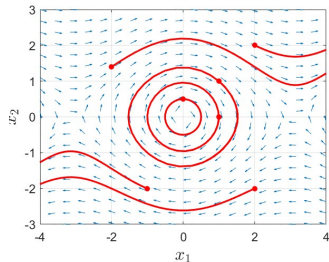


## Chapitre 2 : Phase Plane

Yassine ARIBA



# Sommaire

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- ➊ Introduction and definitions
- ➋ Construction of phase portrait
- ➌ Linear systems case
- ➍ Closed orbits
- ➎ Case study

# Sommaire

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- 1 Introduction and definitions
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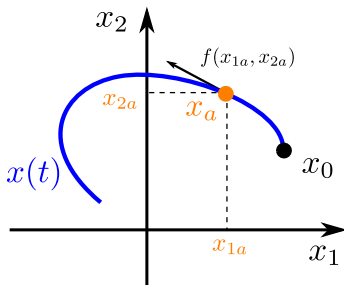
## Second-order systems

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

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

↪ Phase plane



## 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  $x_0$ .

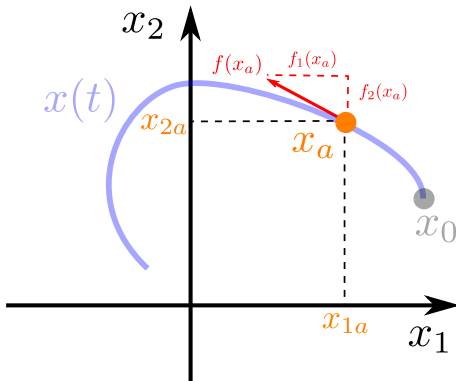
### Vector field

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

## Vector field

The vector  $f(x) = \begin{pmatrix} f_1(x) \\ f_2(x) \end{pmatrix}$  is tangent to the trajectory at point  $x$

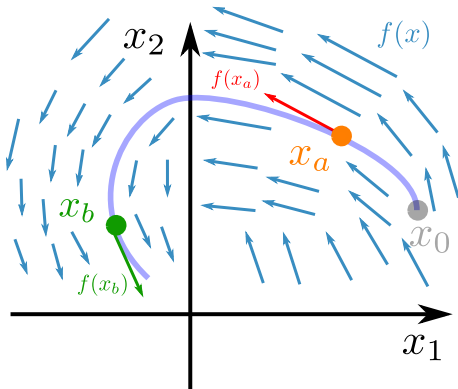
$$\frac{dx_2}{dx_1} = \frac{f_1(x)}{f_2(x)}$$



## Vector field

The vector  $f(x) = \begin{pmatrix} f_1(x) \\ f_2(x) \end{pmatrix}$  is tangent to the trajectory at point  $x$

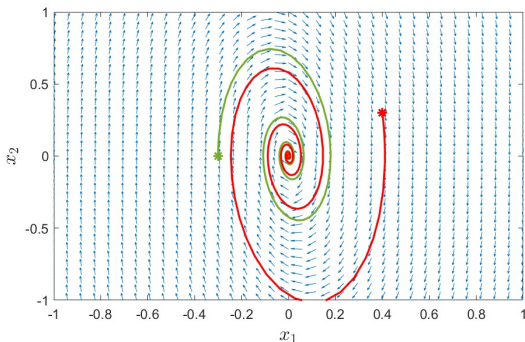
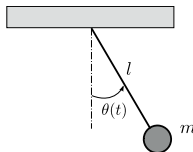
$$\frac{dx_2}{dx_1} = \frac{f_1(x)}{f_2(x)}$$



## Pendulum example

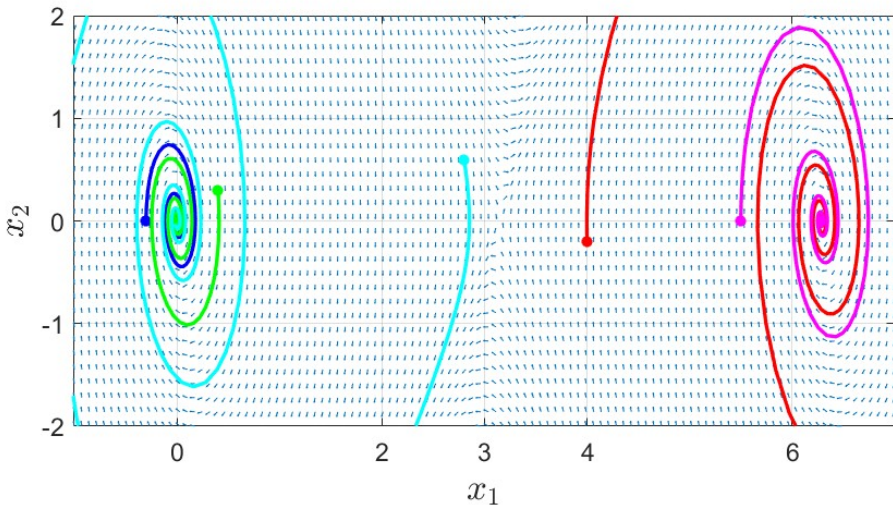
Variables :  $x_1 = \theta$  and  $x_2 = \dot{\theta}$

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





## Pendulum example



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- 1 Introduction and definitions
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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)

- ◇ Theses methods are restricted to quite simple/particular nonlinearities

## Example

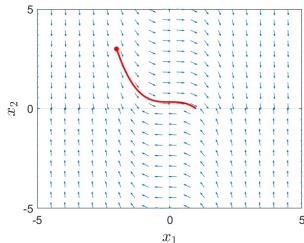
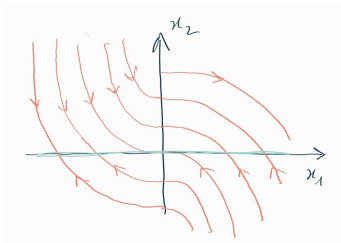
Consider the system

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

- ▶ 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  $\alpha$

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

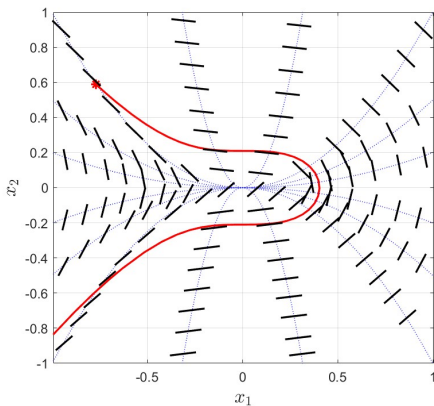
Step :

- ▶ 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

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

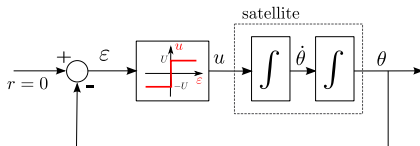
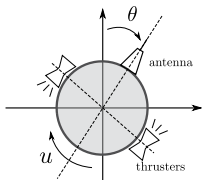
$$\text{slope : } \alpha = \frac{f_2(x)}{f_1(x)} = \frac{-x_1^2}{x_2} \Leftrightarrow x_2 = -\frac{1}{\alpha} x_1^2$$



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.



**Solution :**

## Exercise (isocline method)

Consider the previous (controlled) system

- ▶ Apply the isocline method to retrieve the phase portrait

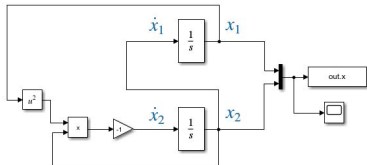
**Solution :**

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

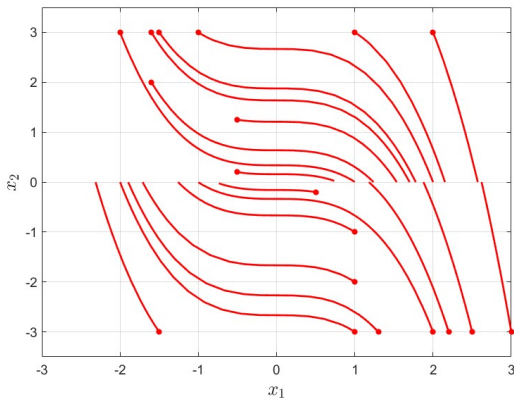
```
x0 = [-2;3];
[t,x] = ode45(f,[0 20],x0);
x1 = x(:,1);
x2 = x(:,2);
plot(x1,x2);
plot(x1(1),x2(2),'*');
```



- ▶ Repeat from step 2

## Numerical simulations

Resulting plot for several  $x_0$



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

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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 :  $Mz = x$

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

► 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

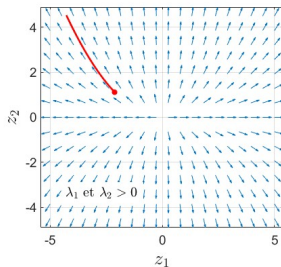
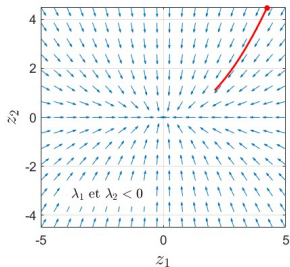
$$\begin{cases} \dot{z}_1 = \lambda_1 z_1 \\ \dot{z}_2 = \lambda_2 z_2 \end{cases} \Rightarrow \begin{cases} z_1(t) = z_{10} e^{\lambda_1 t} \\ z_2(t) = z_{20} e^{\lambda_2 t} \end{cases}$$

- ▶ Eliminate the time  $t$

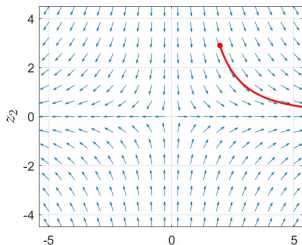
$$z_2 = c z_1^{\lambda_2/\lambda_1} \quad \text{with } c = \frac{z_{20}}{z_{10}^{\lambda_2/\lambda_1}}$$

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**



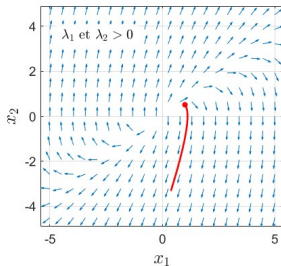
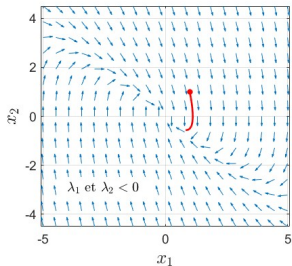
- ▶ Opposite signs  $\Rightarrow$  the equilibrium point is a **saddle point**



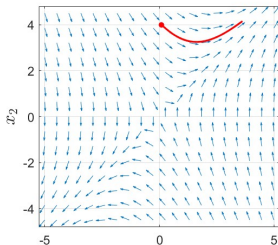


Back in the  $x$ -coordinates basis :  $x = Mz$

- ▶ 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

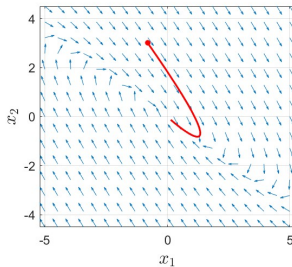
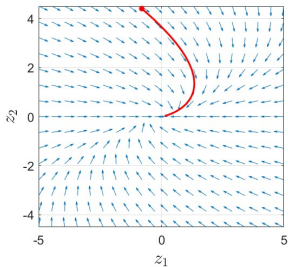
$$\begin{cases} \dot{z}_1 = \lambda z_1 + k z_2 \\ \dot{z}_2 = \lambda z_2 \end{cases} \quad \Rightarrow \quad \begin{cases} z_1(t) = (z_{10} + k z_{20} t) e^{\lambda t} \\ z_2(t) = z_{20} e^{\lambda t} \end{cases}$$

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

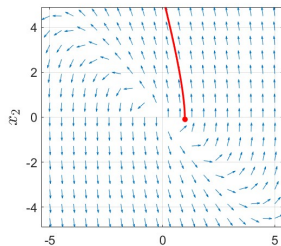
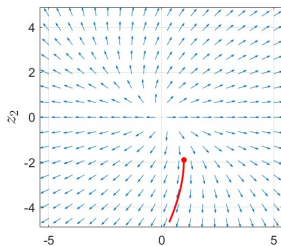
$$z_1 = z_2 \left( \frac{z_{10}}{z_{20}} + \frac{k}{\lambda} \ln \left( \frac{z_2}{z_{20}} \right) \right) \quad \text{and also} \quad \frac{dz_2}{dz_1} = \frac{\lambda z_2}{\lambda z_1 + k z_2}$$

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

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

► Change of basis matrix with  $M = \begin{bmatrix} \text{Re}[v_1] & \text{Im}[v_1] \end{bmatrix}$

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

► Change of variable → polar coordinates :  $z_1 = r \cos \theta$  and  $z_2 = r \sin \theta$

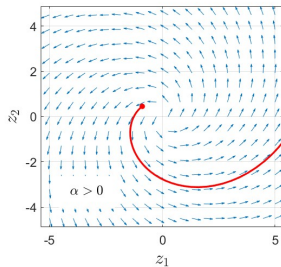
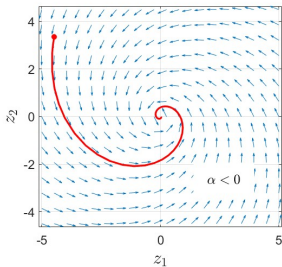
$$\begin{cases} \dot{r} = \alpha r \\ \dot{\theta} = -\beta \end{cases}$$

► that has for solution :

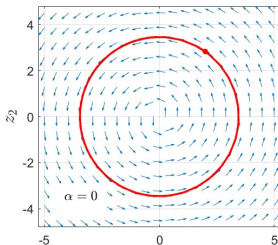
$$\begin{cases} r(t) = r_0 e^{\alpha t} \\ \theta(t) = -\beta t + \theta_0 \end{cases} \quad \text{with} \quad \begin{cases} r_0 = \sqrt{z_{10}^2 + z_{20}^2} \\ \theta_0 = \arctan \frac{z_{20}}{z_{10}} \end{cases}$$

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**

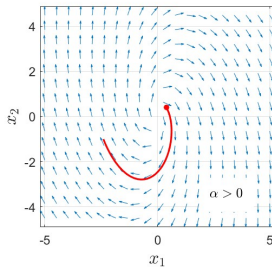
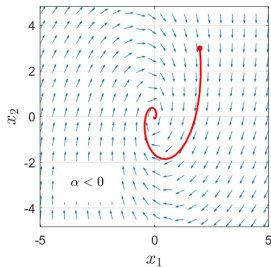


- ▶ Pure imaginary  $\Rightarrow$  the equilibrium point is a **center** (circle of radius  $r_0$ )

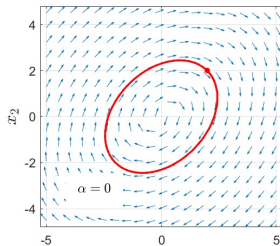


Back in the  $x$ -coordinates basis :  $x = Mz$

- ▶ 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

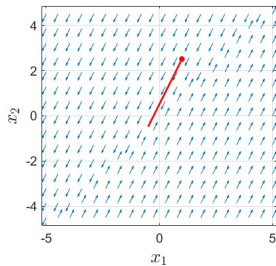
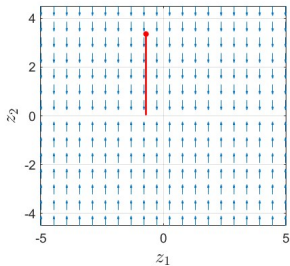
**Second case** :  $\lambda_1 = \lambda_2 = 0$

- ▶ Change of basis gives

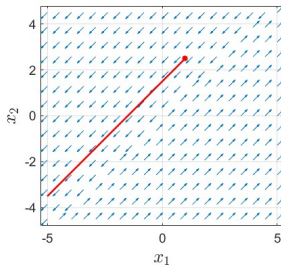
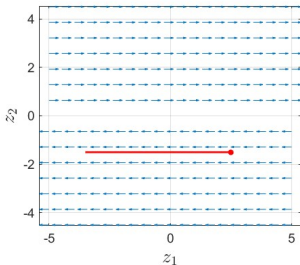
$$\begin{cases} \dot{z}_1 = z_2 \\ \dot{z}_2 = 0 \end{cases} \Rightarrow \begin{cases} z_1(t) = z_{10} + z_{20} t \\ z_2(t) = z_{20} \end{cases}$$

- ▶  $z_1$  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

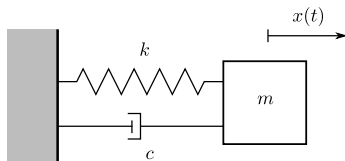
## Example : simple mass-spring system

Equation of motion :

mass ( $m = 1 \text{ kg}$ )

spring (stiffness :  $k = 1 \text{ N/m}$ )

damper (viscous coefficient :  $c \text{ N/m/s}$ )

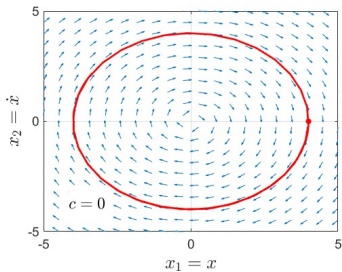
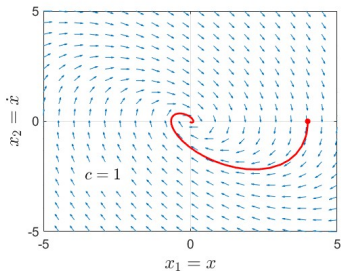
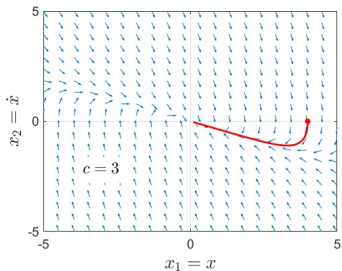


$$\ddot{x} + c\dot{x} + x = 0 \quad \Rightarrow \quad \begin{bmatrix} \dot{x} \\ \ddot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & -c \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} \quad \text{with } \begin{cases} x(0) = x_0 \\ \dot{x}(0) = 0 \end{cases}$$

Eigenvalues of the dynamic matrix

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

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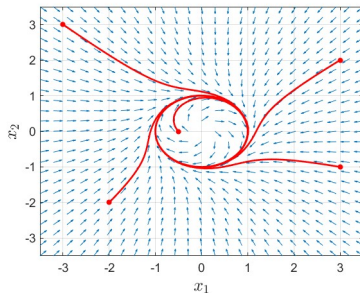
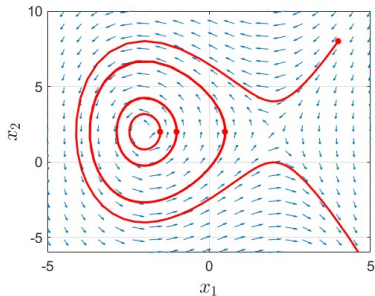


## Closed orbits

A closed orbit is a periodic trajectory

Two cases can be distinguished :

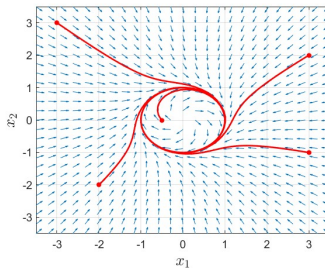
- ▶ **non-isolated** : there are other closed curves in the neighborhood, depend on initial conditions (left)
- ▶ **isolated** : from initial conditions in the neighborhood, trajectories converge or diverge from it → limit cycle (right)



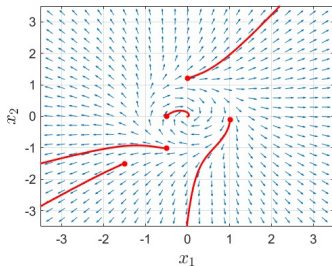
## Limit cycles

Three kinds of limit cycle can be observed

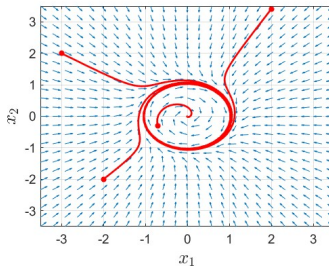
**Stable limit cycle**



**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)

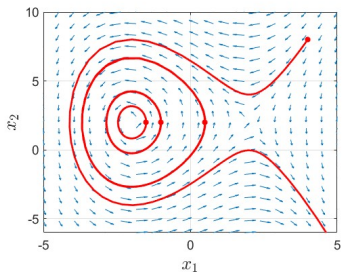
If a trajectory remains in a closed bounded region  $\mathcal{D}$  in the phase plane, then one of the following is true :

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

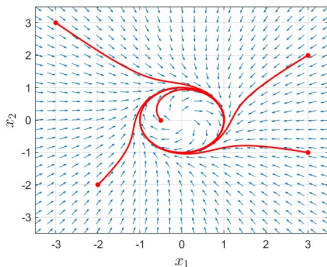
↔ Asymptotic properties of trajectories

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

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

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

# Sommaire

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- 2 Construction of phase portrait
- 3 Linear systems case
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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

## Solution



## 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)