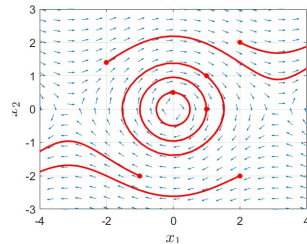


## Chapitre 2 : Phase Plane

Yassine ARIBA



## Sommaire

- ① Introduction and definitions
- ② Construction of phase portrait
- ③ Linear systems case
- ④ Closed orbits
- ⑤ Case study

## Sommaire

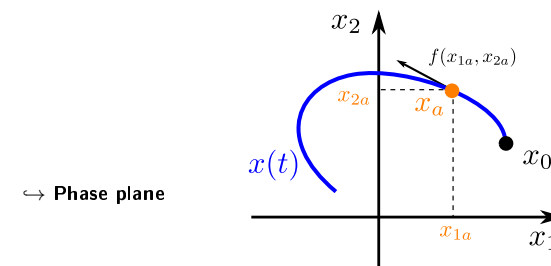
- ① Introduction and definitions
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## Second-order systems

In general, one can not find solution  $x(t)$  of a nonlinear system

Some techniques exist to draw  $x(t)$  for second-order system in a plane

$$\dot{x} = f(x) \equiv \begin{cases} \dot{x}_1 = f_1(x_1, x_2) \\ \dot{x}_2 = f_2(x_1, x_2) \end{cases} \quad \text{with } x(0) = x_0 = \begin{bmatrix} x_{10} \\ x_{20} \end{bmatrix}$$



### Definitions

#### Trajectory or orbit

The curve of  $x(t)$  in the  $x_1 - x_2$  plane is called a *trajectory* or *orbit* of the system from the point  $x_0$ .

#### Phase portrait

The *phase portrait* of the system is the set of all trajectories for different initial conditions  $x_0$ .

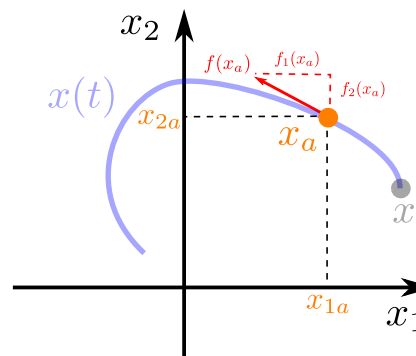
#### Vector field

The *vector field* is the representation, in the  $x_1 - x_2$  plane, of the vector  $f(x) = (f_1(x_1, x_2), f_2(x_1, x_2))$ . It is drawn with arrows.

### Vector field

The vector  $f(x) = (f_1(x), f_2(x))$  is tangent to the trajectory at point  $x$

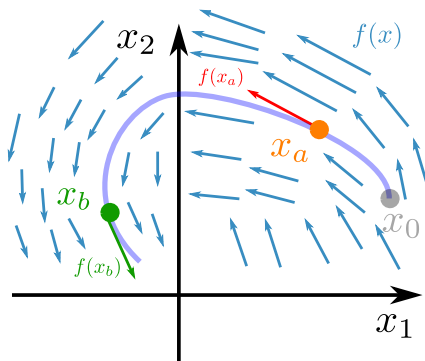
$$\frac{dx_2}{dx_1} = \frac{f_1(x)}{f_2(x)}$$



### Vector field

The vector  $f(x) = (f_1(x), f_2(x))$  is tangent to the trajectory at point  $x$

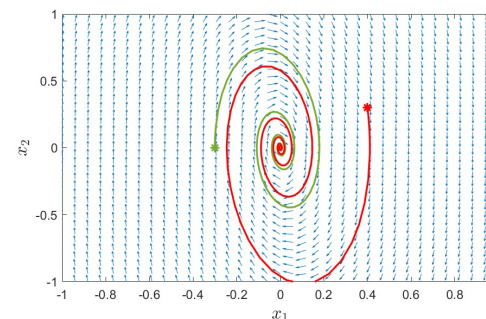
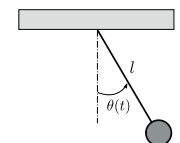
$$\frac{dx_2}{dx_1} = \frac{f_1(x)}{f_2(x)}$$



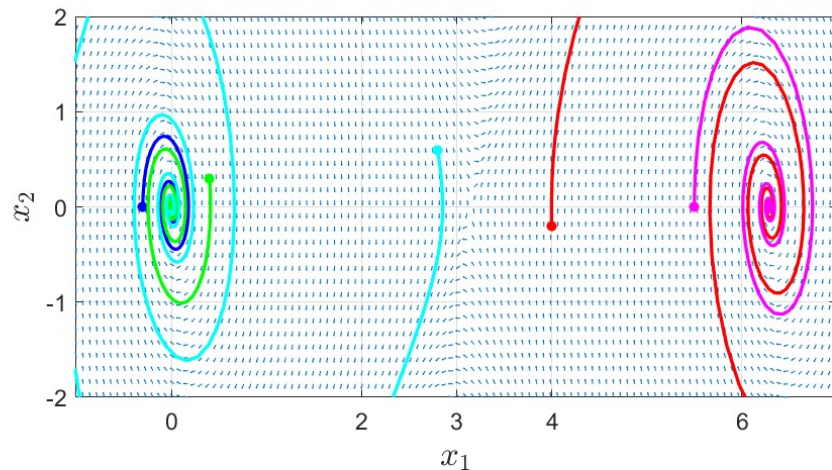
### Pendulum example

Variables :  $x_1 = \theta$  and  $x_2 = \dot{\theta}$

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ -\frac{g}{l} \sin x_1 - \frac{k}{m} x_2 \end{bmatrix}$$



### Pendulum example



## Sommaire

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### Construction of phase portrait

Several techniques exist to draw trajectories on the phase plane

Two will be presented here :

- ▶ analytical method - solve the differential equations
- ▶ isoclines method - graphical method

◊ But nowadays numerical computing softwares are used (MATLAB, Scilab, Python)

### Analytical method

The objective is to get a relationship between  $x_1$  and  $x_2$

$$g(x_1, x_2) = 0$$

- ▶ First approach : solve the state equation

$$\begin{cases} \dot{x}_1 = f_1(x_1, x_2) \\ \dot{x}_2 = f_2(x_1, x_2) \end{cases} \Rightarrow \begin{cases} x_1 = g_1(t) \\ x_2 = g_2(t) \end{cases}$$

Eliminate the time  $t$  between the two parametric curves

- ▶ Second approach : Eliminate the time  $t$  first

$$\frac{dx_2}{dx_1} = \frac{f_2(x_1, x_2)}{f_1(x_1, x_2)}$$

Solve the new differential equation (with separated variables)

◊ These methods are restricted to quite simple/particular nonlinearities

### Example

Consider the system

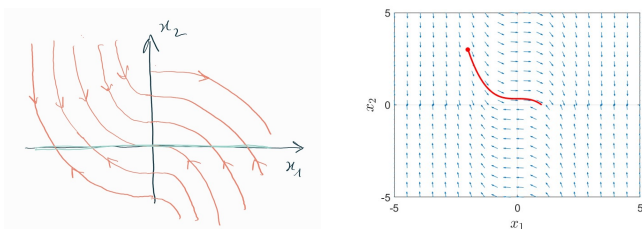
$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -x_2 x_1^2 \end{cases} \quad \text{with} \quad x_0 = \begin{bmatrix} x_{10} \\ x_{20} \end{bmatrix}$$

► Equilibrium points :  $x_1^* \in \mathbb{R}$  and  $x_2^* = 0 \Rightarrow x_1$ -axis

► Analytical resolution :

$$x_2 = -\frac{1}{3}x_1^3 + \underbrace{x_{20} + \frac{1}{3}x_{10}^3}_{\text{cst}}$$

► Sketch and simulation



### Isoclines method

Isocline = locus in the phase plane of trajectory's points of given slope  $\alpha$

$$s(x_1, x_2) = \alpha = \frac{dx_2}{dx_1} = \frac{f_2(x_1, x_2)}{f_1(x_1, x_2)}$$

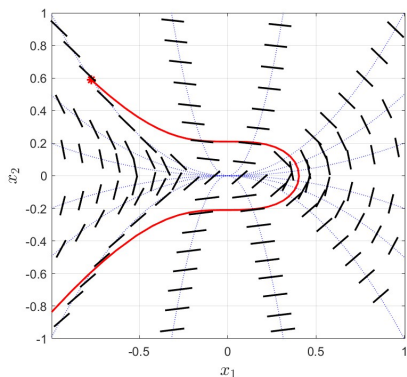
Step :

- For a given  $\alpha$ , draw the curve such that  $s(x_1, x_2) = \alpha$
- Along the curve, draw small segments of slope  $\alpha$
- Each segment is tangent to a trajectory, the direction  $s$  given by sign of  $f_1(x)$  and  $f_2(x)$
- Repeat from first step to draw several isoclines, for different  $\alpha$
- Then, from a given initial condition  $x_0$ , sketch a solution joining segments

◊ Also restricted to quite simple/particular nonlinearities

### Example

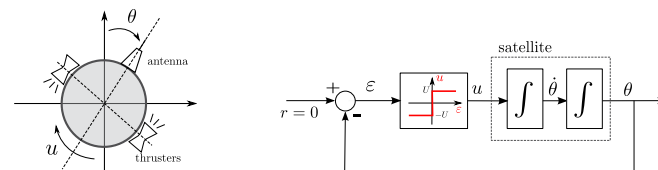
$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -x_1^2 \end{cases} \quad \text{slope :} \quad \alpha = \frac{f_2(x)}{f_1(x)} = \frac{-x_1^2}{x_2} \Leftrightarrow x_2 = -\frac{1}{\alpha} x_1^2$$



Plot for  $\alpha = \{-5, -2, -1, -0.1, 0.1, 1, 2, 5\}$

### Exercise (analytical method)

Consider the simple control of a simple satellite model



- Write the state space model
- What is (are) the the equilibrium point(s)?
- Express  $x_1$  as a function of  $x_2$
- Draw a sketch of the phase portrait.

**Exercise (isocline method)**

Consider the previous (controlled) system

- ▶ Apply the isocline method to retrieve the phase portrait

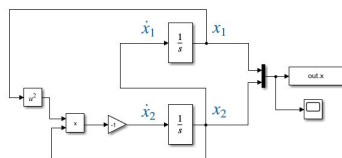
**Solution :**

**Numerical simulations**

General steps with MATLAB

- ▶ Define the system (function  $f$ ) with a MATLAB function or Simulink

```
% anonymous functions
f = @(t,x) [x(2); -x(2)*x(1)^2];
```



- ▶ Select an initial point  $x_0$
- ▶ Solve the differential equation  $\dot{x} = f(x)$

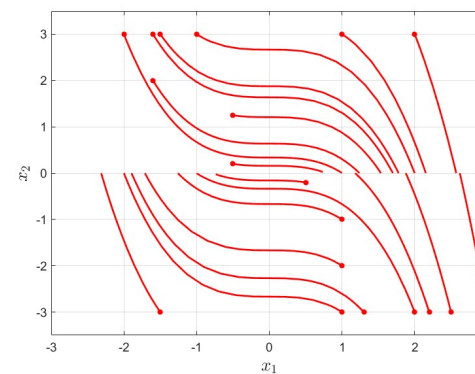
```
x0 = [-2; 3];
[t,x] = ode45(f,[0 20],x0);
x1 = x(:,1);
x2 = x(:,2);
plot(x1,x2);
plot(x1(1),x2(2),'*');
```



- ▶ Repeat from step 2

**Numerical simulations**

Resulting plot for several  $x_0$



In MATLAB, the instruction `quiver` plots the vector field

## Sommaire

- 1 Introduction and definitions
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## What about linear systems ?

Autonomous linear system :

$$\begin{cases} \dot{x}_1 = a_{11} x_1 + a_{12} x_2 \\ \dot{x}_2 = a_{21} x_1 + a_{22} x_2 \end{cases} \Leftrightarrow \dot{x} = Ax$$

► Solution :  $x(t) = e^{At}x_0$

► Jordan canonical form with a change of basis :  $Mz = x$

Simpler system :  $\dot{z} = \underbrace{M^{-1}AM}_J z \Rightarrow$  Solution :  $z(t) = e^{Jt}z_0$

► According to eigenvalues of  $A \rightarrow$  different forms for  $J$

$$\begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \quad \begin{bmatrix} \lambda & k \\ 0 & \lambda \end{bmatrix} \quad \begin{bmatrix} \alpha & -\beta \\ \beta & \alpha \end{bmatrix}$$

( $k = 0$  or  $1$ ) / (if an eigenvalue = 0  $\rightarrow$  specific study)

## Case 1 : real distinct eigenvalues

Two eigenvalues :  $\lambda_1 \neq \lambda_2 \neq 0$

► Change of basis matrix  $M = [v_1, v_2]$  made of the eigenvectors

► Give two decoupled first-order differential equation

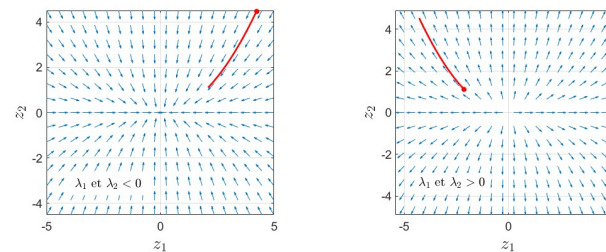
$$\begin{cases} \dot{z}_1 = \lambda_1 z_1 \\ \dot{z}_2 = \lambda_2 z_2 \end{cases} \Rightarrow \begin{cases} z_1(t) = z_{10} e^{\lambda_1 t} \\ z_2(t) = z_{20} e^{\lambda_2 t} \end{cases}$$

► Eliminate the time  $t$

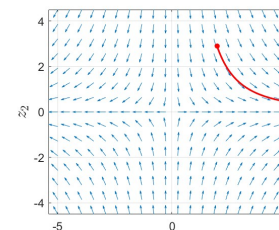
$$z_2 = c z_1^{\lambda_2/\lambda_1} \quad \text{with } c = \frac{z_{20}}{z_{10}^{\lambda_2/\lambda_1}}$$

The shape of the curves depends on signs of  $\lambda_1$  and  $\lambda_2$

► Same signs  $\Rightarrow$  the equilibrium point is a **stable** or **unstable node**

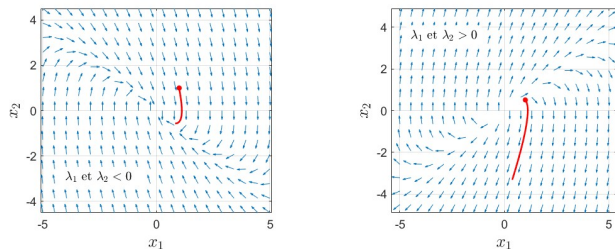


► Opposite signs  $\Rightarrow$  the equilibrium point is a **saddle point**

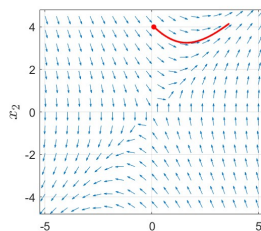


Back in the  $x$ -coordinates basis :  $x = Mz$

- ▶ Same signs  $\Rightarrow$  the equilibrium point is a **stable or unstable node**



- ▶ Opposite signs  $\Rightarrow$  the equilibrium point is a **saddle point**



### Case 2 : real identical eigenvalues

Two eigenvalues :  $\lambda_1 = \lambda_2 = \lambda \neq 0$

- ▶ Change of basis matrix  $x = Mz$  (eigenvectors or chain of eigenvect.)
- ▶ Give two first-order differential equation

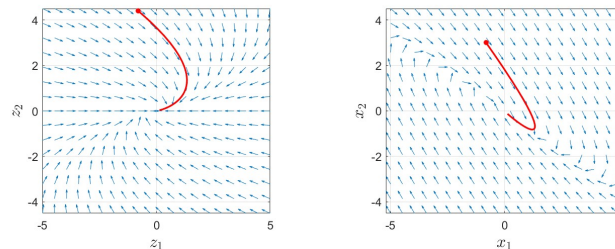
$$\begin{cases} \dot{z}_1 = \lambda z_1 + k z_2 \\ \dot{z}_2 = \lambda z_2 \end{cases} \Rightarrow \begin{cases} z_1(t) = (z_{10} + k z_{20} t) e^{\lambda t} \\ z_2(t) = z_{20} e^{\lambda t} \end{cases}$$

- ▶ If  $k = 0$ , particular case of the previous one
- ▶ Eliminate the time  $t$

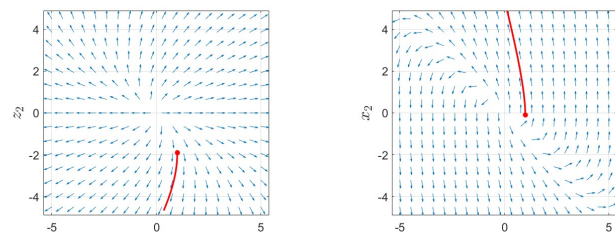
$$z_1 = z_2 \left( \frac{z_{10}}{z_{20}} + \frac{k}{\lambda} \ln \left( \frac{z_2}{z_{20}} \right) \right) \quad \text{and also} \quad \frac{dz_2}{dz_1} = \frac{\lambda z_2}{\lambda z_1 + k z_2}$$

Again, the shape of the curves depends on sign of  $\lambda$

- ▶ negative  $\Rightarrow$  the equilibrium point is a **stable node**



- ▶ positive  $\Rightarrow$  the equilibrium point is an **unstable node**



### Case 3 : complex conjugate eigenvalues

Two eigenvalues :  $\lambda_{1,2} = \alpha \pm j\beta$

$\rightarrow$  Two complex conj. eigenvectors  $v_1$  and  $v_2 = \bar{v}_1$

- ▶ Change of basis matrix with  $M = [Re[v_1], Im[v_1]]$

$$\begin{cases} \dot{z}_1 = \alpha z_1 + \beta z_2 \\ \dot{z}_2 = -\beta z_1 + \alpha z_2 \end{cases}$$

- ▶ Change of variable  $\rightarrow$  polar coordinates :  $z_1 = r \cos \theta$  and  $z_2 = r \sin \theta$

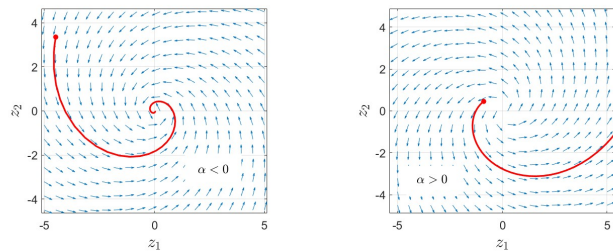
$$\begin{cases} \dot{r} = \alpha r \\ \dot{\theta} = -\beta \end{cases}$$

- ▶ that has for solution :

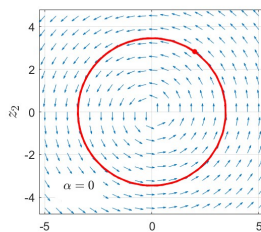
$$\begin{cases} r(t) = r_0 e^{\alpha t} \\ \theta(t) = -\beta t + \theta_0 \end{cases} \quad \text{with} \quad \begin{cases} r_0 = \sqrt{z_{10}^2 + z_{20}^2} \\ \theta_0 = \arctan \frac{z_{20}}{z_{10}} \end{cases}$$

The shape of the curves depends on signs of  $\alpha = \text{Re}[\lambda]$

- ▶ negative or positive real part  $\Rightarrow$  the equ. pt is a **stable** or **unstable focus**

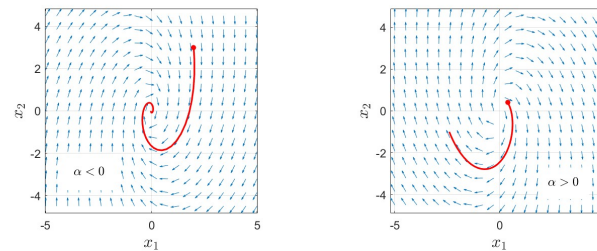


- ▶ Pure imaginary  $\Rightarrow$  the equilibrium point is a **center** (circle of radius  $r_0$ )

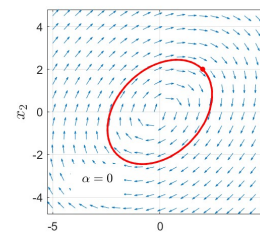


Back in the  $x$ -coordinates basis :  $x = Mz$

- ▶ negative or positive real part  $\Rightarrow$  the equ. pt is a **stable** or **unstable focus**



- ▶ Pure imaginary  $\Rightarrow$  the equilibrium point is a **center** (circle of radius  $r_0$ )



### Case 4 (degenerate) : one or both eigenvalues are zero

Matrix  $A$  is singular  $\rightarrow$  an equilibrium subspace (infinitely many points)

First case :  $\lambda_1 = 0$  and  $\lambda_2 \neq 0$

- ▶ Change of basis gives

$$\begin{cases} \dot{z}_1 = 0 \\ \dot{z}_2 = \lambda_2 z_2 \end{cases} \Rightarrow \begin{cases} z_1(t) = z_{10} \\ z_2(t) = z_{20} e^{\lambda_2 t} \end{cases}$$

- ▶ if  $\lambda_2 < 0$ , trajectories converge, and if  $\lambda_2 > 0$ , they diverge

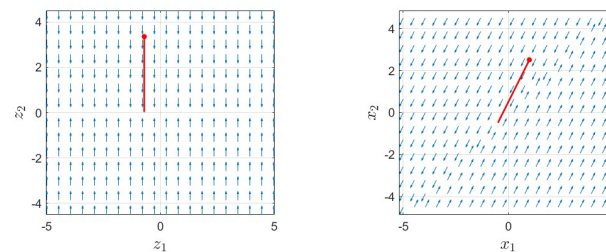
Second case :  $\lambda_1 = \lambda_2 = 0$

- ▶ Change of basis gives

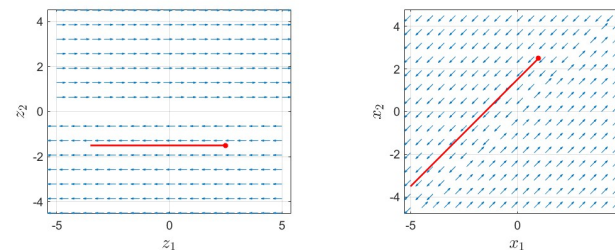
$$\begin{cases} \dot{z}_1 = z_2 \\ \dot{z}_2 = 0 \end{cases} \Rightarrow \begin{cases} z_1(t) = z_{10} + z_{20} t \\ z_2(t) = z_{20} \end{cases}$$

- ▶  $z_1$  increases or decreases depending on the sign of  $z_{20}$

- ▶ First case,  $\lambda_1 = 0$  and  $\lambda_2 \neq 0$  (below  $\lambda_2 < 0$ )



- ▶ Second case,  $\lambda_1 = \lambda_2 = 0$





### Recap

Qualitative behavior for linear systems around the isolated equilibrium  $x = 0$

- ▶ Real eigenvalues
  - $\lambda_1$  and  $\lambda_2$  positive  $\Rightarrow$  **unstable node**
  - $\lambda_1$  and  $\lambda_2$  negative  $\Rightarrow$  **stable node**
  - $\lambda_1$  and  $\lambda_2$  opposite  $\Rightarrow$  **saddle point**
- ▶ Complex conjugate eigenvalues
  - real part  $\alpha > 0 \Rightarrow$  **unstable focus**
  - real part  $\alpha < 0 \Rightarrow$  **stable focus**
  - real part  $\alpha = 0 \Rightarrow$  **center**

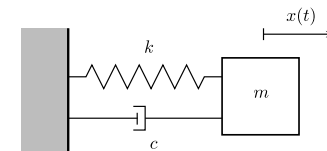
Behavior determined by the eigenvalues of  $A$

- ▶ Determined for the whole plane (**global**), characteristic of linear systems
- ▶ For nonlinear systems, study interesting to get the **local** behavior around an equilibrium point

### Example : simple mass-spring system

Equation of motion :

mass ( $m = 1 \text{ kg}$ )  
spring (stiffness :  $k = 1 \text{ N/m}$ )  
damper (viscous coefficient :  $c \text{ N/m/s}$ )

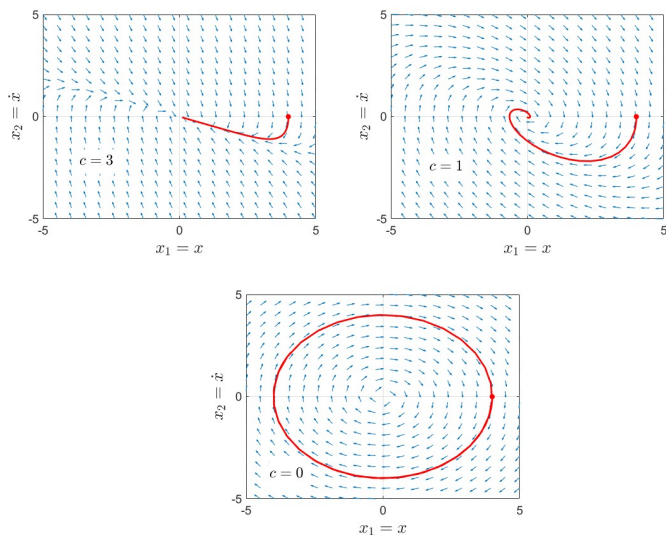


$$\ddot{x} + c\dot{x} + x = 0 \quad \Rightarrow \quad \begin{bmatrix} \dot{x} \\ \ddot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & -c \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} \quad \text{with} \quad \begin{cases} x(0) = x_0 \\ \dot{x}(0) = 0 \end{cases}$$

Eigenvalues of the dynamic matrix

$c \geq 2$	$0 < c < 2$	$c = 0$
$\lambda_{1/2} = \frac{-c \pm \sqrt{c^2 - 4}}{2}$	$\lambda_{1/2} = -\frac{c}{2} \pm i \frac{\sqrt{ c^2 - 4 }}{2}$	$\lambda_{1/2} = \pm i$
noeud stable	foyer stable	centre

### Simulation of the mass-spring system



### Exercise 1

Consider the system

$$\dot{x} = \begin{bmatrix} -2 & 2 \\ 1 & -3 \end{bmatrix} x \quad \text{with} \quad x_0 = \begin{bmatrix} x_{10} \\ x_{20} \end{bmatrix}$$

- ▶ What is the qualitative behavior of the equilibrium point 0 ?
- ▶ What is the representation of the system in the z-coordinates ?
- ▶ Draw a sketch of the phase portrait in z and x-coordinates.

Solution :

## Exercise 2

Consider the system

$$\dot{x} = \begin{bmatrix} 1 & -1 \\ 9 & 1 \end{bmatrix} x \quad \text{with} \quad x_0 = \begin{bmatrix} x_{10} \\ x_{20} \end{bmatrix}$$

- ▶ What is the qualitative behavior of the equilibrium point 0 ?
- ▶ What is the representation of the system in the z-coordinates ?
- ▶ Draw a sketch of the phase portrait in z-coordinates.

Solution :

## Sommaire

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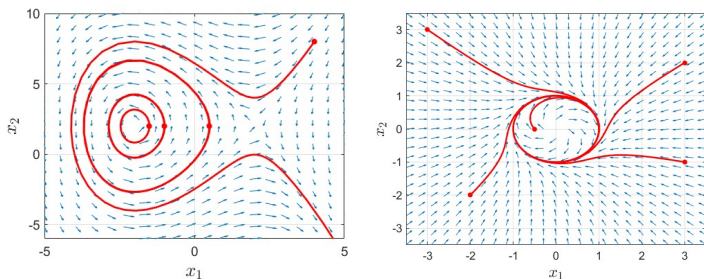
- ① Introduction and definitions
- ② Construction of phase portrait
- ③ Linear systems case
- ④ Closed orbits
- ⑤ Case study

### Closed orbits

A closed orbit is a periodic trajectory

Two cases can be distinguished :

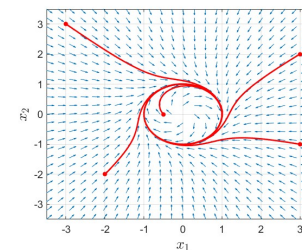
- ▶ **non-isolated** : there are other closed curves in the neighborhood, depend on initial conditions (left)
- ▶ **isolated** : from initial conditions in the neighborhood, trajectories converge or diverge from it → limit cycle (right)



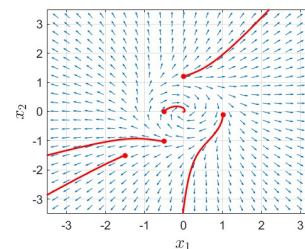
### Limit cycles

Three kinds of limit cycle can be observed

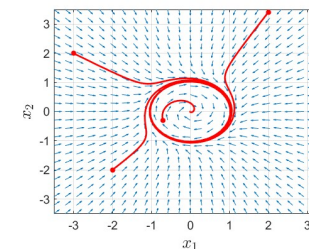
Stable limit cycle



Unstable limit cycle



Semi-stable limit cycle



### Existence of limit cycles

Can we predict the existence of a limit cycle?

3 theorems are stated that may help (valid only for 2<sup>nd</sup> order autonomous systems)

#### Theorem (Poincaré)

If a closed orbit exists, then  $N = S + 1$ , with

- $N$ , the number of nodes/centers/foci enclosed by the closed orbit
- $S$ , the number of saddle points enclosed by the closed orbit

↔ A closed orbit must enclose at least one equilibrium point

#### Theorem (Poincaré-Bendixson)

If a trajectory remains in a closed bounded region  $\mathcal{D}$  in the phase plane, then one of the following is true :

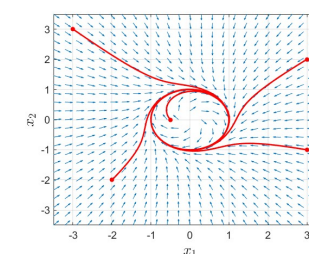
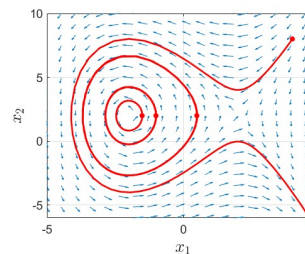
- the trajectory goes to an equilibrium
- the trajectory tends to a closed orbit
- the trajectory is itself a closed orbit

↔ Asymptotic properties of trajectories

These results can be easily verified on previous examples

$$\begin{cases} \dot{x}_1 = 4 - 2x_2 \\ \dot{x}_2 = 12 - 3x_1^2 \end{cases}$$

$$\begin{cases} \dot{x}_1 = x_2 - x_1(x_1^2 + x_2^2 - 1) \\ \dot{x}_2 = -x_1 - x_2(x_1^2 + x_2^2 - 1) \end{cases}$$



### Non-existence condition

This last theorem provides a sufficient condition for the non-existence of a limit cycle

#### Theorem (Bendixson)

No limit cycle can exist in a region  $\mathcal{D}$  of the phase plane in which

$$\frac{\partial f_1}{\partial x_1} + \frac{\partial f_2}{\partial x_2}$$

does not vanish and does not change sign

**Example :**

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -ax_1(1 - bx_1^2) - cx_2 \end{cases} \quad \text{with positive parameters } a, b, c > 0$$

Let's apply formula

$$\frac{\partial f_1}{\partial x_1} + \frac{\partial f_2}{\partial x_2} = 0 - c$$

$\Leftrightarrow \neq 0$  and no change of sign  $\Rightarrow$  no limit cycle

## Sommaire

- 1 Introduction and definitions
- 2 Construction of phase portrait
- 3 Linear systems case
- 4 Closed orbits
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### Case study

**Prey-Predator model** (or Lotka-Volterra model)

study the evolution of two populations  $x_1$  (preys) and  $x_2$  (predators)

$$\begin{cases} \dot{x}_1 = \alpha x_1 - \beta x_1 x_2 \\ \dot{x}_2 = \gamma x_2 x_1 - \delta x_2 \end{cases}$$

$\alpha, \beta, \gamma$  and  $\delta$  are positive constant parameters

- ▶  $\alpha x_1$  is the growth rate of preys if there is no predators
- ▶  $\beta x_1 x_2$  is the death rate of preys because of predators
- ▶  $\gamma x_2 x_1$  is the growth rate of predators with  $x_1$  preys available
- ▶  $\delta x_2$  is the death rate of predators

To simplify, let's set  $\alpha = \beta = \gamma = \delta = 1$

Model :

$$\begin{cases} \dot{x}_1 = x_1(1 - x_2) \\ \dot{x}_2 = x_2(x_1 - 1) \end{cases}$$

- ▶ What is (are) the equilibrium point(s) ?
- ▶ Calculate the linearized model around it (them).
- ▶ What is (are) their nature ? Then, how heights will evolve ?
- ▶ Simulate the system to draw the phase portrait.

## Solution

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## In short

- ▶ Phase plane : study of the time evolution of the state for second order systems
  - ↔ trajectories of  $x = (x_1, x_2)$  in the plane and vector field
- ▶ Usually, numerical software are used to simulate system responses
  - ↔ with MATLAB, Scilab, Python... or your own program implementing numerical methods
- ▶ In the linear case, analytical solutions can be found and the nature of equilibrium point can be derived from eigenvalues
  - ↔ node, saddle point, focus, center, stable/unstable
- ▶ Useful when linearizing nonlinear systems to have the local behavior (around an equilibrium point)