

# Equilibria and isoclines

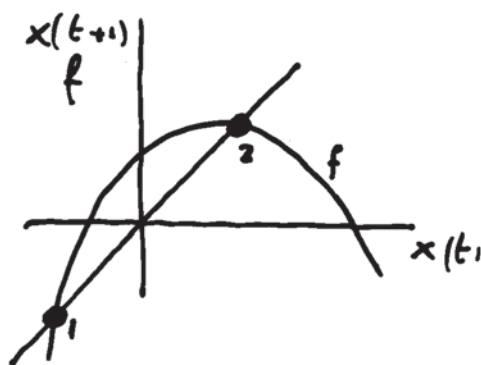
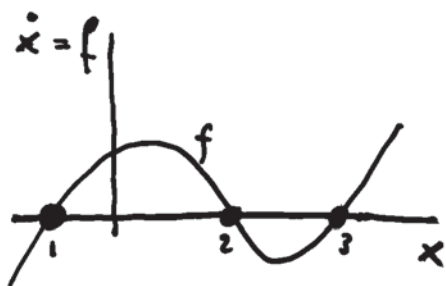
## Definition 1 (equilibrium)

An equilibrium is a state  $\bar{x}$  such that  $x(0) = \bar{x}$  implies  $x(t) = \bar{x} \quad \forall t \geq 0$ .

## Consequence

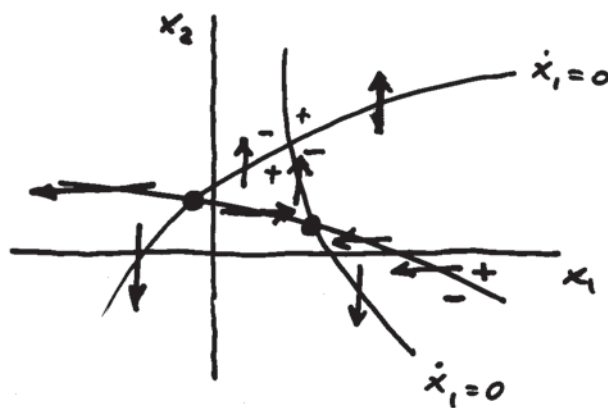
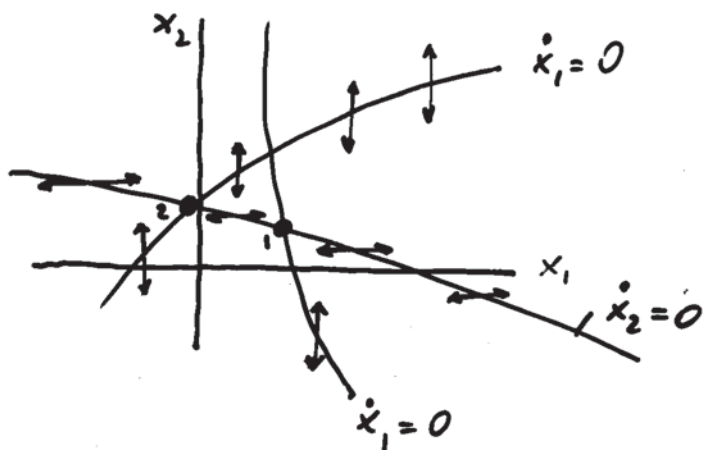
In continuous-time systems  $\dot{x} = f(x)$  equilibria  $\bar{x}$  are such that  $f(\bar{x}) = 0$ , while in discrete-time systems  $x(t+1) = f(x(t))$  equilibria  $\bar{x}$  are such that  $\bar{x} = f(\bar{x})$ .

## Ex. 1 1-st order systems



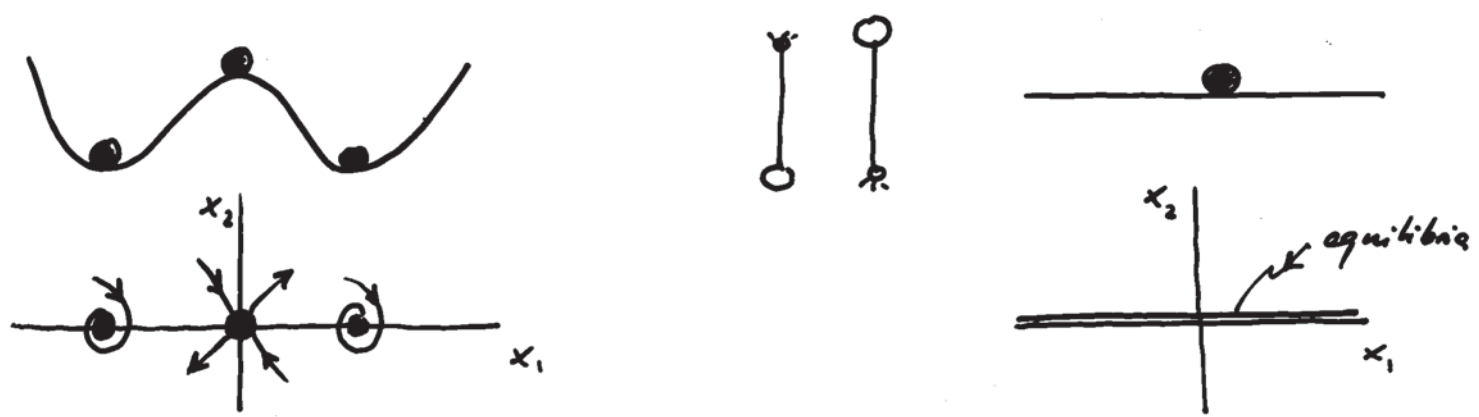
## Ex. 2 2-nd order continuous time systems

$$\begin{aligned} \dot{x}_1 &= f_1(x_1, x_2) & \Rightarrow & \dot{x}_1 = 0 & f_1(x_1, x_2) = 0 \\ \dot{x}_2 &= f_2(x_1, x_2) & \Rightarrow & \dot{x}_2 = 0 & f_2(x_1, x_2) = 0 \end{aligned} \quad \left. \vphantom{\begin{aligned} \dot{x}_1 &= f_1(x_1, x_2) \\ \dot{x}_2 &= f_2(x_1, x_2) \end{aligned}} \right\} \begin{array}{l} \text{(null)} \\ \text{isoclines} \end{array}$$



# Multiplicity of equilibria

We have already seen that equilibria can be multiple

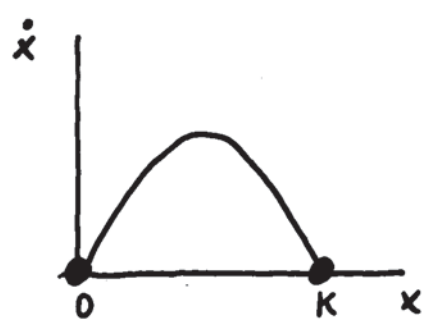


## Ex. 3. Logistic growth.

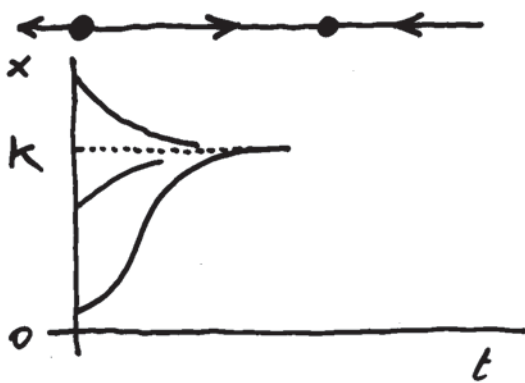
$x(t)$  = biomass at time  $t$

$$\left. \begin{aligned} \dot{x} &= n(x) \cdot x - m(x)x \\ n(x) &= n_0 - \alpha x \\ m(x) &= m_0 + \beta x \end{aligned} \right\} \Rightarrow \dot{x} = r x \left( 1 - \frac{x}{k} \right)$$

$r = n_0 - m_0$   
 $k = \frac{n_0 - m_0}{\alpha + \beta}$



two equilibria  $\bar{x} = 0$   
 $\bar{x} = k$  carrying capacity



the population biomass tends toward its carrying capacity

# Stability

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## Definition 2 ((local) stability) (Liapunov 1892)

An equilibrium  $\bar{x}$  is (locally) stable if for any  $\epsilon > 0$

there exists a  $\delta > 0$  such that

$$\|x(0) - \bar{x}\| < \delta \Rightarrow \|x(t) - \bar{x}\| < \epsilon \quad \forall x(0), t > 0$$



There is no small perturbation of the state, after which the system moves far away from the equilibrium.

## Definition 3 (asymptotic stability)

A stable equilibrium  $\bar{x}$  is asymptotically stable if all perturbed trajectories  $x(t)$  tend to  $\bar{x}$  for  $t \rightarrow \infty$

In words: any small perturbation is asymptotically absorbed

## Definition 4 ((local) instability)

An equilibrium which is not stable is called unstable.

## Definition 5 (basin of attraction)

Given an asymptotically stable equilibrium  $\bar{x}$  the set

$$B(\bar{x}) = \{x(0) : x(t) \rightarrow \bar{x}\}$$

is called basin of attraction

## Definition 6 (global stability)

If  $B(\bar{x})$  coincides with  $\mathbb{R}^n$  (with the exception of a set of zero measure)  $\bar{x}$  is called globally stable

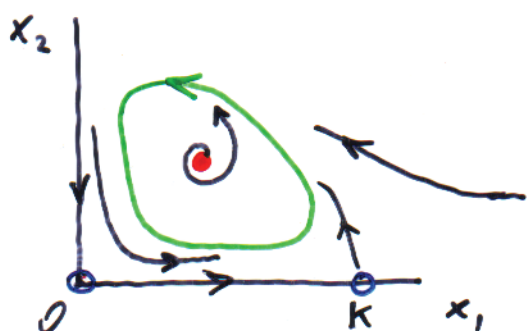
# Examples

## Ex. 4 Logistic growth



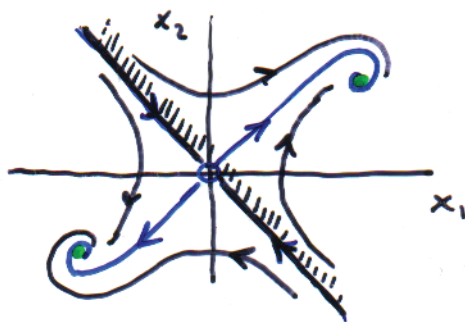
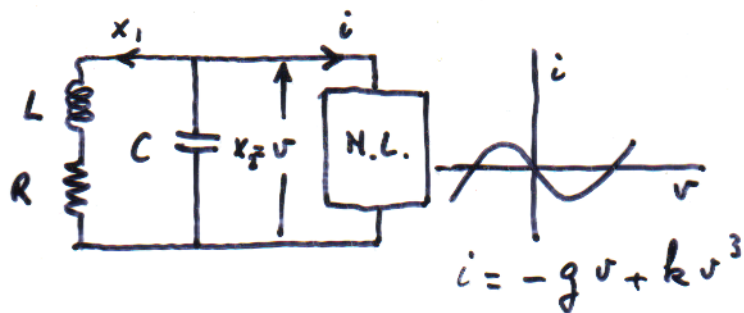
0 is unstable and K is stable  
 K is asymptotically stable  
 K is globally stable (in  $\mathbb{R}_+$ )

## Ex. 5 Prey-predator model



There are three equilibria  
 They are all unstable (one is a focus and two are saddles)

## Ex. 6 Electric circuit (bistable)



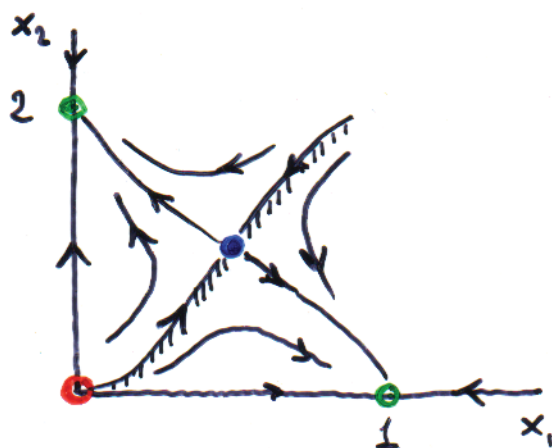
## Ex. 7 Bacteria competition

$$\dot{x}_1 = r_1 x_1 \left(1 - \frac{x_1}{K_1}\right) - \alpha_1 x_1 x_2$$

$$\dot{x}_2 = r_2 x_2 \left(1 - \frac{x_2}{K_2}\right) - \alpha_2 x_1 x_2$$

4 equilibria : 2 stable, 2 unstable

control from 2 to 1 : antibiotic + yeast



# Problems

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P. 1. <sup>(1)</sup> Prove that a linear system  $\dot{x} = Ax$  has a unique equilibrium (namely, the origin  $\bar{x} = 0$ ) if  $A$  is non singular (i.e. if  $A$  does not have a null eigenvalue). What happens if  $A$  is singular? Formulate the analogous results for discrete-time linear systems.

P. 2. <sup>(1)</sup> Sketch the state portrait of the following mechanical system and indicate the basin of attraction of the two stable equilibria



P. 3. <sup>(1)</sup> Consider the following discrete-time 1-st order model

$$x(t+1) = \frac{1}{2 - x(t)}$$

and show that this model has only one equilibrium, namely  $\bar{x} = 1$ . Then, use Moran's construction to prove that the equilibrium is unstable. Finally, observe that all trajectories tend toward  $\bar{x}$ , even if  $\bar{x}$  is unstable. (This model describes the evolution of a genetic disease in a population:  $t$  is the generation and  $x(t)$  is the probability that a randomly selected individual of generation  $t$  is sick).

P. 4. <sup>(2)</sup> Using the model described in Ex. 7 and the isodines try to give a formal support to the state portrait shown in Ex. 7.

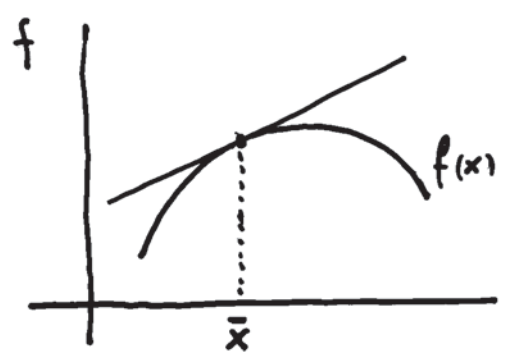
# Local approximation

$$y = f(x)$$

$$y = f(\bar{x}) + \left. \frac{df}{dx} \right|_{\bar{x}} (x - \bar{x}) + \frac{1}{2} \left. \frac{d^2f}{dx^2} \right|_{\bar{x}} (x - \bar{x})^2 + \dots$$

local approximation

$$y = f(\bar{x}) + \left. \frac{df}{dx} \right|_{\bar{x}} (x - \bar{x})$$



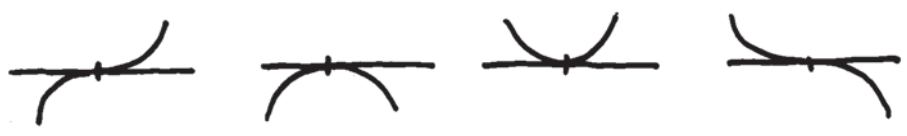
geometric interpretation of linearization

The sign of  $\left. \frac{df}{dx} \right|_{\bar{x}}$  is important

$$\left. \frac{df}{dx} \right|_{\bar{x}} > 0 \quad f(\cdot) \text{ is locally increasing}$$

$$\left. \frac{df}{dx} \right|_{\bar{x}} < 0 \quad f(\cdot) \text{ is locally decreasing}$$

$$\left. \frac{df}{dx} \right|_{\bar{x}} = 0 \quad \text{critical case : nothing can be said}$$





## Jacobian matrix (1)

$$\dot{x}(t) = f(x(t))$$

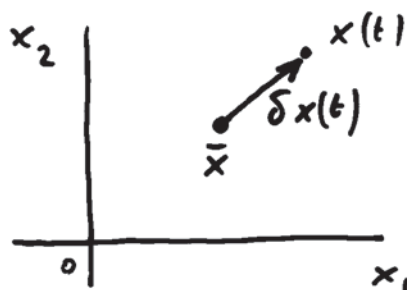
$$\bar{x} = \text{equilibrium } (f(\bar{x}) = 0)$$

### Taylor's expansion

$$\dot{x} = f(\bar{x}) + \left[ \frac{\partial f}{\partial x} \right]_{\bar{x}} (x - \bar{x}) + \dots$$

$$\uparrow O(x - \bar{x})^2$$

$$\left[ \frac{\partial f}{\partial x} \right] = \text{Jacobian (matrix)} = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial f_n}{\partial x_1} & \dots & \frac{\partial f_n}{\partial x_n} \end{pmatrix} = J$$



$$\delta x(t) = x(t) - \bar{x}$$

$$\delta \dot{x}(t) = (\dot{x}(t) - \dot{\bar{x}}) = \left[ \frac{\partial f}{\partial x} \right]_{\bar{x}} \delta x(t) + \dots$$

### Linearized system

$$\boxed{\delta \dot{x} = J \delta x}$$

### Ex. 1 - Logistic growth

$$\dot{x} = r x \left( 1 - \frac{x}{K} \right) \quad \bar{x} = 0 \Rightarrow \delta \dot{x} = r x$$

$$\left( J = r - \frac{2rx}{K} \right) \quad \bar{x} = K \Rightarrow \delta \dot{x} = -r x$$

## Jacobian matrix (2)

$$x(t+1) = f(x(t))$$

$$\bar{x} = \text{equilibrium} \quad (\bar{x} = f(\bar{x}))$$

$$x(t+1) = f(\bar{x}) + \left. \frac{\partial f}{\partial x} \right|_{\bar{x}} (x(t) - \bar{x}) + \dots$$

$$x(t+1) = \bar{x} + \left. \frac{\partial f}{\partial x} \right|_{\bar{x}} (x(t) - \bar{x}) + \dots$$

$$\delta x(t+1) = \left. \frac{\partial f}{\partial x} \right|_{\bar{x}} \delta x(t) + \dots$$

$$\delta x(t+1) = J \delta x(t)$$

linearized system

Ex. 2 - Age structured population

$x_i(t)$  = # individuals of age  $i$  in year  $t$

$$x_1(t+1) = x_3(t) \cdot F(x_3(t))$$

←  $F$  = fertility

$$x_2(t+1) = s_2 x_1(t)$$

↔  $s$  = survival

$$\rightarrow x_3(t+1) = s_3 x_2(t) + s^* x_3(t)$$

↳ mature individuals

$$\delta x(t+1) = J \delta x(t)$$

$$J = \begin{vmatrix} 0 & 0 & F(\bar{x}_3) + \left. \frac{\partial F}{\partial x_3} \right|_{\bar{x}_3} \\ s_2 & 0 & 0 \\ 0 & s_3 & s^* \end{vmatrix}$$

$\swarrow$   $(1-s^*)/s_2 s_3$



# Linearization method

$$\left. \begin{aligned} \dot{x} &= f(x) \\ x(t+1) &= f(x(t)) \end{aligned} \right\} \bar{x} = \text{equilibrium} \Rightarrow J = \left. \frac{\partial f}{\partial x} \right|_{\bar{x}}$$

Theorem 1  $J = \text{asympt. stable} \Rightarrow \bar{x} = \text{asympt. stable}$

In words: if the linearized system is asymptotically stable, the equilibrium  $\bar{x}$  is such

Theorem 2  $J = \text{exponentially unstable} \Rightarrow \bar{x} = \text{unstable}$



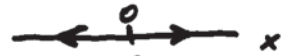

↑  
at least one eigenvalue  
of  $J$  is strictly unstable  
( $\text{Re}(\lambda_i) > 0$  or  $|\lambda_i| > 1$ )

Comment If  $J$  is simply stable or weakly unstable nothing can be said on the stability of  $\bar{x}$ .

## Ex. 3. Logistic growth

$$\begin{aligned} \dot{x} &= r x \left(1 - \frac{x}{K}\right) & \bar{x} = 0 & J = r \Rightarrow \bar{x} = \text{unstable} \\ & & \bar{x} = K & J = -r \Rightarrow \bar{x} = \text{asympt. stable} \end{aligned}$$

## Ex. 4. Quadratic and cubic systems

$\dot{x} = x^2$	$\bar{x} = 0$	$J = 2x \Big _{\bar{x}} = 0 \Rightarrow ?$	
$\dot{x} = -x^2$	$\bar{x} = 0$	$J = -2x \Big _{\bar{x}} = 0 \Rightarrow ?$	
$\dot{x} = x^3$	$\bar{x} = 0$	$J = 0 \Rightarrow ?$	
$\dot{x} = -x^3$	$\bar{x} = 0$	$J = 0 \Rightarrow ?$	

## Problems

P. 1. <sup>(3)</sup> Consider the prey-predator model

$$\begin{cases} \dot{x}_1 = r x_1 \left(1 - \frac{x_1}{K}\right) - \frac{a x_1}{b + x_1} x_2 & \text{prey} \\ \dot{x}_2 = e \frac{a x_1}{b + x_1} x_2 - d x_2 & \text{predator} \end{cases}$$

and assume that the positive parameters  $a, b, e, d, r, K$  satisfy the following inequalities

$$b < K \quad (1) \qquad \frac{ea}{d} > \frac{K+b}{K-b} \quad (2)$$

Show that under conditions (1, 2) the system has two trivial equilibria

$$\bar{x}^{(1)} = \begin{vmatrix} 0 \\ 0 \end{vmatrix} \quad \text{and} \quad \bar{x}^{(2)} = \begin{vmatrix} K \\ 0 \end{vmatrix}$$

and one strictly positive equilibrium

$$\bar{x}^{(3)} = \begin{vmatrix} \bar{x}_1 \\ \bar{x}_2 \end{vmatrix} \quad \bar{x}_i \geq 0 \quad i = 1, 2$$

- Show, through linearization, that  $\bar{x}^{(1)}$ ,  $\bar{x}^{(2)}$  and  $\bar{x}^{(3)}$  are unstable.
- Compare with Ex. 5 of Lecture 3.

P. 2. <sup>(3)</sup> Discuss, through linearization, the stability of the four equilibria shown in Ex. 7 of Lecture 3 (pay attention to inequalities among parameters).

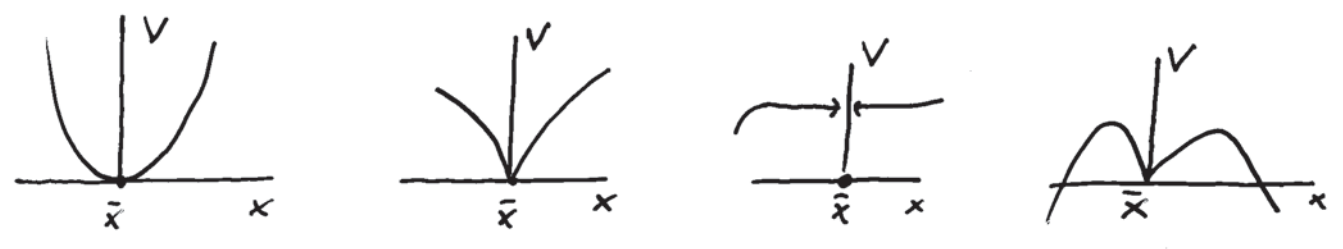
# POSITIVE DEFINITE FUNCTIONS

$$x \in \mathbb{R}^n \quad V(x) : \mathbb{R}^n \rightarrow \mathbb{R}$$

Definition  $V(\cdot)$  is positive definite at  $\bar{x}$  iff

(a)  $V(\bar{x}) = 0$

(b)  $V(x) > 0 \quad \forall x \neq \bar{x}$  in a neighborhood of  $\bar{x}$



In  $\mathbb{R}^2$

$$V(x) = x_1^2 + x_2^2$$

$$V(x) = x_1 - \log x_1 + x_2 - \log x_2 - 2$$

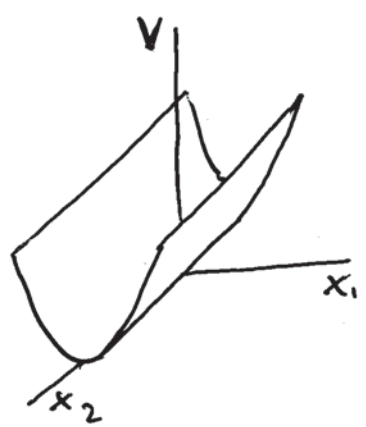
$$V(x) = a x_1^2 + 2b x_1 x_2 + c x_2^2 \quad \text{with } a > 0, ac > b^2$$

Definition  $V(\cdot)$  is positive semidefinite at  $\bar{x}$  iff

(a)  $V(\bar{x}) = 0$

(b)  $V(x) \geq 0 \quad \forall x \neq \bar{x}$  in a neighborhood of  $\bar{x}$

(c)  $V(x) = 0$  for some  $x$  close to  $\bar{x}$



$V = x_1^2$  is posit. semid. in  $\mathbb{R}^2$

Definition  $V(\cdot)$  is negative (semi)definite iff  $-V(\cdot)$  is positive (semi)definite

# LIAPUNOV METHOD [1892]

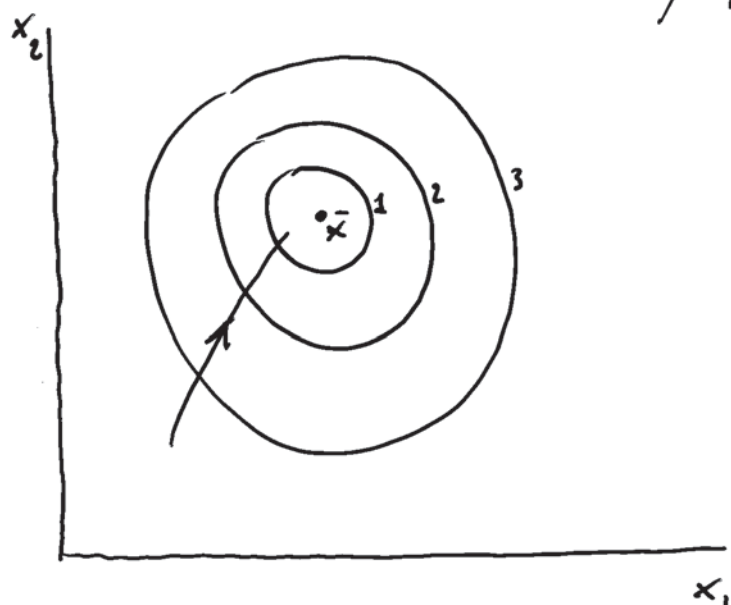
$$\dot{x} = f(x) \quad f(\bar{x}) = 0$$

(a)  $V(\cdot)$  posit. def. at  $\bar{x}$

(b)  $V(\cdot)$  regular (continuous with continuous derivatives)

(c)  $\dot{V}(x) = \frac{\partial V}{\partial x} f = \frac{\partial V}{\partial x_1} f_1 + \frac{\partial V}{\partial x_2} f_2 + \dots + \frac{\partial V}{\partial x_n} f_n$  negative definite at  $\bar{x}$

⇓  
 $\bar{x} = \text{asympt. stable}$



(a) + (b)  $\Rightarrow$  the lines at  $V = \text{const.}$  are closed and ordered close to  $\bar{x}$ .

$\dot{V}$  neg. def.  $\Rightarrow$  trajectories cross the lines  $V = \text{const.}$  transversally from outside toward inside

Remark A function  $V(\cdot)$  satisfying conditions (a), (b) and (c) is called a Liapunov function

## Krasowskii's method

(a) + (b) +  $\dot{V}$  negative semidefinite + Krasowskii condition (\*)

⇓  
 $\bar{x} = \text{asympt. stable}$

$$K = \{x : \dot{V}(x) = 0\}$$

(\*)  $K$  does not contain trajectories close to  $\bar{x}$  but  $\neq \bar{x}$

## LA SALLE METHOD

$$\dot{x} = f(x) \quad f(\bar{x}) = 0$$

$V(\cdot)$  is a Liapunov (or Krasovskii) function

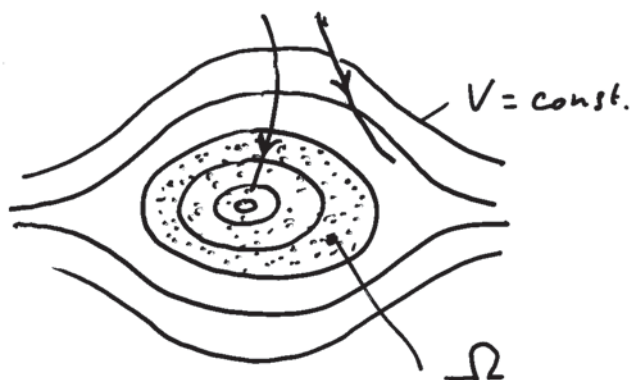
$$\Downarrow$$

$$\bar{x} = \text{asympt. stable}$$

$$\Downarrow$$

$$B(\bar{x}) = ?$$

La Salle :  $B(\bar{x}) \supset \Omega$  = the largest region contained in a closed line  $V(x) = \text{const.}$  in which Liapunov (or Krasows.) conditions hold



Remark  $\Omega$  is an estimate (from below) of  $B(\bar{x})$

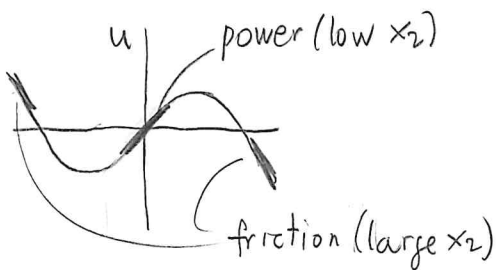
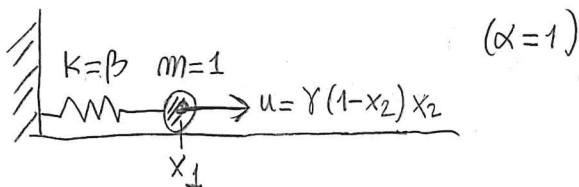
Remark quadratic positive definite functions  $V(\cdot)$  have always closed lines  $V = \text{const.}$

# EXAMPLE (Van der Pol equation)

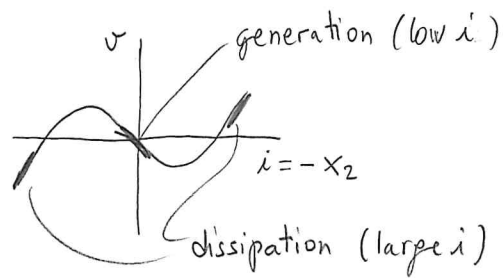
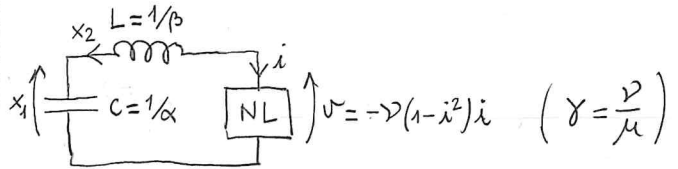
$$\dot{x}_1 = \alpha x_2$$

$$\dot{x}_2 = -\beta x_1 + \gamma(1-x_2^2)x_2$$

Mechanical interpretation



Electrical interpretation



Equilibrium  $\bar{x} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

$$J = \begin{bmatrix} 0 & \alpha \\ -\beta & \gamma \end{bmatrix} \quad \text{tr } J = \gamma \quad \det J = \alpha\beta$$

$$\bar{x} \text{ as. stab. for } \begin{cases} \gamma < 0 \\ \alpha\beta > 0 \end{cases}$$

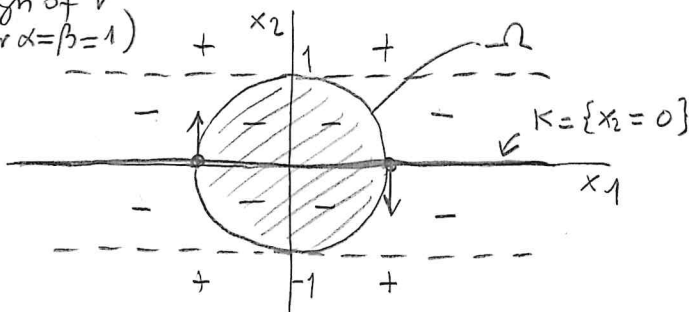
Assume  $\alpha > 0, \beta > 0, \gamma < 0$  (note: the Van der Pol oscillator has  $\gamma > 0$ !)

$V(x) = \frac{c_1}{2} x_1^2 + \frac{c_2}{2} x_2^2$ , energy-like function with elliptic contour lines  
coefficient  $c_1$  and  $c_2$  to be determined to simplify the expression of  $\dot{V}$

$$\begin{aligned} \dot{V} &= c_1 x_1 \dot{x}_1 + c_2 x_2 \dot{x}_2 = \underbrace{c_1 x_1}_{\beta} \alpha x_2 + \underbrace{c_2 x_2}_{\alpha} (-\beta x_1 + \gamma(1-x_2^2)x_2) \\ &= \alpha \gamma x_2^2 (1-x_2^2) \leq 0 \text{ for } |x_2| < 1 \end{aligned}$$

$$= 0 \text{ for } x_2 = 0 \text{ and } x_2 = \pm 1$$

sign of  $\dot{V}$   
(for  $\alpha = \beta = 1$ )



$\Rightarrow \bar{x}$  as. stab