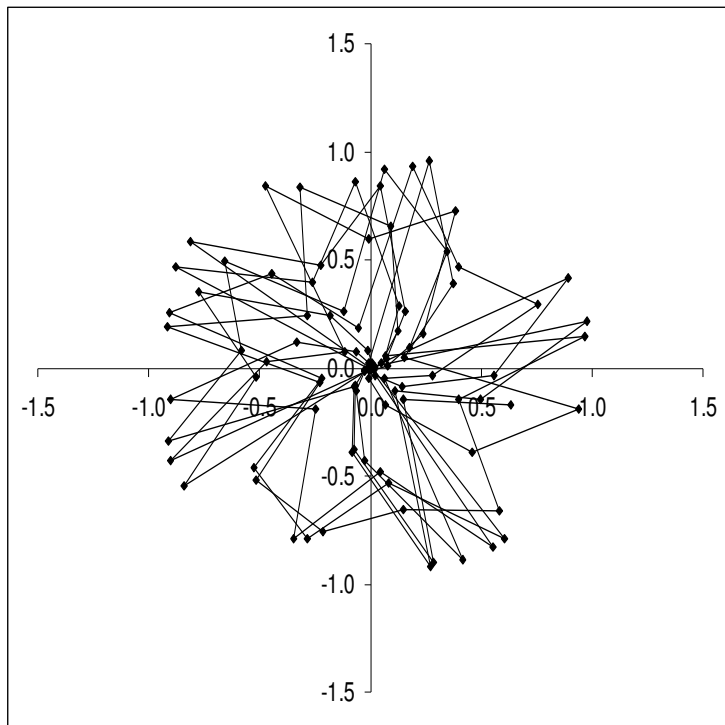
Nonlinear dynamical systems

- $\lambda = 3.5, \omega = 0.5, x_0 = 0.2, y_0 = 0$



Nonlinear dynamical systems

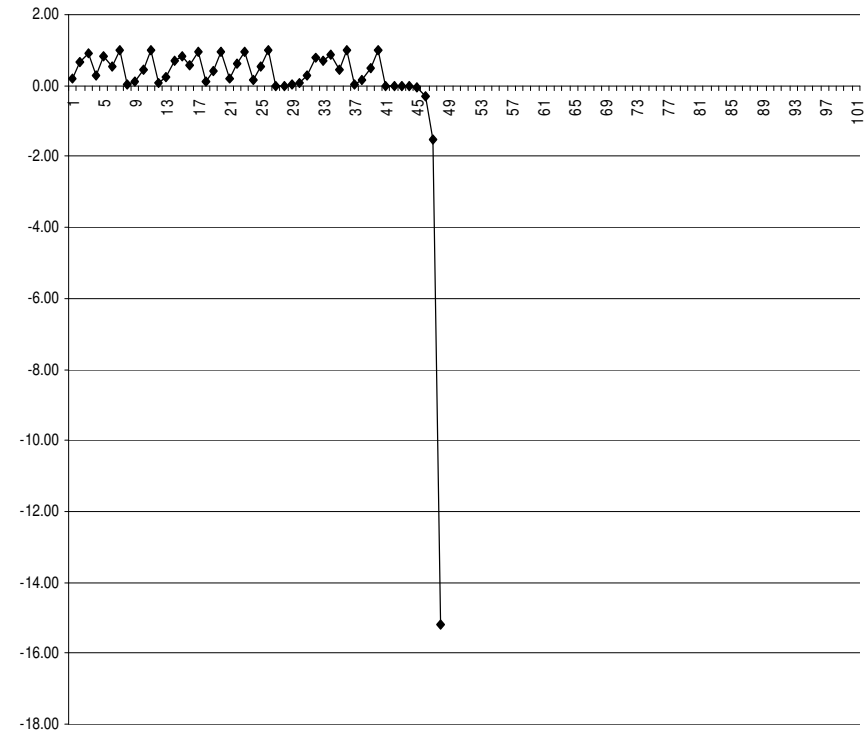
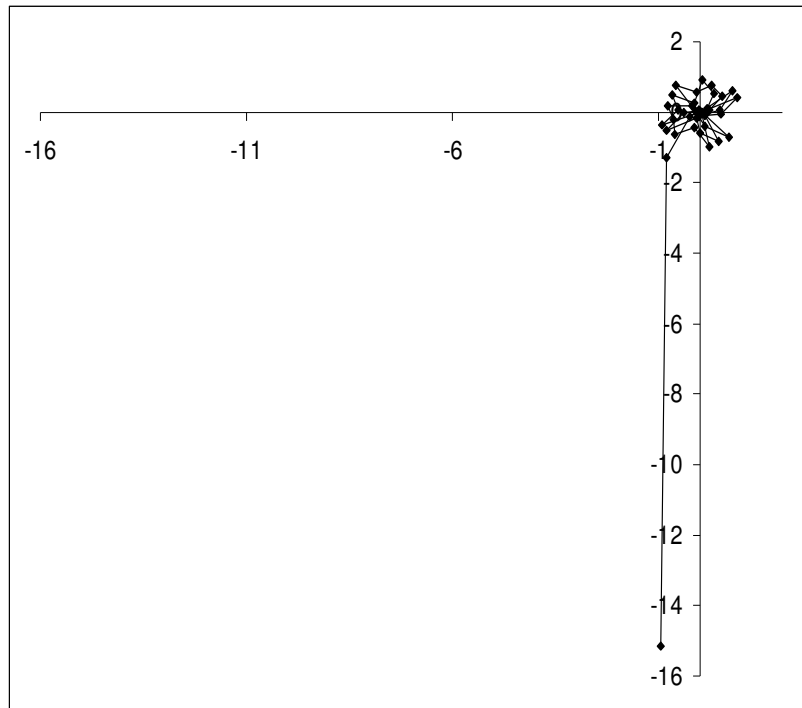
- $\lambda = 4.001, \omega = 0.5, x_0 = 0.2, y_0 = 0$

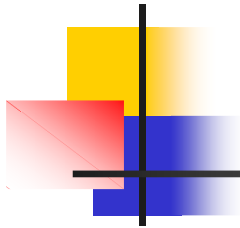




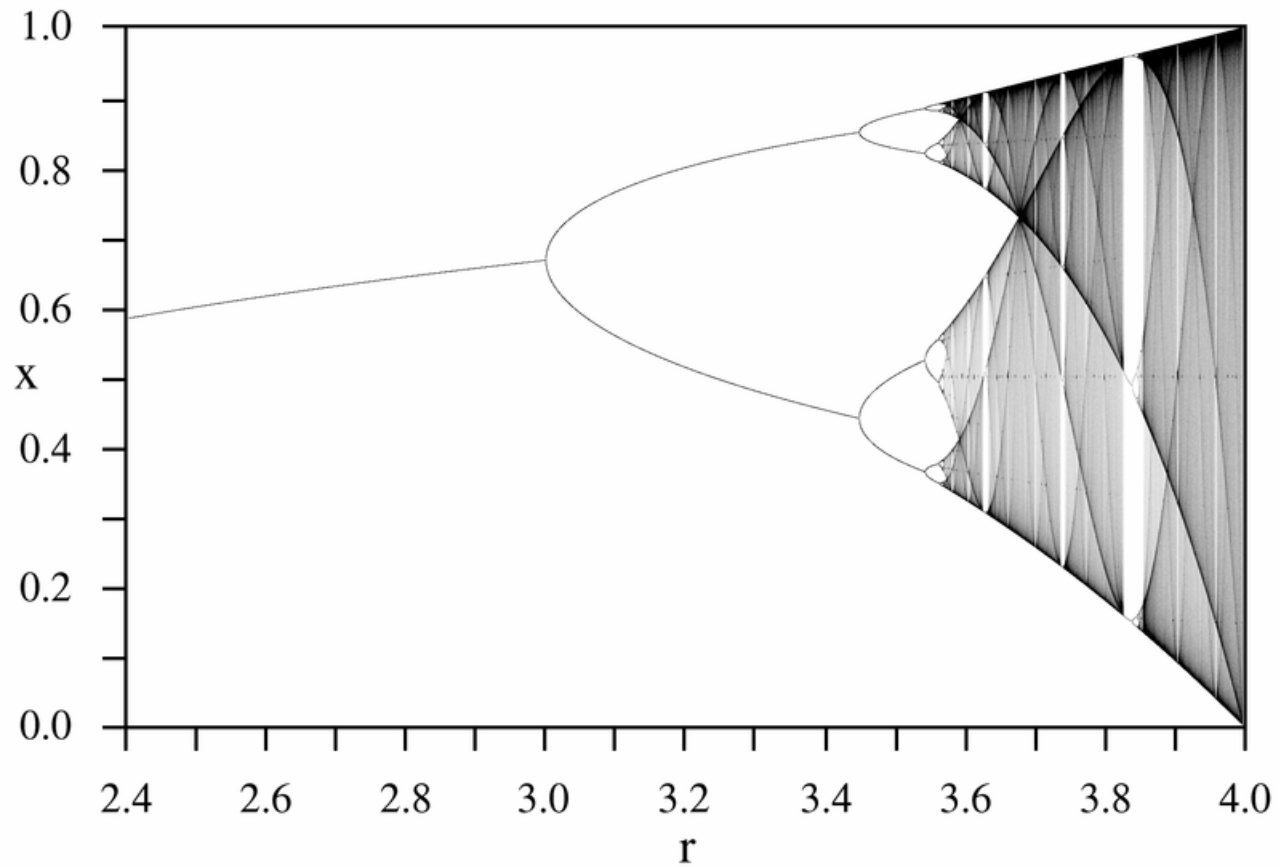
Nonlinear dynamical systems

- $\lambda = 4.001, \omega = 0.5, x_0 = 0.201, y_0 = 0$





Nonlinear dynamical systems





Nonlinear dynamical systems

By varying the parameter λ , the following behaviour is observed:

- With λ between 0 and 1, the population will eventually die, independent of the initial population.
- With λ between 1 and 2, the population will quickly stabilize on the value $(\lambda-1)/\lambda$, independent of the initial population.
- With λ between 2 and 3, the population will also eventually stabilize on the same value, but first oscillates around that value for some time.
- With λ between 3 and $1+\sqrt{6}$ (approximately 3.45), the population will oscillate between two values forever. These two values are dependent on λ but independent of the initial population.
- With λ between 3.45 and 3.54 (approximately), the population will oscillate between four values forever; again, this behavior does not depend on the initial population.



Nonlinear dynamical systems

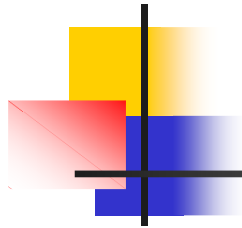
- With λ slightly bigger than 3.54, the population will oscillate between 8 values, then 16, 32, etc. The lengths of the parameter intervals which yield the same number of oscillations decrease rapidly; the ratio between the lengths of two successive such bifurcation intervals approaches the Feigenbaum constant $\delta = 4.669$. None of these behaviors depend on the initial population.
- At $\lambda = 3.57$ (approximately) is the onset of chaos, at the end of the period-doubling cascade. We can no longer see any oscillations. Slight variations in the initial population yield dramatically different results over time, a prime characteristic of chaos.
- Most values beyond 3.57 exhibit chaotic behaviour, but there are still certain isolated values of λ that appear to show non-chaotic behavior; these are sometimes called islands of stability. These behaviours are again independent of the initial value.
- Beyond $\lambda = 4$, the values eventually leave the interval $[0,1]$ and diverge for almost all initial values.



Nonlinear dynamical systems

A chaotic system

- “Globally stable”
- Never exactly repeats its previous trajectory
- **Locally predictable** (if the system returns to a position in state space close to an earlier trajectory, its trajectory will follow for a while the previous trajectory closely)
- **Exhibits regime shifts** (what happens if $r=1$?)
- **Trajectories depend on initial conditions/parameter values** (one trajectory, starting from one set of initial conditions, may be very different in detail from the trajectory starting from a very slightly different set of initial conditions)



Complexity

Types of complexity

- Type 1. Convergence
- Type 2. Repetitive patterns
- Type 3. Chaotic
- Type 4. Chaotic with regime shifts



Nonlinear dynamical systems

Examples of chaotic systems?



Nonlinear dynamical systems

Solovian neo-classical growth model

Assumptions

- Competitive markets
- Constant returns to scale
- Constant growth rate of labor, n
- Constant savings rate, $S = sQ$



Nonlinear dynamical systems

Equilibrium

$$k^* = k \quad (k = K/L)$$

Growth rate of K = Growth rate of labor (n)

$$sQ/K = n$$

$$sq = nk$$

Dynamics

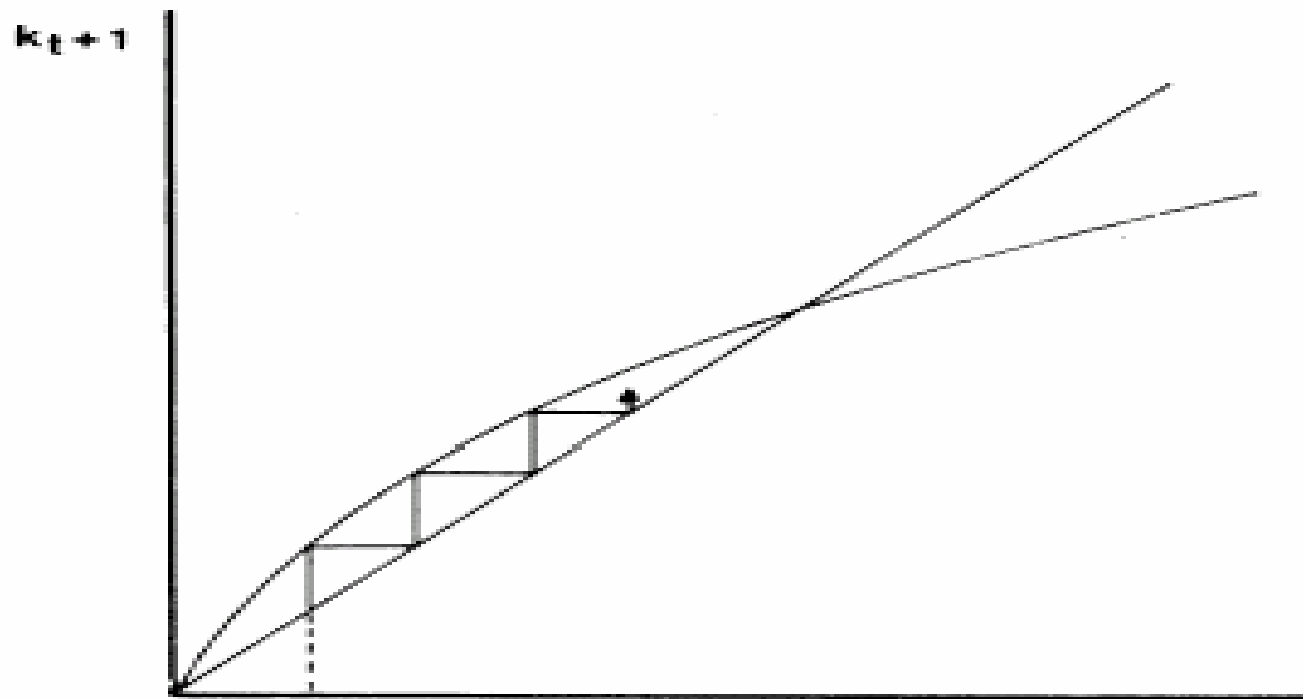
$$k_{t+1} = s f(k_t)/(1+n)$$

Assume a Cobb Douglas PF

$$k_{t+1} = s Bk_t^\beta/(1+n)$$



Nonlinear dynamical systems



(a) Monotonic Convergence to a Steady State



Nonlinear dynamical systems

Assume the “pollution effect”

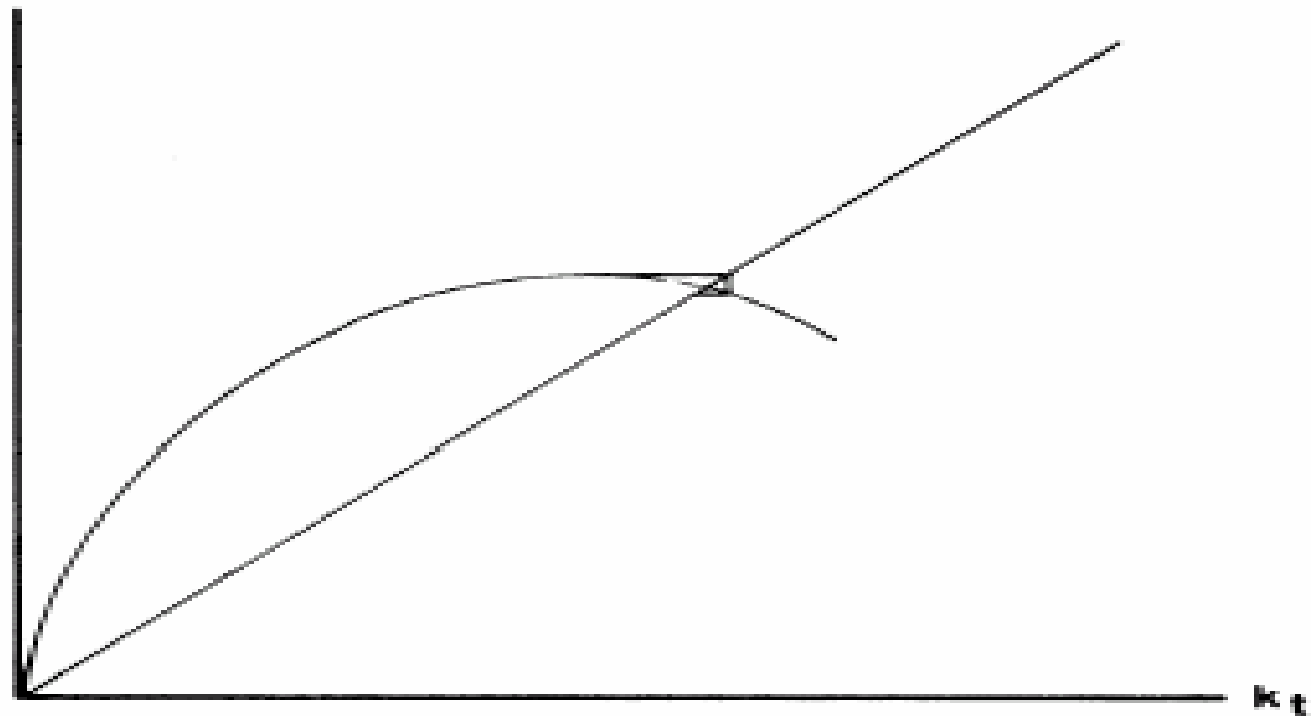
$$f^P(k) = f(k)(m - k)^\gamma$$

Then,

$$k_{t+1} = s B k_t^\beta (m - k_t)^\gamma / (1+n)$$



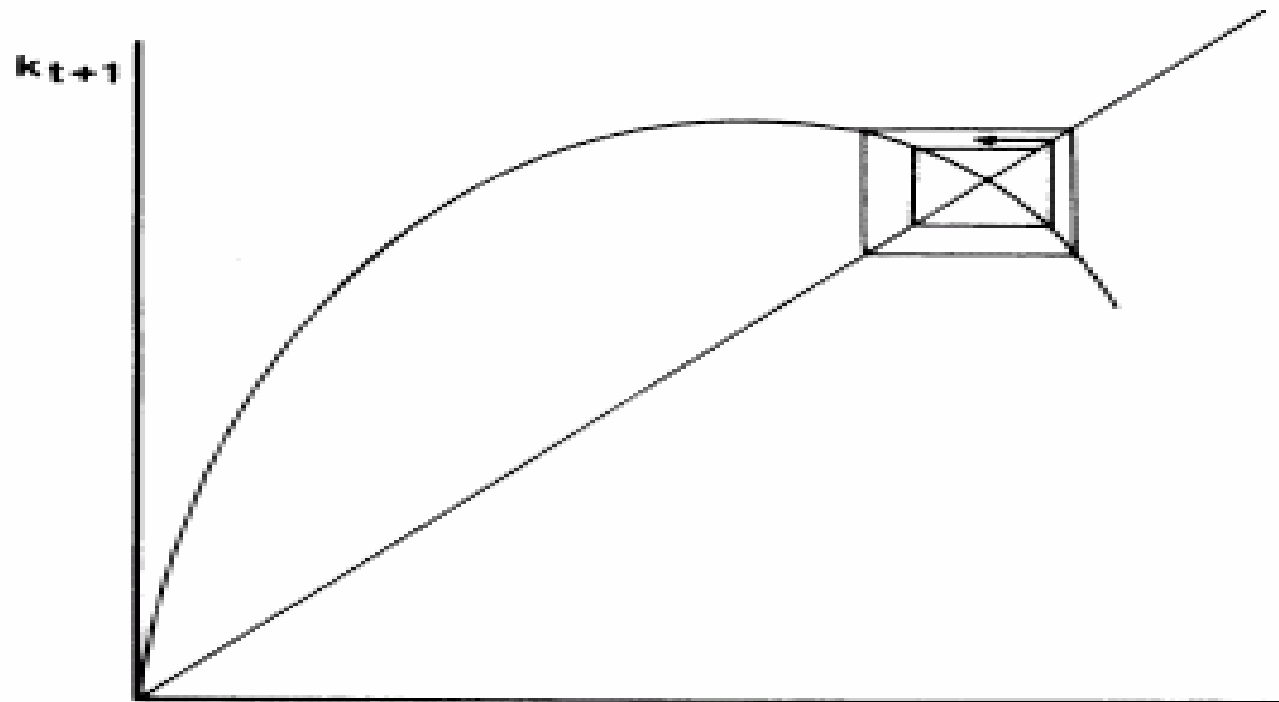
Nonlinear dynamical systems



(b) Damped Cycles



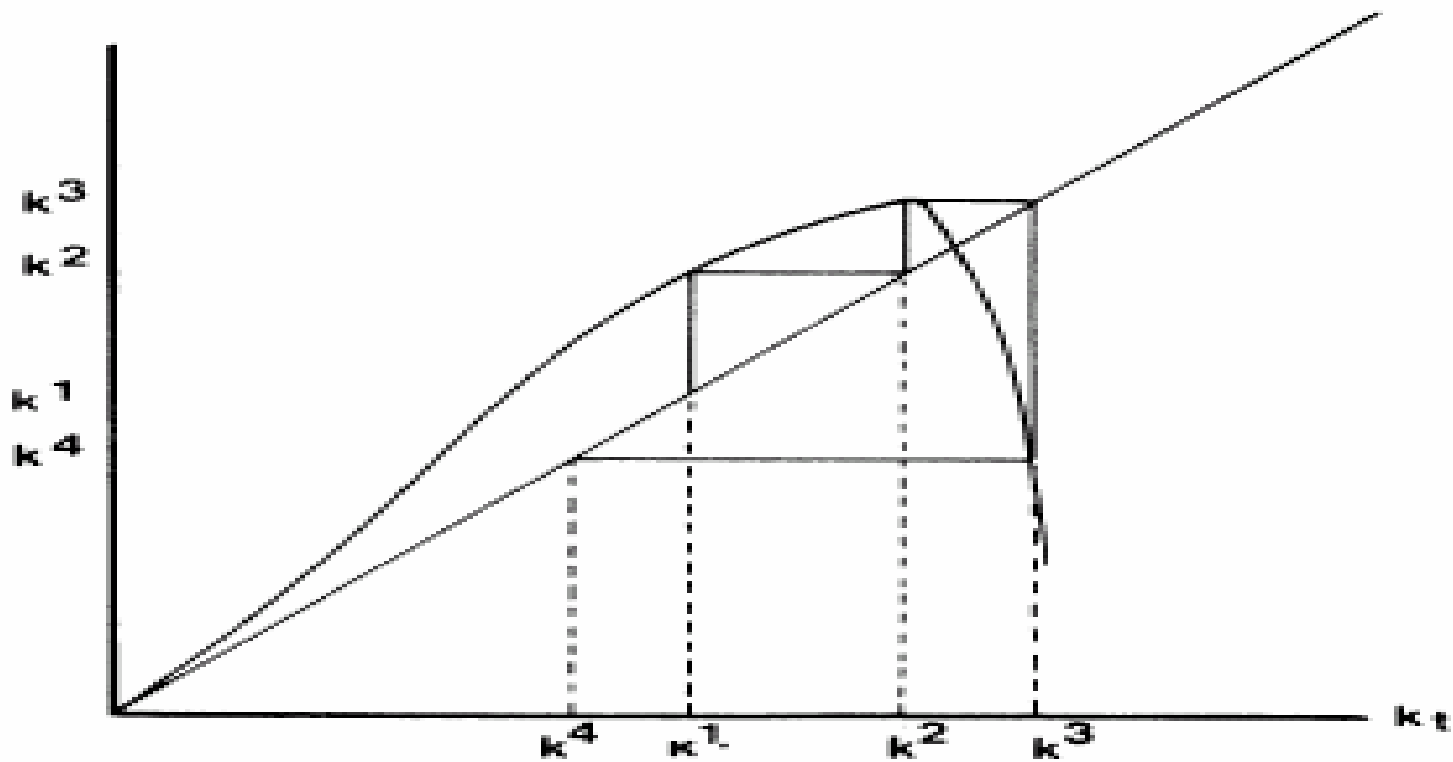
Nonlinear dynamical systems



(c) Stable Two-Period Cycles



Nonlinear dynamical systems



(d) A More Complex Trajectory



Cellular automata

- The Cellular Automaton is an array of cells with a particular geometry (arranged on a line segment, a circle, a segment of the 2-dimensional plane, or a torus).
- In each period of simulation each cell is characterized by a certain state, chosen from a finite set of possibilities.
- The state of each cell in the next period depends on the current state of cells in the system as a whole.
- In most of the CA, the evolution of a single cells depends only on the past states of the cells in a neighborhood containing it.



Cellular automata

A simple case

- The CA might consist of n cells arranged on a line segment.
- Each cell can be in one of two states (on or off).
- The neighborhood of each cell contains one cell on either side of it.
- The current configuration of neighboring cells is described by each of the eight possibilities (000, 001, 010, ..., 111)

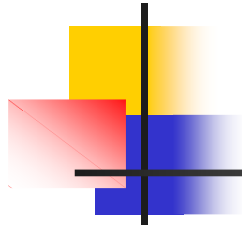


Cellular automata

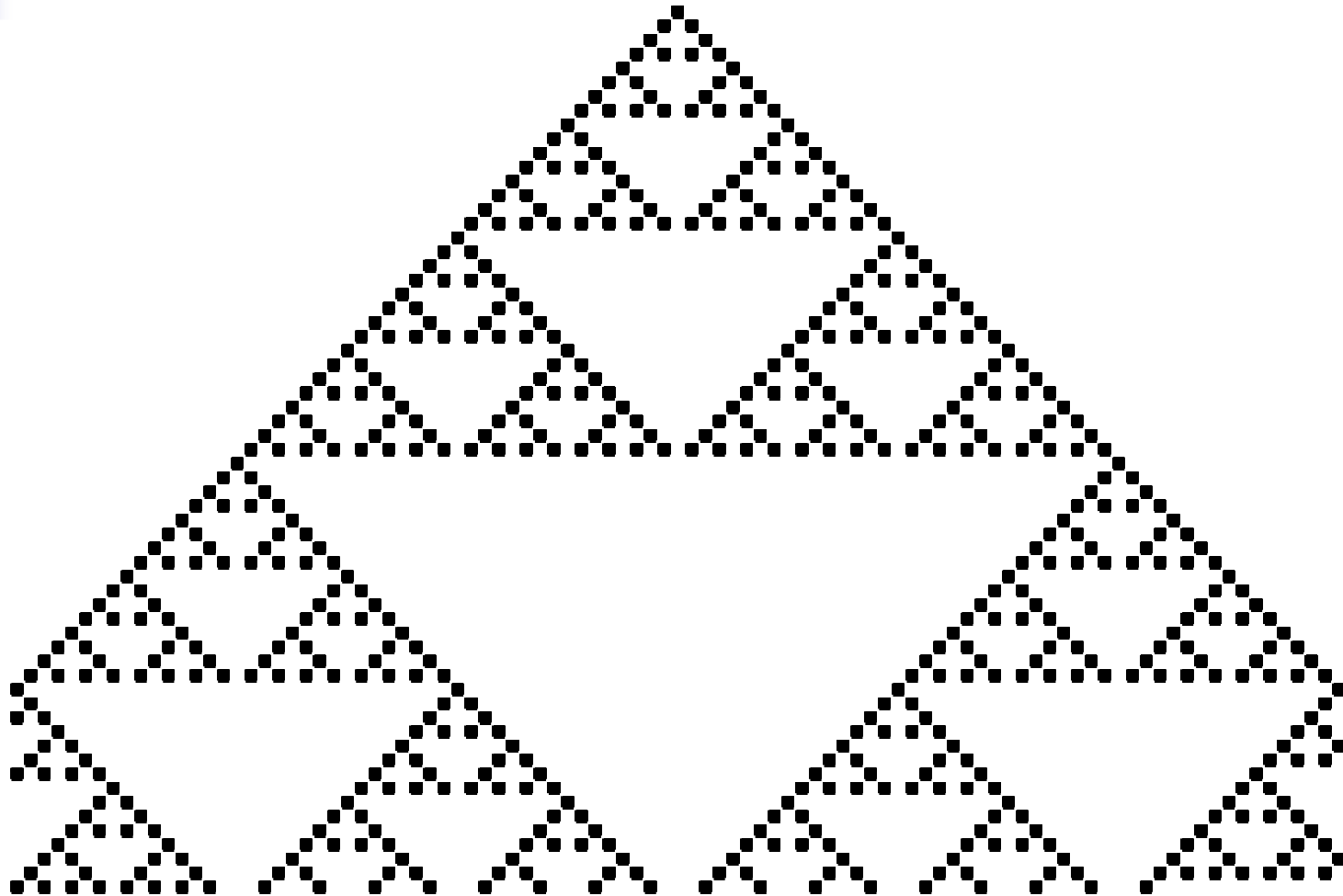
A simple case

- $x_{it+1} = 1$ if ($x_{i-1t} = 1$ and $x_{it} = 1$ and $x_{i+1t} = 0$)
or ($x_{i-1t} = 0$ and $x_{it} = 1$ and $x_{i+1t} = 1$)
or ($x_{i-1t} = 0$ and $x_{it} = 0$ and $x_{i+1t} = 1$)
or ($x_{i-1t} = 1$ and $x_{it} = 0$ and $x_{i+1t} = 0$)

= 0 otherwise



Cellular automata





Cellular automata

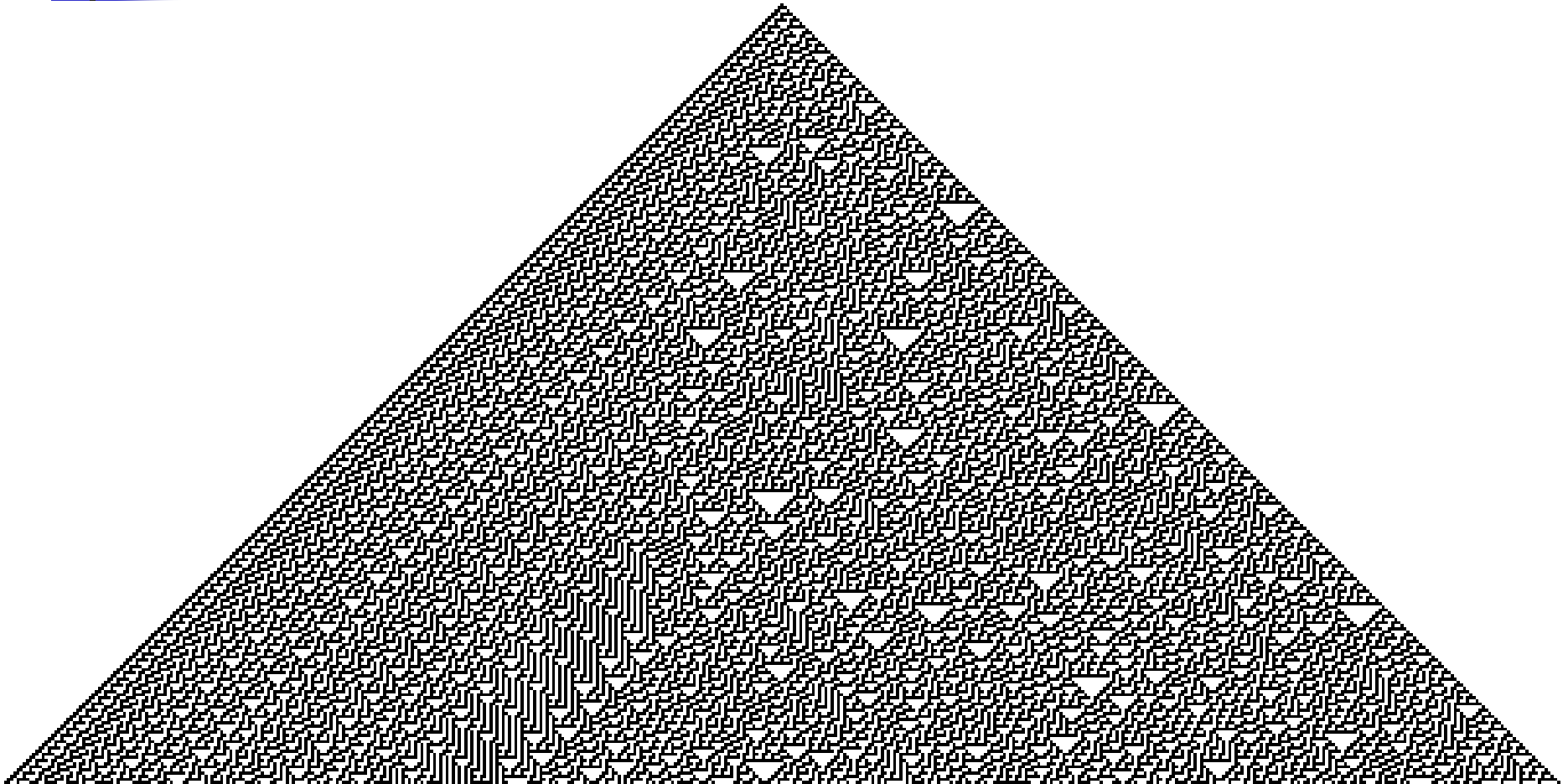
A simple case

- $x_{it+1} = 1$ if ($x_{i-1t} = 0$ and $x_{it} = 1$ and $x_{i+1t} = 0$)
or ($x_{i-1t} = 0$ and $x_{it} = 1$ and $x_{i+1t} = 1$)
or ($x_{i-1t} = 0$ and $x_{it} = 0$ and $x_{i+1t} = 1$)
or ($x_{i-1t} = 1$ and $x_{it} = 0$ and $x_{i+1t} = 0$)

= 0 otherwise



Cellular automata



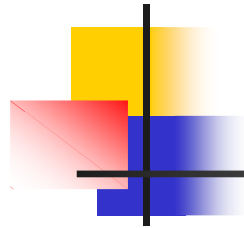


Cellular automata

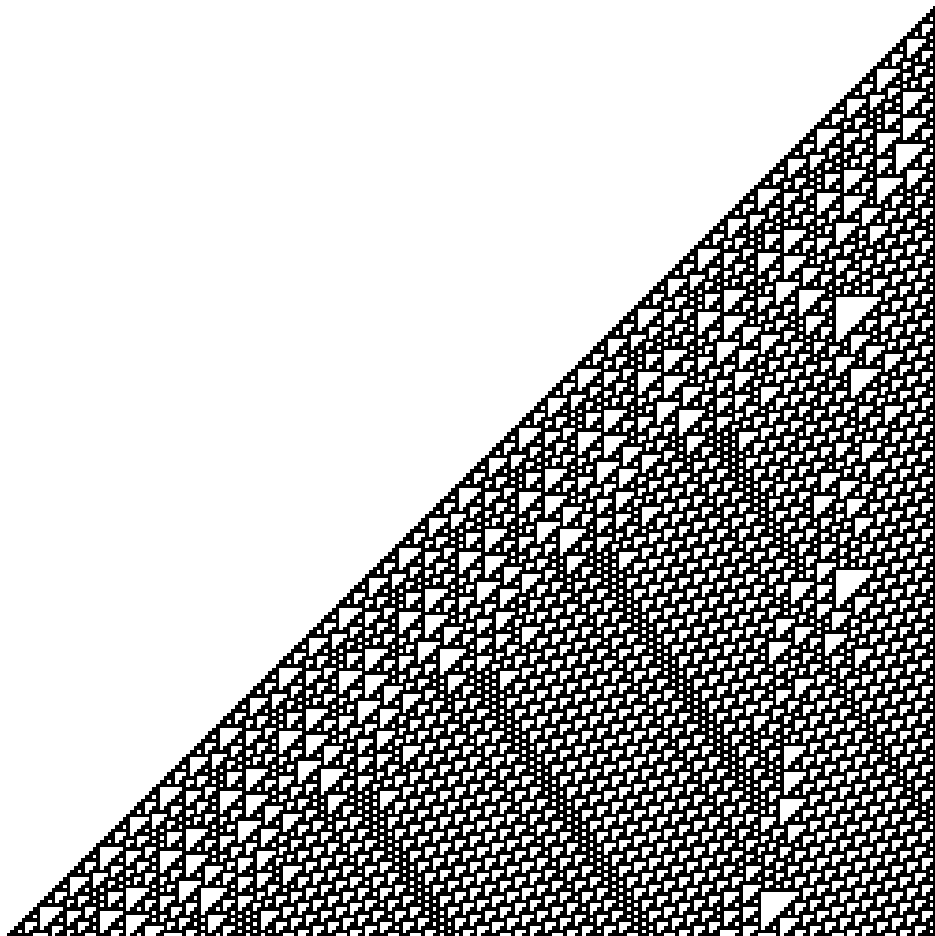
A simple case

- $x_{it+1} = 1$ if ($x_{i-1t} = 1$ and $x_{it} = 1$ and $x_{i+1t} = 0$)
or ($x_{i-1t} = 0$ and $x_{it} = 1$ and $x_{i+1t} = 1$)
or ($x_{i-1t} = 0$ and $x_{it} = 1$ and $x_{i+1t} = 0$)
or ($x_{i-1t} = 1$ and $x_{it} = 0$ and $x_{i+1t} = 1$)
or ($x_{i-1t} = 0$ and $x_{it} = 0$ and $x_{i+1t} = 1$)

= 0 otherwise



Cellular automata





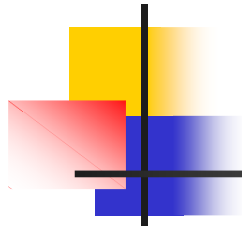
Cellular automata

2-dimensional CA

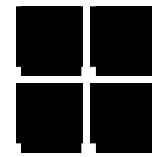
- Game of Life

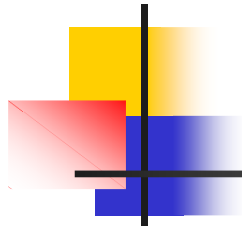
- Rules

- A dead cell with exactly three live neighbors becomes a live cell (birth).
- A live cell with two or three live neighbors stays alive (survival).
- In all other cases, a cell dies or remains dead (overcrowding or loneliness).



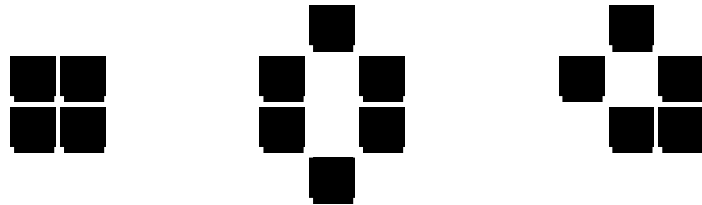
Cellular automata





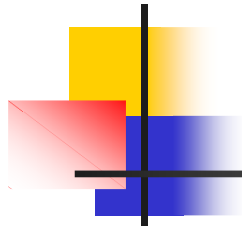
Cellular automata

- Still life objects



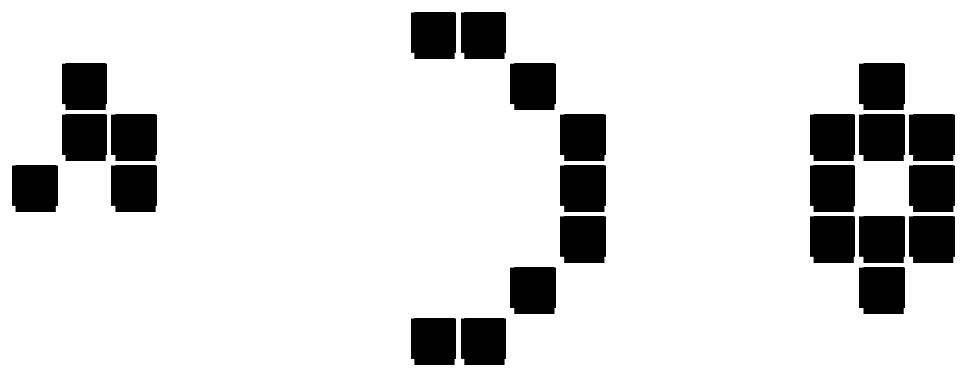
- Oscillators

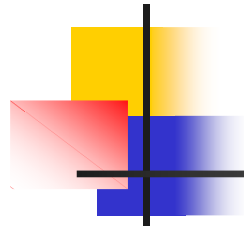




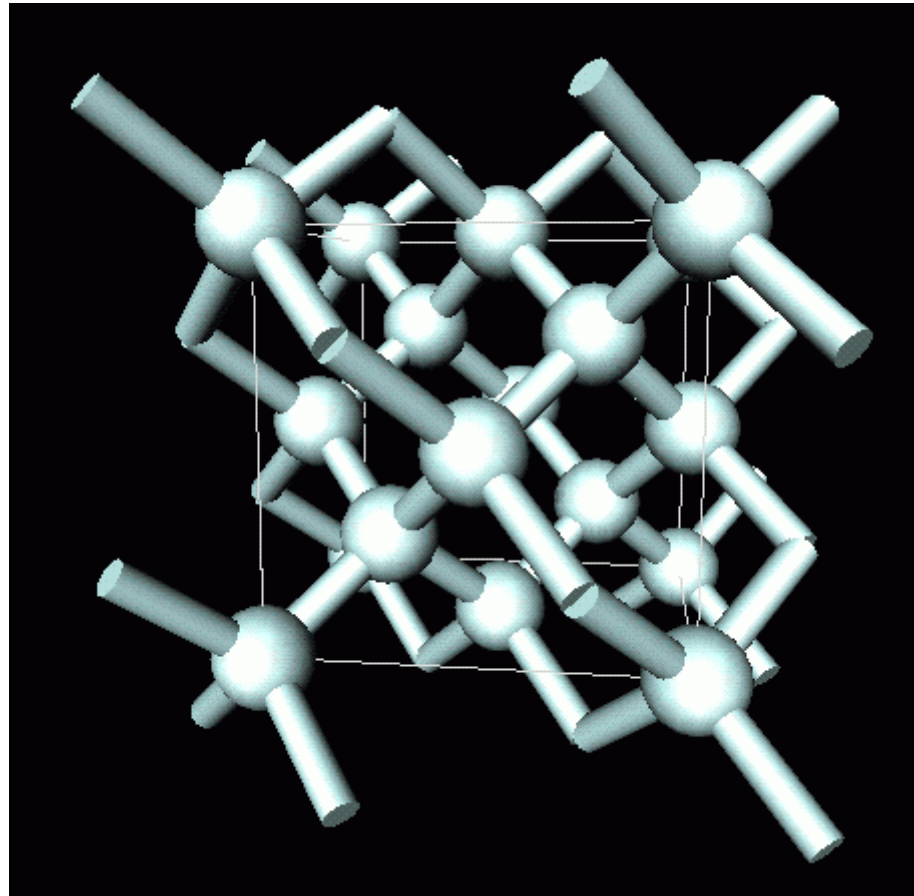
Cellular Automata

- Gliders/Shuttles/Pulsars





Cellular Automata



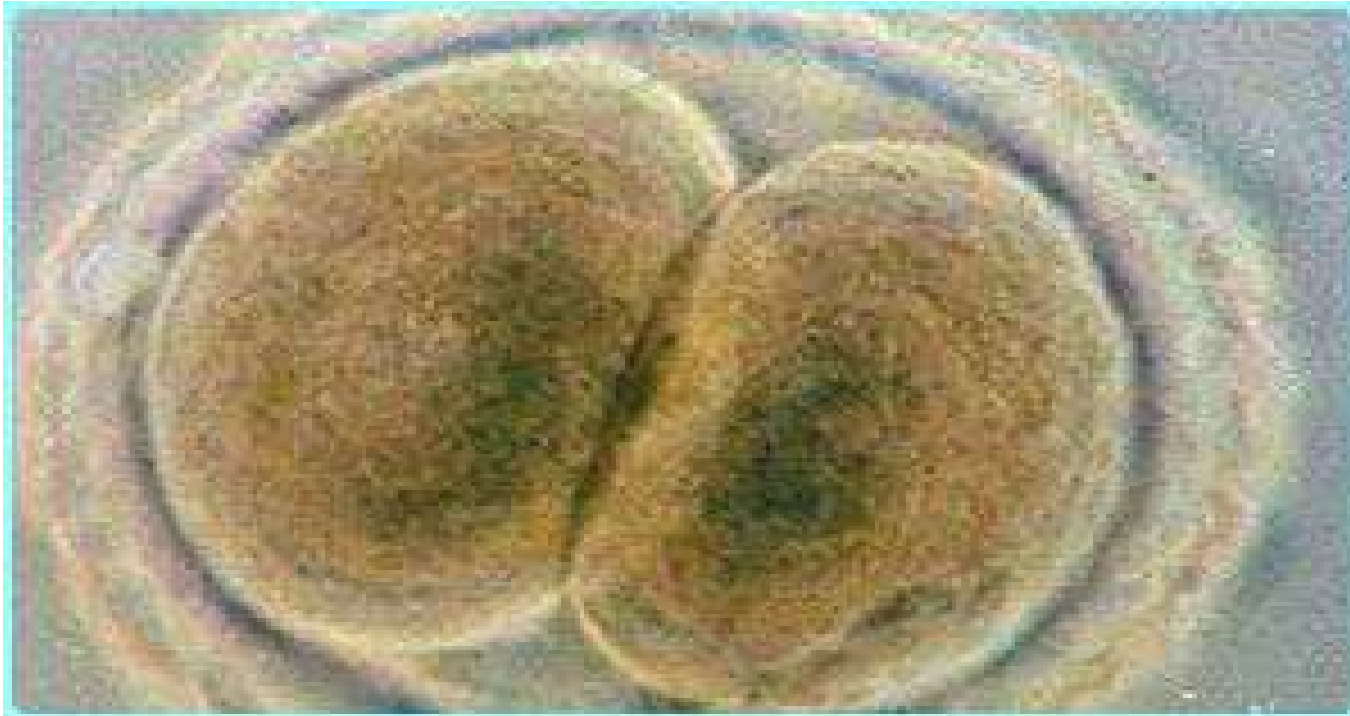


Cellular Automata





Cellular Automata





Cellular Automata

