

Appendix C

(Stability analysis in discrete dynamical systems.)

Consider the one-dimensional discrete dynamical system

$$\textcircled{1} \quad n_{t+1} = f(n_t), \quad t = 0, 1, 2, \dots$$

for given cont. diff. $f: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ and suppose that

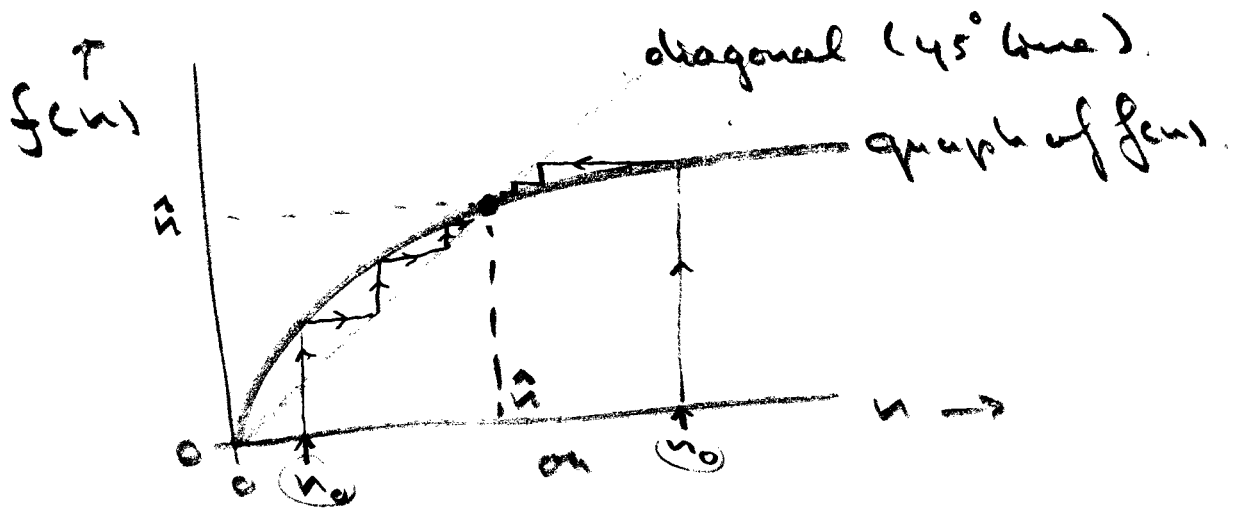
$$\textcircled{2} \quad \hat{n} = f(\hat{n})$$

(i.e., \hat{n} is an equilibrium of $\textcircled{1}$).

Question: Is \hat{n} stable in the sense that $n_t \rightarrow \hat{n}$ as $t \rightarrow \infty$ for initial values n_0 in some open set?

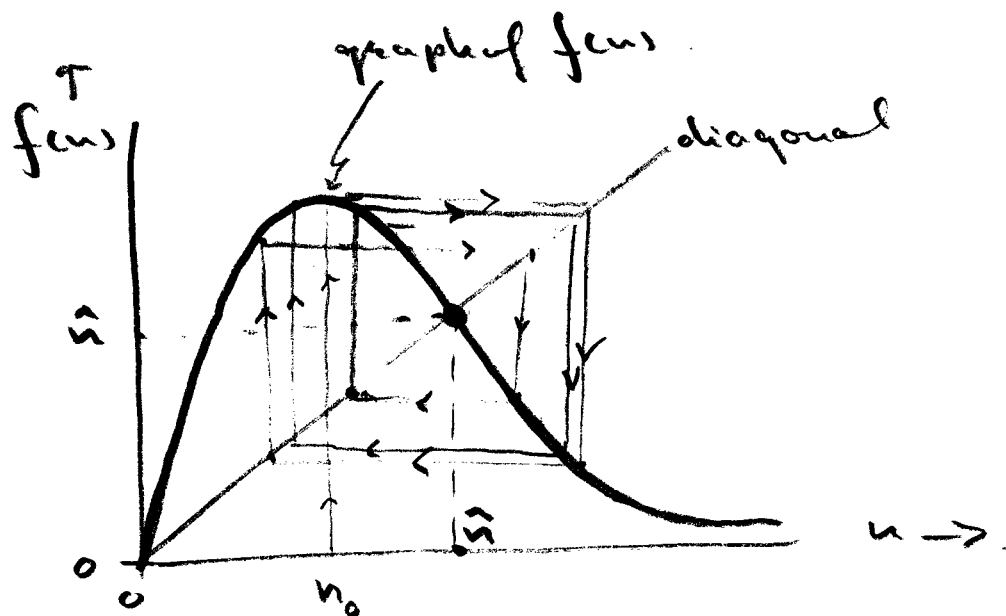
First answer: "Cobweb method"

Ex 1



conclusion: \hat{x} is stable.

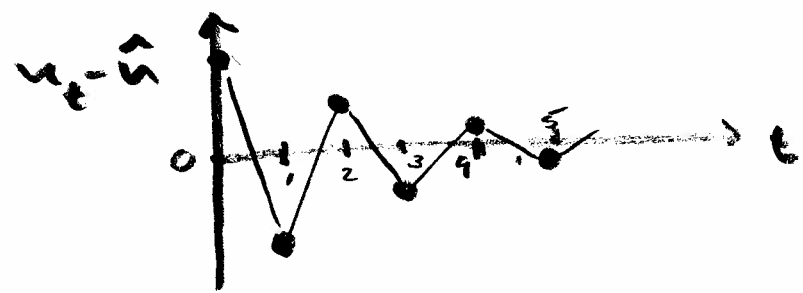
Ex 2



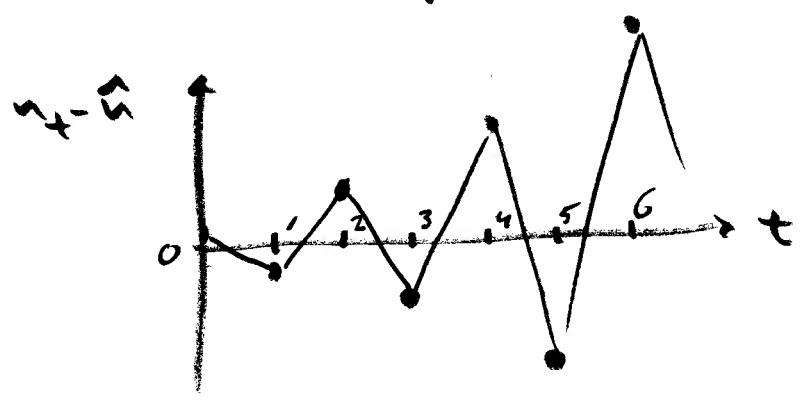
conclusion: \hat{x} does not to be stable.

The cobweb method gives global info about stability (or lack thereof), but is not always clear.

$| -1 < f'(\hat{u}) < 0 | \Rightarrow$ non-monotonous convergence



$| f'(\hat{u}) < -1 | \Rightarrow$ non-monotone divergence



In short: \hat{u} in the linearized system (4) is locally stable if $|f'(\hat{u})| < 1$ and unstable if $|f'(\hat{u})| > 1$.

The theorem of Hartman & Grobman (see Appendix A) says that if $|f'(\hat{u})| \neq 0$ and $|f'(\hat{u})| \neq 1$, then the non-linear system (1) and the linearized system (4) are locally topologically equivalent.

The local stability analysis directly generalizes to multi-dimensional systems.

$$\textcircled{6} \quad v_{t+1} = f(v_t) \quad , \quad t = 0, 1, 2, \dots$$

with $f: \mathbb{R}_+^k \rightarrow \mathbb{R}_+^k$, cont. diff. and

$$\textcircled{7} \quad \hat{v} = f(\hat{v})$$

Linearization of $\textcircled{6}$ about the equilibrium \hat{v} gives

$$\textcircled{8} \quad v_{t+1} - \hat{v} = f'(\hat{v})(v_t - \hat{v})$$

where $f'(\hat{v})$ is a $k \times k$ matrix of partial derivatives (Jacobi-matrix).

Again, Hartman-Grobman says that if none of the eigenvalues of $f'(\hat{v})$ is zero or lies on the unit circle in the complex plane, then the systems $\textcircled{6}$ and $\textcircled{8}$ are topologically equivalent.

5

This motivates us to investigate the linear system

$$(9) \quad \boxed{x_{t+1} = Ax_t \in \mathbb{R}^k} \quad (x=0 \text{ is equil})$$

General solution:

$$(10) \quad x_t = A^t x_0$$

but what does this say about the stability of $x=0$?

Let

$$Ab_i = \lambda_i b_i \quad (i=1, \dots, k)$$

where b_i is an eigenvector of A and λ_i the corresponding eigenvalue.

Let further

$$B := (b_1, \dots, b_k) \quad (\text{matrix})$$

and

$$\Lambda := \text{diagonal}(\lambda_1, \dots, \lambda_k).$$

(c.f. Appendix A)

Then

$$AB = BA.$$

Suppose further that the eigenvectors b_1, \dots, b_n are linearly independent.

Then B^{-1} exists and

$$A = B \Lambda B^{-1}$$

Hence, t times

$$\begin{aligned} A^t &= \underbrace{B \Lambda B^{-1} \cdot B \Lambda B^{-1} \cdot \dots \cdot B \Lambda B^{-1}}_{t \text{ times}} \cdot B \Lambda B^{-1} = \\ &= B \Lambda^t B^{-1} = B \begin{pmatrix} \lambda_1^t & & 0 \\ & \ddots & \\ 0 & & \lambda_n^t \end{pmatrix} B^{-1} \end{aligned}$$

Subst into (10) gives

$$(11) \quad x_t = B \begin{pmatrix} \lambda_1^t & & 0 \\ & \ddots & \\ 0 & & \lambda_n^t \end{pmatrix} B^{-1} x_0$$

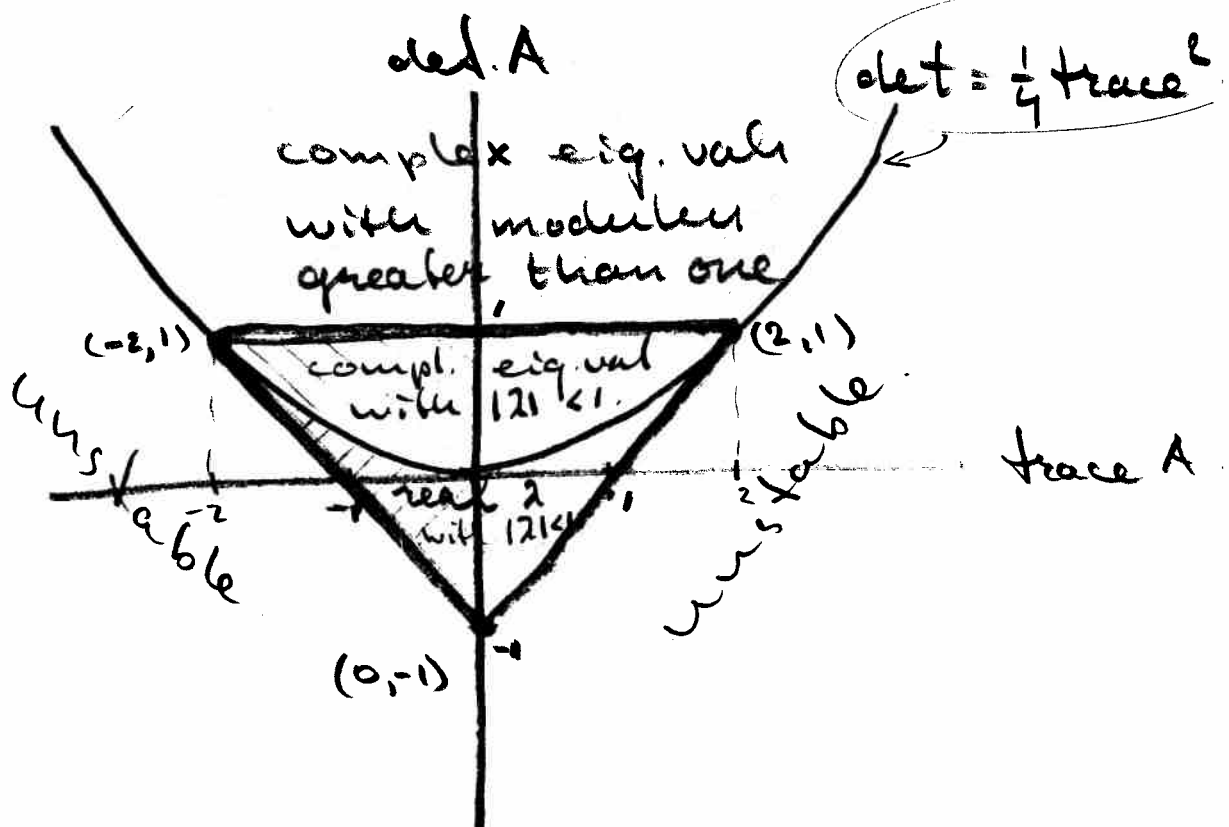
It immediately follows that if $|\lambda_i| < 1$ for all i , then $x=0$ is stable. However, if there exists even a single i such that $|\lambda_i| > 1$, then $x=0$ is unstable.

Planar systems.

Suppose $A \in \mathbb{R}^{2 \times 2}$

Then the eigenvalues

λ_1 and λ_2 are fully determined by $\det A$ and $\text{trace } A$ in the following way



Only for $(\text{trace } A, \det A)$ inside the triangle $(-2, 1), (2, 1), (0, -1)$ the $x=0$ is stable.

Example

(Model of Nicholson-Bailey)

$n_{t+1} = \alpha n_t e^{-\beta p_t}$	(host)
$p_{t+1} = n_t (1 - e^{-\beta p_t})$	(parasitoid)

Equilibrium

$n_{t+1} = n_t = \hat{n} > 0$ & $p_{t+1} = p_t = \hat{p} > 0$

\Leftrightarrow $\hat{n} = \frac{\alpha \log \alpha}{(\alpha - 1) \beta}$ & $\hat{p} = \frac{\log \alpha}{\beta}$

provided $\alpha > 1$.

Jacobi-matrix

$A := \begin{pmatrix} 1 & -\frac{\alpha \log \alpha}{\alpha - 1} \\ 1 - \frac{1}{\alpha} & \frac{\log \alpha}{\alpha - 1} \end{pmatrix}$

obvious
($\alpha > 1$)

trace $A = 1 + \frac{\log \alpha}{\alpha - 1} > 1$

det $A = \frac{\alpha \log \alpha}{\alpha - 1} > 1$

whenever
 $\alpha > 1$

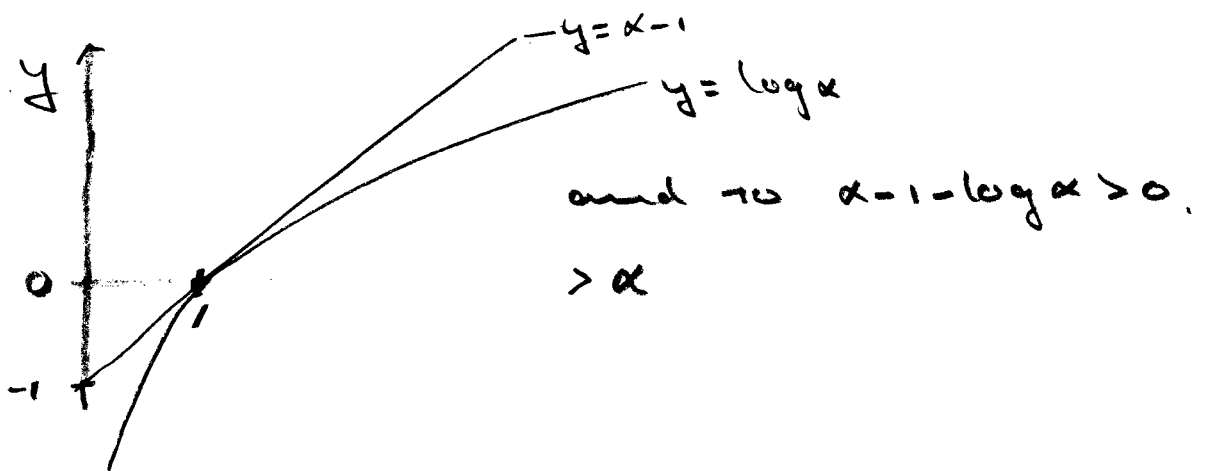
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Since $\det A > 1$, (\hat{u}, \hat{p}) is unstable (see fig. on page 7)

To see that $\det A$ is greater than one, define

$$g(\alpha) := \det A = \frac{\alpha \log \alpha}{\alpha - 1}$$

$$\Rightarrow g'(\alpha) = \frac{\alpha - 1 - \log \alpha}{(\alpha - 1)^2} > 0$$



$$\Rightarrow g(\alpha) > \lim_{\alpha \downarrow 1} g(\alpha) = 1.$$
