Complex Systems Workshop Lecture I: Non-linear Dynamics, Chaos, Bifurcation & Strange Attractors

Cars Hommes

CeNDEF, UvA

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Outline

- The 1-D quadratic map
- Bifurcations
- Chaos
- The 2-D Hénon map
- 6 Hopf bifurcation
- 6 Homoclinic orbits
- Lyapunov Exponents
- 8 Implications Nonlinear Dynamics for Economics

Quadratic Map

Example of one-dimensional system:

$$x_{t+1} = f_{\lambda(x_t)} = \lambda x_t (1 - x_t)$$
 (2.2)

- initial state $x_0 \in [0,1]$
- **2** parameter λ , $0 \le \lambda \le 4$.
- Problem: what do the orbits look like?

Convergence to a steady state

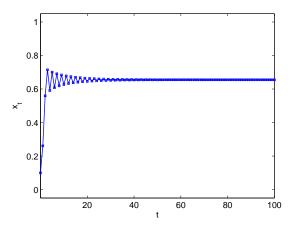


Figure: $\lambda = 2.9$ and $x_0 = 0.1$



Convergence to a 2-cycle

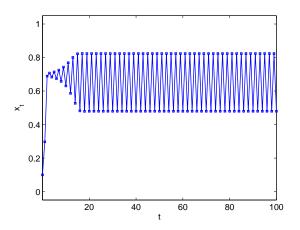


Figure: $\lambda = 3.3$ and $x_0 = 0.1$



Convergence to a 4-cycle

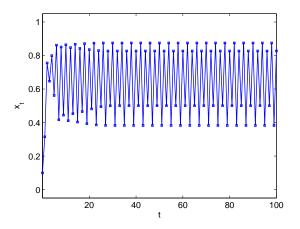


Figure: $\lambda = 3.5$ and $x_0 = 0.1$



Convergence to a 3-cycle

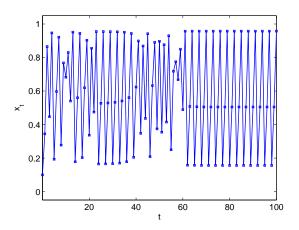


Figure: $\lambda = 3.83$ and $x_0 = 0.1$



Sensitive dependence on initial conditions

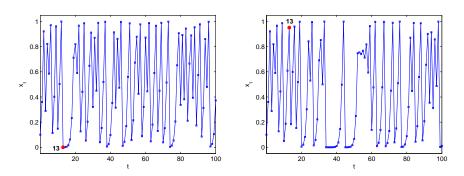


Figure: $\lambda = 4$ and $x_0 = 0.1$ (left) and (f) $\lambda = 4$ and $x_0 = 0.1001$ (right).

Periodic Orbits and Stability

A point x is called a **periodic point with period k** if

$$f^k(x) = x$$
 and $f^i(x) \neq x$, $0 < i < k$.

(Note: periodic point with period k is fixed point of k-th iterate f^k)

$$\{x_1, x_2, ..., x_k\} = \{x_1, f(x_1), f^2(x_1), ..., f^{k-1}(x_1)\}$$
 periodic orbit or k-cycle.

If x_i stable fixed point of f^k , then $\{x_1, x_2, ..., x_k\}$ stable periodic orbit; from the chain rule we have

$$(f^{k})'(x_{i}) = (f^{k})'(x_{1}) = f'(f^{k-1}(x_{1})) \cdot f'(f^{k-2}(x_{1})) \dots f'(f(x_{1})) \cdot f'(x_{1})$$
$$= \prod_{i=0}^{k-1} f'(f^{i}(x_{1})).$$

(Note: $(f^k)'(x_j)$ is the product of derivatives along the orbit)

Aperiodic Point

A point x is called an **aperiodic point** if

- the orbit of x is bounded,
- the orbit of x is not periodic, and
- 1 the orbit of x does not converge to a periodic orbit

Bifurcation diagram of the quadratic map

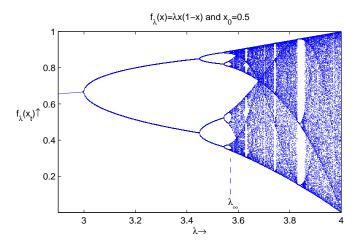


Figure: Bifurcation diagram of the quadratic map.



Period doubling bifurcation at $\lambda = 3$

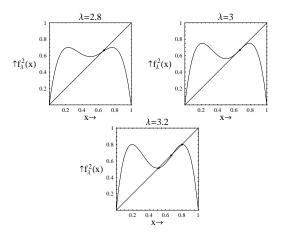


Figure: Graphs of the second iterate f^2 for three different λ -values close to the period-doubling bifurcation at $\lambda = 3$.



Tangent bifurcation for $x_{t+1} = x_t^2 + c$ at c = 1/4

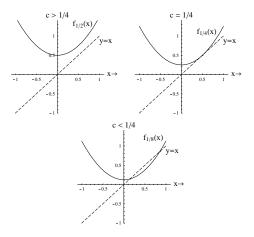


Figure: Tangent bifurcation for $x_{t+1} = x_t^2 + c$ at c = 1/4.



Creation of a 3-cycle by tangent bifurcation

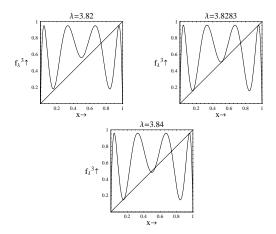


Figure: Creation of 3-cycle by tangent bifurcation at $\lambda \approx 3.8283$.



Tangent bifurcation of a 3-cycle in the quadratic map

Proposition 2.1

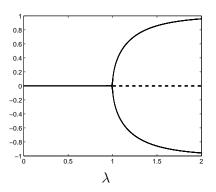
For $\lambda=\lambda^*\approx 3.8283$ f_λ has a tangent bifurcation in which two 3-cycles are created, one stable and one unstable. Equivalently, at $\lambda=\lambda^*$ the third iterate f_λ^3 has a tangent bifurcation in which simultaneously 6 steady states are created, 3 stable and 3 unstable. We have

- (1) for $\lambda < \lambda^*$: f_{λ} has no 3-cycle,
- (2) for $\lambda = \lambda^*$: f_{λ} has one 3-cycle $\{x_1, x_2, x_3\}$, and $(f^3)'(x_i) = +1$, for $1 \le i \le 3$,
- (3) for $\lambda > \lambda^*$ (and λ close to λ^*): f_{λ} has two 3-cycles, one stable and one unstable.

Pitchfork Bifurcation

Example: symmetric S-shaped map

$$x_{t+1} = \frac{e^{\lambda x_t} - e^{-\lambda x_t}}{e^{\lambda x_t} + e^{-\lambda x_t}}$$



Definition of topoligical chaos

The dynamics of a difference equation $x_{t+1} = f(x_t)$ is called **(topologically) chaotic** if the following three properties are satisfied:

- There exists an infinite set *P* of (unstable) periodic points with different periods.
- There exists an uncountable set S of aperiodic points (i.e. poinst whose orbits are bounded, not periodic and not converging to a periodic orbit).
- **3** *f* has sensitive dependence on initial conditions w.r.t. Λ = P ∪ S, that is, there exists a positive distance C such that for all initial states $x_0 ∈ Λ$ and any ε-neighbourhood U of x_0 , there exists an initial state $y_0 ∈ Λ ∩ U$ and a time T > 0 such that the distance $d(x_T, y_T) = d(f^T(x_0), f^T(y_0)) > C$.



Example of Chaos: quadratic map for $\lambda = 4$

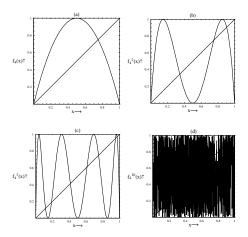


Figure: Graphs of (a) $f_4(x) = 4x(1-x)$, (b) f_4^2 , (c) f_4^3 and (d) f_4^{10} .



Properties of quadratic map f_4

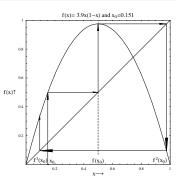
It can be shown that for any n, the graph of f_4^n has the following properties:

- f_4^n has 2^{n-1} maxima equal to 1 and $2^{n-1} + 1$ minima equal to 0 (including minima at x = 0 and x = 1).
- 2 f_4^n 'oscillates' 2^{n-1} times on the interval [0,1].
- **1** the map f_4^n has 2^n fixed points.
- **9** for any interval I of arbitrarily small length ε , there exists an N > 0 such that I contains points x, y with $f_4^N(x) = 0$ and $f_4^N(y) = 1$.

Period three implies chaos

Theorem 1

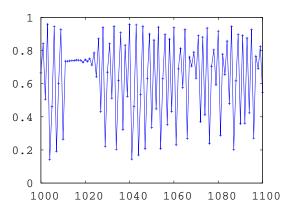
("Period 3 implies Chaos", Li & Yorke [1975]). Let $x_{t+1} = f(x_t)$ be a 1-D difference equation with f a continuous map. If there exist a point x_0 such that $f^3(x_0) \le x_0 < f(x_0) < f^2(x_0)$ (or with > instead of <) then the dynamics is topologically chaotic



Topological Chaos with Noise

Quadratic map with small noise

$$x_{t+1} = 3.83x_t(1-x_t)$$



Definition of true chaos

The dynamics of a difference equation $x_{t+1} = f(x_t)$ is called **'truly' chaotic** if there exists a set Λ of positive Lebesgue measure, such that f has sensitive dependence on initial conditions w.r.t. Λ , that is, there exists a positive distance C such that for all initial states $x_0 \in \Lambda$ and any ε -neighbourhood U of x_0 , there exists an initial state $y_0 \in \Lambda \cap U$ and a time T > 0 such that the distance $d(x_T, y_T) = d(f^T(x_0), f^T(y_0)) > C$.

Lyapunov exponents

- The **Lyapunov exponent** is defined as $\lambda(x_0) = \lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} \ln(|f'(f^i(x_0))|).$
- Derivation:

$$|f^{n}(x_{0} + \delta) - f^{n}(x_{0})| \approx |(f^{n})'(x_{0})\delta| = e^{n\lambda(x_{0})} |\delta|$$

 $\Leftrightarrow e^{n\lambda(x_{0})} = |(f^{n})'(x_{0})| \Rightarrow \lambda(x_{0}) = \frac{1}{n} \ln(|(f^{n})'(x_{0})|).$

• The Lyapunov exponent measures the average rate of divergence of nearby initial states. It is the average of the logs of the absolute values of the derivative along the orbit.



Lyapunov exponent plot of the quadratic map

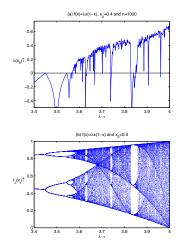


Figure: Lyapunov exponent L as a function of the parameter λ .

The asymmetric tent map

The asymmetric tent map T_{β} is the continuous, piecewise linear map $T_{\beta}:[0,1]\to[0,1]$ defined as

$$T_{\beta}(x) = \begin{cases} \frac{2}{1+\beta}x, & 0 \le x \le \frac{(\beta+1)}{2} \\ \frac{2}{1-\beta}(1-x), & \frac{\beta+1}{2} < x \le 1, \end{cases}$$
 (1)

where the parameter $-1 < \beta < +1$.

Graph of asymmetric tent maps

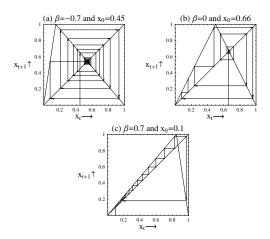


Figure: Graphs of the asymmetric tent map: (a) $\beta=-0.7$, (b) $\beta=0$ and (c) $\beta=0.7$

The properties of the asymmetric tent map

The piecewise linear difference equation $x_{t+1} = T_{\beta}(x_t)$ has the following properties:

- For any integer $j \ge 1$, T_{β} has a periodic point of period j; all periodic orbits are unstable.
- ② For Lebesgue almost all initial states $x_0 \in [0, 1]$, the time path $\{x_t\}_{t=0}^{\infty}$ is chaotic and dense in the interval [0, 1].
- **③** For Lebesgue almost all initial states $x_0 \in [0,1]$, the sample average of the (chaotic) time path is $\bar{x} = \lim_{T \to \infty} \frac{1}{T+1} \sum_{t=0}^{T} x_t = 1/2$.
- **4** For Lebesgue almost all initial states $x_0 \in [0, 1]$, the sample autocorrelation coefficient at lag j is $\rho_i = \beta^j$.

Two-dimensional (2-D) systems

$$(x_{t+1}, y_{t+1}) = F_{\lambda}(x_t, y_t),$$

 F_{λ} nonlinear 2-D map and λ is a parameter.

The *orbit* with *initial state* (x_0, y_0) is the set

$$\{(x_0, y_0), (x_1, y_1), (x_2, y_2), \ldots\} = \{(x_0, y_0), F_{\lambda}(x_0, y_0), F_{\lambda}^2(x_0, y_0), \ldots\}.$$

Problem: what do these orbits look like and how does it depend on initial states and parameters?

Example: Hénon map:

$$x_{t+1} = 1 - ax_t^2 + y_t$$

 $y_{t+1} = bx_t$,

where a and b are parameters.

(special case b = 0 yields 1-D quadratic map)



Strange attractor of the Hénon map

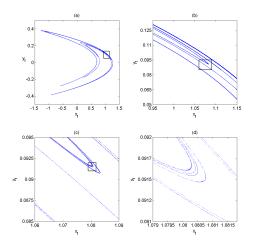


Figure: The strange attractor for the Hénon map $H_{a,b}$ with a=1.4 and b=0.3.

Chaotic time-series and SDIC

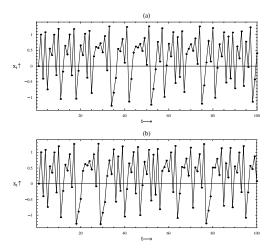


Figure: Chaotic time series and sensitive dependence for the Hénon map $H_{a,b}$ with a=1.4 and b=0.3. (a) $(x_0,y_0)=(0,0)$ and (b) $(x_0,y_0)=(0.001,0)$.

Attractor and Strange Attractor

An **attractor** of a K-dimensional system $X_{t+1} = F(X_t)$ is a compact set A with the following properties:

- **1** The set A is invariant, i.e. $F(A) \subset A$.
- ② There exists an open neighborhood U of A (i.e. $A \subset U$), such that all initial states X_0 converge to the attractor A, i.e. for all $X_0 \in U$, $\lim_{n\to\infty} dist(F^n(X_0), A) = 0$.
- **3** There exists an initial state $X_0 \in A$ for which the orbit is dense in A.

An attractor A is called a **strange attractor** of the *N*-dimensional dynamical system $x_{t+1} = F(x_t)$, if the map F has sensitive dependence w.r.t. the set of initial states converging to A.

The delayed logistic map

- Delayed logistic map: $N_{t+1} = aN_t(1 N_{t-1})$.
- Equivalently $(x_t = N_t \text{ and } y_t = N_{t-1})$:

$$\begin{array}{rcl}
 x_{t+1} & = & y_t \\
 y_{t+1} & = & ay_t(1-x_t).
 \end{array}$$

steady states

$$(x_1, y_1) = (0, 0)$$
 and $(x_2, y_2) = (\frac{a-1}{a}, \frac{a-1}{a}).$

• The eigenvalues of the system are $\lambda_1 = \frac{1}{2} - \frac{1}{2}\sqrt{5-4a}$ and $\lambda_2 = \frac{1}{2} + \frac{1}{2}\sqrt{5-4a}$.



Dynamical properties of the delayed logistic map

The eigenvalues λ_1 and λ_2 of $JF_a(\frac{a-1}{a},\frac{a-1}{a})$ satisfy the following properties:

- $0 \le a < 1$: real eigenvalues with $-1 < \lambda_1 < 1 < \lambda_2$, so $(\frac{a-1}{a}, \frac{a-1}{a})$ is a **saddle**.
- $1 < a < \frac{5}{4}$: real eigenvalues with $0 < \lambda_1 < \lambda_2 < 1$, so $(\frac{a-1}{a}, \frac{a-1}{a})$ is attracting (**stable node**).
- $\frac{5}{4} < a < 2$: complex eigenvalues with $\lambda_1 \lambda_2 = a 1 < 1$, so $\left(\frac{a-1}{a}, \frac{a-1}{a}\right)$ is a **stable focus**.
- a > 2: complex eigenvalues with $\lambda_1 \lambda_2 = a 1 > 1$, so $(\frac{a-1}{a}, \frac{a-1}{a})$ is an **unstable focus**.
- Hopf bifurcation (or Neimark-Sacker) for a = 2 complex eigenvalues on the unit circle



Attractors delayed logistic map

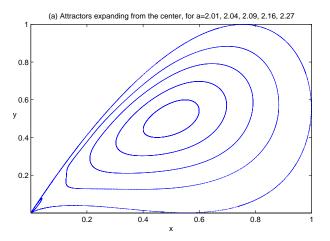


Figure: (a) Attractors for the logistic delayed equation for different a-values after the Hopf bifurcation.

Strange attractor of the delayed logisitic map

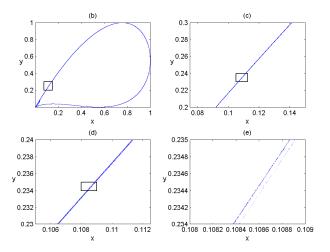


Figure: (b-e) The strange attractor for a = 2.27 and some enlargements.

Time series for the logistic delayed equation

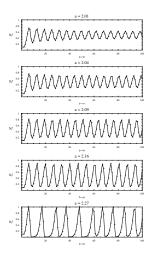


Figure: Time series for for different values of the parameter a.



Bifurcation diagrams for logistic delayed equation

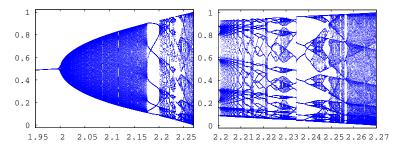


Figure: Hopf-bifurcation and breaking of an invariant circle bifurcation route to chaos

Types of Local Bifurcations

- **1** $\lambda = +1$:
 - saddle-node bifurcation: two new steady states, a saddle and a node;
 - pitchfork bifurcation: one steady state becomes unstable and two new stable steady states;
 - transcritical bifurcation; two steady states collide and exchange stability;
- **2** $\lambda = -1$:
 - period-doubling bifurcation: steady state loses stability and new stable 2-cycle;
- ① a pair of complex eigenvalues λ_1 and λ_2 on the unit circle, i.e. $|\lambda_1\lambda_2|=1$:
 - Hopf bifurcation: steady state becomes an unstable focus and an attracting invariant circle emerges with (quasi-)periodic dynamics



Local (un)stable manifolds

Let p be a fixed point of the 2-D map F.

The local stable manifold and local unstable manifold of p are defined as

$$W_{loc}^{s}(p) = \{ x \in U | \lim_{n \to \infty} F^{n}(x) = p \}$$
 (2)

$$W_{loc}^{u}(p) = \{x \in U | \lim_{n \to -\infty} F^{n}(x) = p\}$$
(3)

where U is some small neighbourhood of p.



Global(un)stable manifolds

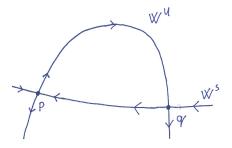
The global stable manifold and the global unstable manifold are now defined as

$$W^{s}(p) = \bigcup_{n=0}^{\infty} F^{-n}(W_{loc}^{s})$$

$$\tag{4}$$

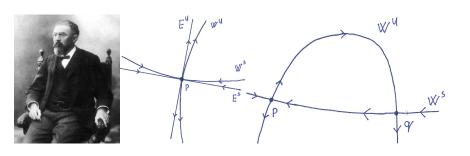
$$W^{u}(p) = \bigcup_{n=0}^{\infty} F^{n}(W_{loc}^{u}). \tag{5}$$

Homoclonic point



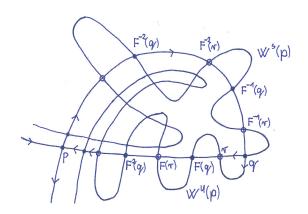
A point q is called a **homoclinic point** if $q \neq p$ and q is an intersection point of the stable and unstable manifolds of the saddle point p, that is, $q \in W^s(p) \cap W^u(p)$.

Henry Poincaré, ca. 1890: motion in the **three body problem** is unpredictable and chaotic



2-D map of suitable plane section has homoclinic orbit

Henry Poincaré, ca. 1890: homoclinic orbit implies sensitive dependence on initial conditions

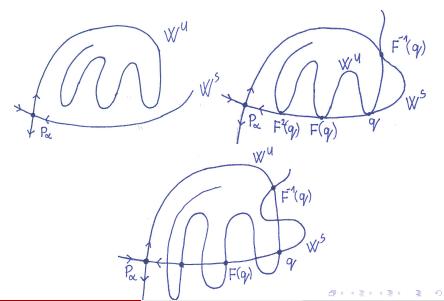


Homoclonic bifurcation

We say that F_{α} has a **homoclinic bifurcation**, associated to the saddle point p_{α} , at $\alpha = \alpha_0$, if

- for $\alpha < \alpha_0$, $W^s(p_\alpha)$ and $W^u(p_\alpha)$ have no intersection point $q \neq p$;
- ② for $\alpha = \alpha_0$, $W^s(p_\alpha)$ and $W^u(p_\alpha)$ have a point of homoclinic tangency;
- **3** for $\alpha > \alpha_0$, $W^s(p_\alpha)$ and $W^u(p_\alpha)$ have a transversal homoclinic intersection point.

Homoclonic bifurcation



Lyapunov exponents

- Consider a dynamic model $x_{t+1} = F(x_t)$, where F is an n-dimensional map.
- Let x_0 be an intial state vector and δ an initial perturbation vector.
- After n time periods, the separation between the two initial state vectors x_0 and $x_0 + \delta$ is approximately

$$|| F^{n}(x_{0} + \delta) - F^{n}(x_{0}) || \approx || (D_{x_{0}}F^{n})(\delta) ||$$
.

 The Lyapunov exponent is a measure of the average exponential rate of divergence:

$$\lambda(x_0,\delta) = \lim_{n\to\infty} \frac{1}{n} ln(\parallel (D_{x_0}F^n)(\delta) \parallel).$$



Lyapunov exponents λ_1 and λ_2

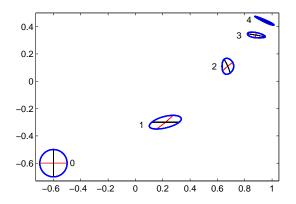


Figure: Lyapunov exponents λ_1 and λ_2

Henry Poincaré to Léon Walras (30 September, 1901):

"You regard men as infinitely selfish and infinitely faresighted.

The first hypothesis can perhaps be admitted as a first approximation, but the second should perhaps be regarded with some reservations"



nonlinear complex systems: limits to predictability and rationality

Léon Walras to Henry Poincaré (3 October, 1901):



In reality, agents are neither infinitely selfish nor infinitely clairvoyant. Theory should indicate these frictions carefully ...

Implications Nonlinear Dynamics for Economics

- The fact that (simple) nonlinear systems exhibit complex dynamics calls for reservations about rational behavior, in particular rational expectations;
- In a **nonlinear** world, simple **heuristics** that work reasonably well may be the best **boundedly rational** agents can achieve