

# Mathematical modelling

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# The dynamics of systems of 2 equations

For an autonomous linear system

$$\dot{x}_1 = a_{11}x_1 + a_{12}x_2, \quad \dot{x}_2 = a_{21}x_1 + a_{22}x_2,$$

the origin  $(0,0)$  is always a stationary point, i.e., an equilibrium solution.

The eigenvalues of the matrix

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

determine the type of the stationary point  $(0,0)$  and the shape of the phase portrait.

We will assume that  $\det A \neq 0$ . Let  $\lambda_1, \lambda_2$  be the eigenvalues of  $A$ . We also assume that there exist two linearly independent vectors  $v_1, v_2$  of  $A$  (even if  $\lambda_1 = \lambda_2$ ).

Case 1:  $\lambda_1, \lambda_2 \in \mathbb{R}$

The general solution is

$$x(t) = C_1 e^{\lambda_1 t} v_1 + C_2 e^{\lambda_2 t} v_2.$$

- ▶ If  $C_1 = 0$ , the trajectory  $x_1(t)$  is a ray in the direction of  $v_2$  if  $C_2 > 0$ , or  $-v_1$  if  $C_2 < 0$ .
- ▶ Similarly, if  $C_2 = 0$  the trajectory  $x_2(t)$  is a ray in the direction of  $v_2$  or  $-v_2$ .
- ▶ The behaviour of other trajectories depends on the signs of  $\lambda_1$  and  $\lambda_2$ .

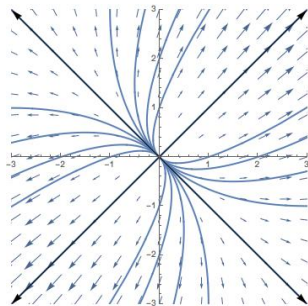
### Subcase 1.1: $0 < \lambda_1 < \lambda_2$

- ▶ as  $t \rightarrow \infty$ ,  $x(t)$  asymptotically approaches the solution  $\pm e^{\lambda_2 t} v_2$ ,
- ▶ as  $t \rightarrow -\infty$ ,  $x(t)$  asymptotically approaches the solution  $\pm e^{\lambda_1 t} v_1$ .

The point  $(0, 0)$  is a source.

**Example.** The general solution of the system  $\dot{x}_1 = 3x_1 + x_2$ ,  $\dot{x}_2 = x_1 + 3x_2$  is

$$x(t) = C_1 e^{4t} \begin{bmatrix} 1 & 1 \end{bmatrix}^T + C_2 e^{2t} \begin{bmatrix} -1 & 1 \end{bmatrix}^T.$$



Algorithm:

[https://zalara.github.io/Algoritmi/phaseportrait\\_1\\_1.m](https://zalara.github.io/Algoritmi/phaseportrait_1_1.m)

### Subcase 1.2: $\lambda_2 < \lambda_1 < 0$

- ▶ as  $t \rightarrow \infty$ ,  $x(t)$  asymptotically approaches the solution  $\pm e^{\lambda_1 t} v_2$ ,
- ▶ as  $t \rightarrow -\infty$ ,  $x(t)$  asymptotically approaches the solution  $\pm e^{\lambda_2 t} v_1$ .

The point  $(0, 0)$  is a sink.

**Example.** The general solution of the system  $\dot{x}_1 = -3x_1 - x_2$ ,  $\dot{x}_2 = -x_1 - 3x_2$  is

$$x(t) = C_1 e^{-4t} \begin{bmatrix} 1 & 1 \end{bmatrix}^T + C_2 e^{-2t} \begin{bmatrix} -1 & 1 \end{bmatrix}^T.$$

Algorithm:

[https://zalara.github.io/Algoritmi/phaseportrait\\_1\\_2.m](https://zalara.github.io/Algoritmi/phaseportrait_1_2.m)

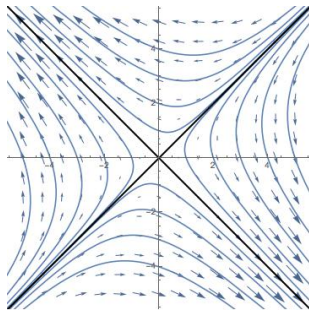
### Subcase 1.3: $\lambda_1 < 0 < \lambda_2$

- ▶ as  $t \rightarrow \infty$ ,  $x(t)$  asymptotically approaches the solution  $\pm e^{\lambda_2 t} v_2$ ,
- ▶ as  $t \rightarrow -\infty$ ,  $x(t)$  asymptotically approaches the solution  $\pm e^{\lambda_1 t} v_1$ .

The point  $(0, 0)$  is a saddle.

**Example.** The general solution of the system  $\dot{x}_1 = x_1 - 3x_2$ ,  $\dot{x}_2 = -3x_1 + x_2$  is

$$x(t) = C_1 e^{-2t} \begin{bmatrix} 1 & 1 \end{bmatrix}^T + C_2 e^{4t} \begin{bmatrix} 1 & -1 \end{bmatrix}^T.$$



Algorithm:

[https://zalara.github.io/Algoritmi/phaseportrait\\_1\\_3.m](https://zalara.github.io/Algoritmi/phaseportrait_1_3.m)

### Subcase 2.1: $\lambda_{1,2} = \alpha \pm i\beta$ , $\alpha \neq 0$

The general solution is

$$x(t) = e^{\alpha t} [(C_1 \cos(\beta t) + C_2 \sin(\beta t))u + (-C_1 \sin(\beta t) + C_2 \cos(\beta t))w].$$

Hence,

- ▶ if  $\alpha < 0$ ,  $x(t)$  spirals towards  $(0, 0)$  as  $t \rightarrow \infty$ , and
- ▶ if  $\alpha > 0$ ,  $x(t)$  spirals away from  $(0, 0)$  as  $t \rightarrow \infty$ .

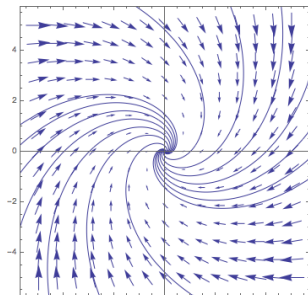
The point  $(0, 0)$  is a spiral sink in the first case and a spiral source in the second case.

### Example

$$\dot{x}_1 = -3x_1 + 2x_2, \quad \dot{x}_2 = -x_1 - x_2$$

$$x(t) = e^{-2t}.$$

$$\left( (C_1 \cos t + C_2 \sin t) \begin{bmatrix} 2 \\ 1 \end{bmatrix} + (-C_1 \sin t + C_2 \cos t) \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right)$$



### Subcase 2.2: $\lambda_{1,2} = \pm i\beta$ , $\alpha \neq 0$

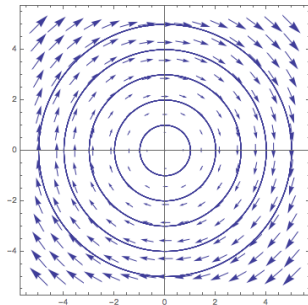
The trajectories are periodic with period  $2\pi/\beta$ , i.e. the point  $x(t)$  circles around  $(0, 0)$ .

The point  $(0, 0)$  is a center.

### Example

$$\dot{x} = v, \quad \dot{v} = -\omega^2 x$$

$$x(t) = (C_1 \cos(\omega t) + C_2 \sin(\omega t)) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (-C_1 \sin(\omega t) + C_2 \cos(\omega t)) \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$



Algorithm:

[https://zalara.github.io/Algoritmi/phaseportrait\\_2\\_1.m](https://zalara.github.io/Algoritmi/phaseportrait_2_1.m)

[https://zalara.github.io/Algoritmi/phaseportrait\\_2\\_2.m](https://zalara.github.io/Algoritmi/phaseportrait_2_2.m)



# Nonlinear autonomous systems of equations

$$\dot{x}_1 = f_1(x_1, x_2), \quad \dot{x}_2 = f_2(x_1, x_2)$$

If  $a = (a_1, a_2)$  is a critical point, that is,

$$f_1(a_1, a_2) = f_2(a_1, a_2) = 0,$$

then the behaviour of trajectories close to  $a$  is approximated by trajectories of the [linearization](#) of the system at the point  $a$ :

$$\dot{x}_1 \doteq \frac{\partial f_1}{\partial x_1}(x_1 - a_1) + \frac{\partial f_1}{\partial x_2}(x_2 - a_2), \quad \dot{x}_2 \doteq \frac{\partial f_2}{\partial x_1}(x_1 - a_1) + \frac{\partial f_2}{\partial x_2}(x_2 - a_2).$$

This is a linear homogeneous system with coefficient matrix the Jacobian matrix of the vector function  $f(x)$ :

$$\dot{x} \doteq Df(a)(x - a) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix} (x - a).$$

The critical point is classified as a [source](#), [sink](#), [saddle](#), [spiral source](#), [spiral sink](#) or [center](#) depending on the eigenvalues of  $Df(a)$ .

In addition to critical points, that is, equilibrium solutions, a plane nonlinear system (that is, a nonlinear system of two differential equations) can also have [limit cycles](#).

A limit cycle is a periodic solutions  $x_\infty(t)$  such that for initial conditions  $x(t_0) = x_0$  in a certain domain the corresponding solutions  $x(t)$

- ▶ either asymptotically tend towards  $x_\infty(t)$  as  $t \rightarrow \infty$  – in this case  $x_\infty$  is an [attracting limit cycle](#), or
- ▶  $x(t) \rightarrow x_\infty(t)$  as  $t \rightarrow -\infty$  – in this case  $x_\infty$  is a [repelling limit cycle](#).

Systems of more than two differential equations can exhibit much more complex, chaotic behaviour.

Algorithm:

[https://zalara.github.io/Algoritmi/example\\_predator\\_preym\\_linearization.m](https://zalara.github.io/Algoritmi/example_predator_preym_linearization.m)

## Differential equations of order 2

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

The general solution is a two-parametric family

$$x = x(t, C_1, C_2).$$

A particular solution is given by specifying

- ▶ initial conditions:  $x(t_0) = \alpha_0$ ,  $\dot{x}(t_0) = \alpha_1$ ,  
where the values of the solution and its derivative are given at some  
initial time  $t_0$   
or
- ▶ boundary conditions:  $x(a) = x_0$ ,  $x(b) = x_1$   
where values of the solution at different times  $a, b$  are given (i.e., on  
the boundary of some interval  $[a, b]$ )

## Differential equations of order $n$

$$x^{(n)} = f(t, x, \dot{x}, \dots, x^{(n-1)})$$

The general solution is an  $n$ -parametric family

$$x = x(t, C_1, \dots, C_n).$$

A particular solution is given by

- ▶ initial conditions:  $x(t_0) = \alpha_0, \dots, x^{(n-1)}(t_0) = \alpha_{n-1}$   
where the values of the solution and its first  $(n-1)$  derivatives are given at some initial time  $t_0$   
or
- ▶ boundary conditions  
where values of the solution or its derivatives are given in different times.

## Linear DE's of order $n$

A linear DE (LDE) of degree  $n$  is of the form

$$x^{(n)} + a_{n-1}(t)x^{(n-1)} + \cdots + a_0(t)x = f(t). \quad (1)$$

The equation is

- ▶ homogeneous if  $f(t) = 0$ , and
- ▶ nonhomogeneous if  $f(t) \neq 0$ .
- ▶ The general solution of the homogeneous part is the family of all linear combinations

$$y(t) = C_1x_1(t) + \cdots + C_nx_n(t)$$

of  $n$  linearly independent solutions  $x_1(t), \dots, x_n(t)$ .

- ▶ If the coefficients  $a_1(t), \dots, a_n(t)$  are continuous functions, then for any  $\alpha_0, \dots, \alpha_n$  there exists exactly one solution satisfying the initial condition

$$x(t_0) = \alpha_0, \quad \dot{x}(t_0) = \alpha_1, \quad \dots, \quad x^{(n-1)}(t_0) = \alpha_n.$$

## LDEs with constant coefficients

Assume that the coefficient functions  $a_1(t), \dots, a_n(t)$  in a homogeneous LDE are constant:

$$x^{(n)} + a_{n-1}x^{(n-1)} + \dots + a_0x = 0, \quad a_1, \dots, a_n \in \mathbb{R} \quad (2)$$

Translating (2) to the system by the usual trick of introducing new variables

$$x_1 = x, \quad x_2 = x'_1, \quad x_3 = x'_2, \quad \dots, \quad x_n = x'_{n-1},$$

(2) becomes

$$x'_n = -a_0x_1 - a_1x_2 - \dots - a_{n-1}x_n,$$

or matrixially  $\vec{x}' = A\vec{x}$ :

$$\vec{x}'(t) = \begin{bmatrix} 0 & 1 & 0 & \dots & & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 \\ \vdots & & \ddots & \ddots & & \vdots \\ \vdots & & & \ddots & \ddots & \vdots \\ 0 & \dots & & \dots & 0 & 1 \\ -a_0 & -a_1 & \dots & \dots & -a_{n-2} & -a_{n-1} \end{bmatrix} \vec{x}(t)$$

- ▶ The solutions to this system are of the form

$$x(t) = p_k(t)e^{\lambda t}v,$$

where  $\lambda$  is the eigenvalue of  $A$ ,  $p_k(t)$  is a polynomial of degree  $k$  in  $t$  and  $v$  is the generalized eigenvector. (This follows most easily by the use of the Jordan form of the matrix.)

- ▶ In particular, if there are  $n$  linearly independent eigenvectors of the matrix  $A$ , then all polynomials  $p_k$  are constants and generalized eigenvectors are usual eigenvectors.
- ▶ By a simple calculation of expressing the determinant of  $A - \lambda I$  according to the coefficients and cofactors of the last row, it turns out that the eigenvalues of  $A$  are precisely the roots of the [characteristic polynomial](#) corresponding to (2):

$$P(\lambda) := \lambda^n + a_{n-1}\lambda^{n-1} + \cdots a_1\lambda + a_0. \quad (3)$$

- ▶ A (trivial) fact with a nontrivial proof, called the [fundamental theorem of algebra](#), states that a polynomial of degree  $n$  has exactly  $n$  roots, counted by multiplicity. In case the matrix  $A$  is real, these roots are real or complex conjugate pairs.

- ▶ From the roots of the characteristic polynomial (3),  $n$  linearly independent solutions of the LDE can be reconstructed.
- ▶ For every real root  $\lambda \in \mathbb{R}$ ,

$$x(t) = e^{\lambda t}$$

is a solution of the homogeneous LDE.

- ▶ For a complex conjugate pair of roots  $\lambda = \alpha \pm i\beta$ , the real and imaginary parts of the complex-valued exponential functions

$$e^{(\alpha \pm i\beta)t} = e^{\alpha t}(\cos(\beta t) + i \sin(\beta t))$$

are two linearly independent solutions

$$x_1 = e^{\alpha t} \cos(\beta t), \quad x_2 = e^{\alpha t} \sin(\beta t).$$

## Proposition

*If a root (or a complex pair of roots)  $\lambda$  has multiplicity  $k > 1$ , then it can be shown that*

$$e^{\lambda t}, te^{\lambda t}, \dots, t^{k-1}e^{\lambda t}$$

*are all linearly independent solutions.*



## Proof of proposition

Let us prove the last fact by an interesting trick. We introduce the operator

$$L : \mathcal{C}^{(n)}(I) \rightarrow \mathcal{C}(I)$$

$$L(u) = u^{(n)} + a_{n-1}u^{(n-1)} \dots + a_0u,$$

where  $\mathcal{C}^{(n)}(I)$  stands for the vector space of  $n$ -times continuously differentiable functions on the interval  $I$  and  $\mathcal{C}(I)$  stands for the vector space of continuous functions on  $I$ .

Let  $\lambda_0$  be the root of the characteristic polynomial (3) of multiplicity  $k$ , i.e.,

$$P(\lambda) = (\lambda - \lambda_0)^k Q(\lambda).$$

Let  $0 \leq q \leq k$  by an integer. We will check that  $t^q e^{\lambda t}$  solves (2).

Notice that

$$t^q e^{\lambda t} = \frac{d^q}{d\lambda^q} e^{\lambda t}.$$

For ease of notation we define  $a_n := 1$ . We have that:

$$\begin{aligned} L(t^q e^{\lambda t}) &= \sum_{i=0}^n a_i \left( \frac{d^q}{d\lambda^q} e^{\lambda t} \right)^{(i)} = \sum_{i=0}^n a_i \frac{d^i}{dt^i} \left( \frac{d^q}{d\lambda^q} e^{\lambda t} \right) \\ &= \frac{d^q}{d\lambda^q} \left( \sum_{i=0}^n a_i \frac{d^i}{dt^i} e^{\lambda t} \right) = \frac{d^q}{d\lambda^q} \left( \sum_{i=0}^n a_i \lambda^i e^{\lambda t} \right) = \frac{d^q}{d\lambda^q} (P(\lambda) e^{\lambda t}). \end{aligned}$$

Since

$$\frac{d^q}{d\lambda^q}(P(\lambda)e^{\lambda t}) = \sum_{i=1}^q \frac{d^i}{d\lambda^i}(P(\lambda)) \cdot Q_i(t, \lambda),$$

where  $Q_i(t, \lambda)$  are functions of  $t$ ,  $\lambda$  and

$$\frac{d^i}{d\lambda^i}(P(\lambda_0)) = 0, \quad \text{for } i = 0, \dots, q,$$

it follows that

$$L(t^q e^{\lambda_0 t}) = 0.$$