

# Nonlinear physics, dynamical systems and chaos theory

Jean-Christophe Loiseau

*jean-christophe.loiseau@ensam.eu*  
DynFluid,  
Arts et Métiers ParisTech, France

# First-order systems

Flows on the line

# First-order systems

An apparently simple system

- ▶ Let us consider the following first-order dynamical system

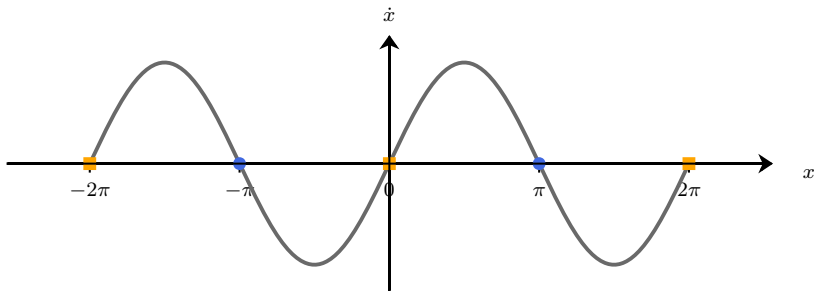
$$\dot{x} = \sin(x).$$

- ▶ Its analytical solution is given by

$$t = \ln \left| \frac{\csc(x_0) + \cot(x_0)}{\csc(x) + \cot(x)} \right|.$$

# First-order systems

## Phase line



Phase line of the first-order dynamical system considered.

# First-order systems

## Fixed points

- ▶ Fixed points  $x^*$  are equilibrium solutions characterized by

$$f(x^*) = 0.$$

- ▶ In the present case, these are given by

$$x^* = n\pi \text{ for } n \in \mathbb{N}.$$

# First-order systems

## Linear stability

- ▶ The dynamics of a perturbation  $\eta(t) = x(t) - x^*$  is given by

$$\dot{\eta} = f(x^* + \eta).$$

- ▶ If  $\eta$  is small enough,  $f(x^* + \eta)$  can be approximated by its first-order Taylor expansion around  $x^*$

$$f(x^* + \eta) = f(x^*) + f'(x^*)\eta + \mathcal{O}(\eta^2).$$

# First-order systems

## Linear stability

- ▶ Given that  $f(x^*) = 0$ , the dynamics of  $\eta$  are governed by

$$\dot{\eta} = f'(x^*)\eta.$$

- ▶ Its analytical solution is given by

$$\eta(t) = \exp(f'(x^*)t) \eta_0.$$

# First-order systems

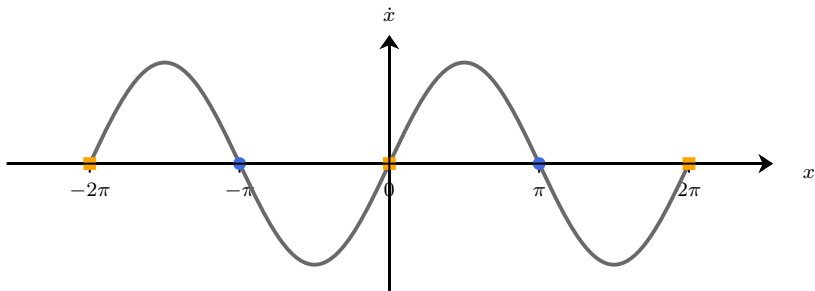
## Linear stability

- ▶ The linear stability of a fixed point  $x^*$  is determined by the sign of  $f'(x^*)$ :
  - ↪ if  $f'(x^*) > 0$ ,  $\eta(t)$  grows exponentially fast. The fixed point is said to be **linearly unstable**.
  - ↪ if  $f'(x^*) < 0$ ,  $\eta(t)$  decays exponentially fast. The fixed point is said to be **linearly stable**.
  - ↪ if  $f'(x^*) = 0$ , one can not conclude and nonlinear analyses are required.
- ▶ Let us now re-analyze our original system and sketch the evolution of  $x(t)$ .



# First-order systems

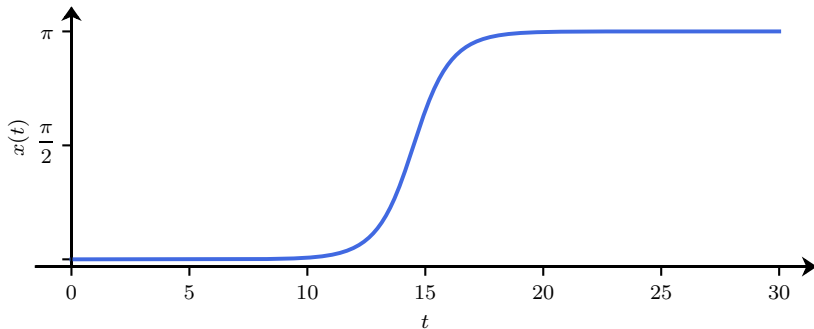
## Phase line



Phase line of the first-order dynamical system considered.

# First-order systems

Evolution of  $x(t)$



Evolution of  $x(t)$  for the initial condition  $x_0 = 10^{-6}$ .

## Warning!

For a first-order system, the trajectories can only vary monotonically: either they end up on a stable fixed point, or they diverge to  $\pm\infty$ .

# Second-order systems

Oscillators, but not only...

# Second-order systems

- ▶ Second-order systems are dynamical systems which can be described by

$$\dot{x} = f(x, y)$$

$$\dot{y} = g(x, y).$$

- ▶ Having two degrees of freedom, they can exhibit a dynamics much richer than simple first-order systems.

# Second-order system

Our working example

- ▶ For the rest of this section, let us consider the following system

$$\begin{aligned}\dot{x} &= x - y^2 + 1.28 + 1.4xy \\ \dot{y} &= 0.2y - x + x^3.\end{aligned}$$

- ▶ Note that this system is considered only for illustration purposes. To the best of my knowledge, it does not model any particular physics.

# Interlude

How to compute fixed points?

# How to compute fixed points?

## Different techniques

- ▶ Fixed points are structuring the phase space of the dynamical system under scrutiny. Unfortunately, it may not be easy (nor possible) to compute them analytically.
- ▶ A number of different numerical techniques exist for that purpose. The following list is by no means exhaustive:
  - ↪ Newton-Raphson method,
  - ↪ Selective Frequency Damping,
  - ↪ BoostConv,
  - ↪ ...



# Newton-Raphson method

$$f : \mathbb{R} \rightarrow \mathbb{R}$$

- ▶ Originally proposed by Isaac Newton (1645–1727) and Joseph Raphson (1648 – 1715) to solve

$$f(x) = 0.$$

- ▶ Given an initial guess  $x_0$ , the idea is to approximate  $f(x)$  by its first-order Taylor expansion around  $x_0$ , i.e.

$$f(x) \simeq f(x_0) + f'(x_0)(x - x_0).$$

- ▶ A better estimate  $x_1$  of the root of  $f$  can then be obtained by solving

$$0 = f(x_0) + f'(x_0)(x_1 - x_0).$$

# Newton-Raphson method

$$f : \mathbb{R} \rightarrow \mathbb{R}$$

- ▶ After  $k$  iterations, the basic iteration scheme can be written as

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}.$$

- ▶ The iterative procedure stops when a user-defined criterion is fulfilled, usually

$$\|f(x_k)\| \leq \epsilon \text{ or } \|x_{k+1} - x_k\| \leq \epsilon,$$

with  $\epsilon \simeq 10^{-10}$ .

# Newton-Raphson method

$$f : \mathbb{R} \rightarrow \mathbb{R}$$

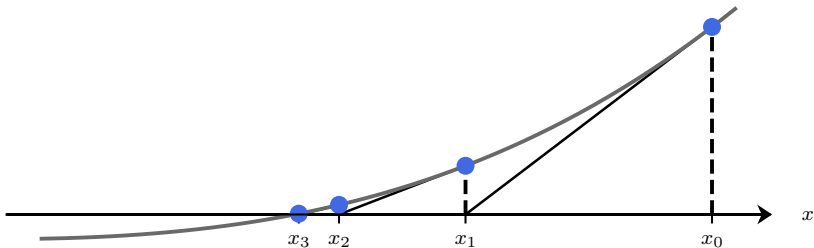


Illustration of the Newton-Raphson for  $f(x) = x^3 - 2x - 5$  and  $x_0 = 3.8$ .

# Newton-Raphson method

$$\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$$

- ▶ Generalization of the Newton-Raphson method to the case  $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is quite straightforward.
- ▶ Given an estimate  $\mathbf{x}_k$ , the basic iteration reads

$$\mathbf{J}\delta\mathbf{x} = -\mathbf{f}(\mathbf{x}_k)$$

$$\mathbf{x}_{k+1} = \mathbf{x}_k + \delta\mathbf{x},$$

where  $\mathbf{J}$  is the Jacobian matrix of  $\mathbf{f}$  evaluated at  $\mathbf{x}_k$ .

# Newton-Raphson method

## Limitations

- ▶ Although efficient, Newton-Raphson method suffers from a number of limitations:
  - ↪ The fixed points computed may depend on the initial guess  $x_0$ .
  - ↪ Evaluating  $f(x)$  might be computationally expensive.
  - ↪ At each iteration, the Jacobian matrix  $J$  needs to be evaluated and inverted ( $\mathcal{O}(n^3)$  operations).
- ▶ A number of variants of the Newton-Raphson method exist as to address these different limitations. This however is algorithmic refinement beyond the scope of the present course.

# Second-order systems

Back to our example

