Linear differential equations

The solution of dx(t)/dt = ax(t) is $x(t) = Ce^{at}$, where C = x(0). Check this:

$$\partial_t C e^{at} = a C e^{at} = ax(t)$$

Now two-dimensional systems:

$$\begin{cases} \mathrm{d}x/\mathrm{d}t = f(x,y) \\ \mathrm{d}y/\mathrm{d}t = g(x,y) \end{cases}$$

where x(t) and y(t) are unknown functions of time t, and f and g are functions of x and y.

An example:

$$\begin{cases} dx/dt = ax + by \\ dy/dt = cx + dy \end{cases} \text{ and } \begin{cases} dx/dt = -2x + y \\ dy/dt = x - 2y \end{cases}$$

where x and y decay at a rate -1, and are converted into one another at a rate 1. Steady state x = y = 0.

In matrix notation:

$$\begin{pmatrix} \mathrm{d}x/\mathrm{d}t\\ \mathrm{d}y/\mathrm{d}t \end{pmatrix} = \begin{pmatrix} a & b\\ c & d \end{pmatrix} \begin{pmatrix} x\\ y \end{pmatrix}$$

We claim that
$$\begin{pmatrix} dx/dt \\ dy/dt \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$
 has as a general solution:

$$x(t) = C_1 x_1 e^{\lambda_1 t} + C_2 x_2 e^{\lambda_2 t}$$

$$y(t) = C_1 y_1 e^{\lambda_1 t} + C_2 y_2 e^{\lambda_2 t}$$

or

$$\begin{pmatrix} x(t) \\ y(t) \end{pmatrix} = C_1 \begin{pmatrix} x_1 \\ y_1 \end{pmatrix} e^{\lambda_1 t} + C_2 \begin{pmatrix} x_2 \\ y_2 \end{pmatrix} e^{\lambda_2 t}$$

where $\lambda_{1,2}$ are eigenvalues and $(x_i \ y_i)$ are the corresponding eigenvectors of the matrix given above.

Like
$$x(t) = Ce^{at}$$
, this has only one steady state: $(x, y) = (0, 0)$.

Notice that the solutions $\begin{pmatrix} x(t) \\ y(t) \end{pmatrix} = C_1 \begin{pmatrix} x_1 \\ y_1 \end{pmatrix} e^{\lambda_1 t} + C_2 \begin{pmatrix} x_2 \\ y_2 \end{pmatrix} e^{\lambda_2 t}$ are a linear combination of the growth along the eigenvectors.

Since x(t) and y(t) grow when $\lambda_{1,2} > 0$ we obtain:

- a stable node when both $\lambda_{1,2} < 0$
- an unstable node when both $\lambda_{1,2} > 0$
- an (unstable) saddle point when $\lambda_1 > 0$ and $\lambda_2 < 0$ (or vice versa)

When $\lambda_{1,2}$ are complex, i.e., $\lambda_{1,2} = \alpha \pm i\beta$, we obtain

- a stable spiral when the real part $\alpha < 0$
- an unstable spiral when the real part $\alpha > 0$
- a neutrally stable center point when the real part $\alpha = 0$

Example:
$$\begin{pmatrix} dx/dt \\ dy/dt \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -2 & 1 \\ 1 & -2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

Since tr = -4 and det = 4 - 1 = 3 we obtain:

$$\lambda_{1,2} = \frac{-4 \pm \sqrt{16 - 12}}{2} = -2 \pm 1$$

so $\lambda_1 = -1$ and $\lambda_2 = -3$.

Hence solutions tend to zero and (x, y) = (0, 0) is a stable node.

To find the eigenvector $\vec{v_1}$ we write:

$$\vec{v_1} = \begin{pmatrix} -b\\ a-\lambda_1 \end{pmatrix} = \begin{pmatrix} -1\\ -1 \end{pmatrix} \quad \text{or} \quad \vec{v_1} \begin{pmatrix} 1\\ 1 \end{pmatrix}$$

For $\vec{v_2}$ we write

$$\vec{v_2} = \begin{pmatrix} -b\\ a - \lambda_2 \end{pmatrix} = \begin{pmatrix} -1\\ 1 \end{pmatrix}$$

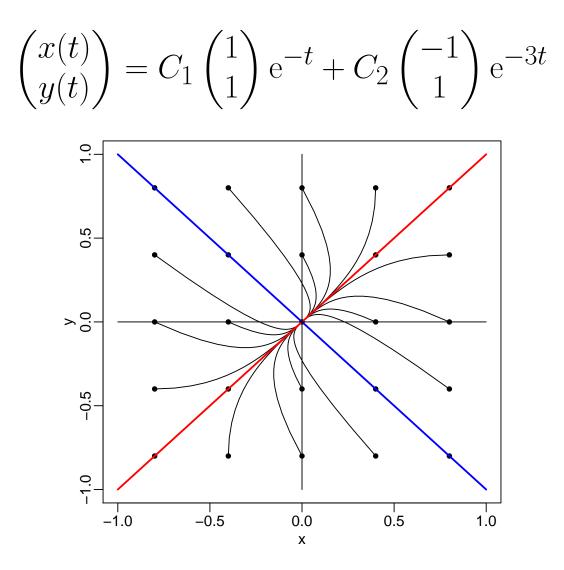
In combination this gives

$$\begin{pmatrix} x(t) \\ y(t) \end{pmatrix} = C_1 \begin{pmatrix} 1 \\ 1 \end{pmatrix} e^{-t} + C_2 \begin{pmatrix} -1 \\ 1 \end{pmatrix} e^{-3t}$$

or

$$\begin{aligned} x(t) &= C_1 e^{-t} - C_2 e^{-3t} \\ y(t) &= C_1 e^{-t} + C_2 e^{-3t} \end{aligned}$$

The integration constants C_1 and C_2 can be solved from the initial condition: i.e., $x(0) = C_1 - C_2$ and $y(0) = C_1 + C_2$.



Finally let's check this solution:

$$\begin{aligned} x(t) &= C_1 e^{-t} - C_2 e^{-3t} \\ y(t) &= C_1 e^{-t} + C_2 e^{-3t} \end{aligned}$$

or

$$\frac{\mathrm{d}x}{\mathrm{d}t} = -C_1 \mathrm{e}^{-t} + 3C_2 \mathrm{e}^{-3t}$$
$$\frac{\mathrm{d}y}{\mathrm{d}t} = -C_1 \mathrm{e}^{-t} - 3C_2 \mathrm{e}^{-3t}$$

which should be equal to

$$\frac{\mathrm{d}x}{\mathrm{d}t} = -2x + y = -2(C_1 \mathrm{e}^{-t} - C_2 \mathrm{e}^{-3t}) + C_1 \mathrm{e}^{-t} + C_2 \mathrm{e}^{-3t} = -C_1 \mathrm{e}^{-t} + 3C_2 \mathrm{e}^{-3t}$$
$$\frac{\mathrm{d}y}{\mathrm{d}t} = x - 2y = C_1 \mathrm{e}^{-t} - C_2 \mathrm{e}^{-3t} - 2(C_1 \mathrm{e}^{-t} + C_2 \mathrm{e}^{-3t}) = -C_1 \mathrm{e}^{-t} - 3C_2 \mathrm{e}^{-3t}$$

Linear approximations

Derivative:

$$f'(\bar{x}) = \lim_{x \to \bar{x}} \frac{f(x) - f(\bar{x})}{x - \bar{x}}$$
 or $f'(\bar{x}) = \lim_{h \to 0} \frac{f(\bar{x} + h) - f(\bar{x})}{h}$,

Rewrite this into:

$$f(x) \simeq f(\bar{x}) + f'(\bar{x}) (x - \bar{x})$$
 or $f(x) \simeq f(\bar{x}) + f'(\bar{x}) h$,

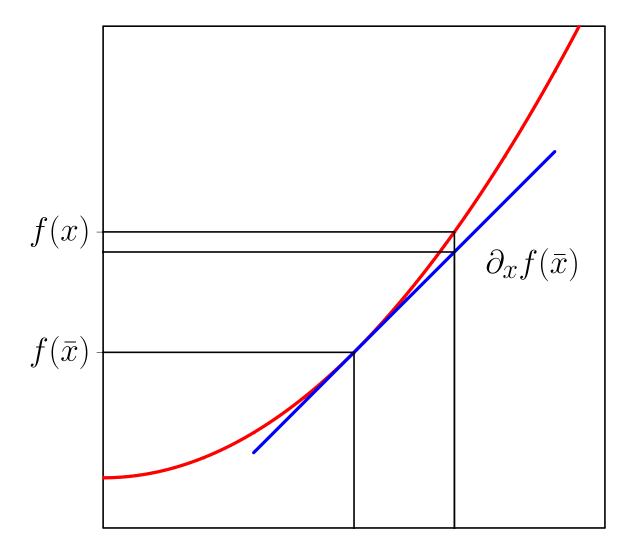
Example:

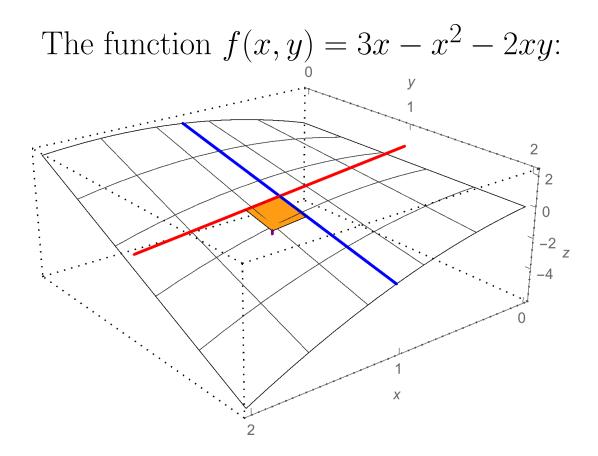
$$f(x) = ax^{2} + b \quad \rightarrow \quad \partial_{x}f(x) = 2ax$$

$$a = 2, b = 1, x = 3 \quad \rightarrow \quad f(3) = 2 \times 9 + 1 = 19, \\ \partial_{x}f(3) = 2 \times 2 \times 3 = 12$$

$$f(3.1) = 20.22 \quad \text{or} \quad f(3.1) \simeq f(3) + \partial_{x}f(3) \times 0.1 = 19 + 12 \times 0.1 = 20.22$$

$$f(x) \simeq f(\bar{x}) + \partial_x f(\bar{x}) \ (x - \bar{x})$$





 $\partial_x f(x, y) = 3 - 2x - 2y$ and $\partial_y f(x, y) = -2x$ and in the point f(1, 1) = 0: $\partial_x f(x, y) = -1$ and $\partial_y f(x, y) = -2$

Generally

$$f(x,y) \simeq f(\bar{x},\bar{y}) + \partial_x f(x-\bar{x}) + \partial_y f(y-\bar{y})$$

Or, after defining $h_x = x - \bar{x}$ and $h_y = y - \bar{y}$:

$$f(x,y) = f(\bar{x} + h_x, \bar{y} + h_y) \simeq f(\bar{x}, \bar{y}) + \partial_x f h_x + \partial_y f h_y$$

Example:

$$f(x,y) = 3x - x^2 - 2xy , \quad f(1,1) = 0 , \quad \partial_x = -1, \partial_y = -2$$
$$f(1.25, 1.25) = 3.75 - 1.5625 - 3.125 = -0.9375$$
$$f(1.25, 1.25) \simeq 0 - 1 \times 0.25 - 2 \times 0.25 = -0.75$$

Consider

$$\begin{cases} \mathrm{d}x/\mathrm{d}t = f(x,y) \\ \mathrm{d}y/\mathrm{d}t = g(x,y) \end{cases}$$

close an equilibrium point at (\bar{x}, \bar{y}) , i.e., $f(\bar{x}, \bar{y}) = g(\bar{x}, \bar{y}) = 0$ Linear approximation of f(x, y) close to the equilibrium: $f(x, y) \simeq f(\bar{x}, \bar{y}) + \partial_x f(x - \bar{x}) + \partial_y f(y - \bar{y})$

As $f(\bar{x}, \bar{y}) = 0$ we obtain $f(x, y) \simeq \partial_x f(x - \bar{x}) + \partial_y f(y - \bar{y})$

For g(x, y) this yields:

$$g(x,y) \simeq \partial_x g \ (x-\bar{x}) + \partial_y g \ (y-\bar{y})$$

$$\begin{cases} dx/dt = f(x,y) \\ dy/dt = g(x,y) \end{cases} \text{ became } \begin{cases} dx/dt \simeq \partial_x f(x-\bar{x}) + \partial_y f(y-\bar{y}) \\ dy/dt \simeq \partial_x g(x-\bar{x}) + \partial_y g(y-\bar{y}) \end{cases}$$

Since the partial derivatives are merely the slopes of f(x, y) and g(x, y) at the point (\bar{x}, \bar{y}) , they are constants that we can write as

$$a = \partial_x f, \quad b = \partial_y f, \quad c = \partial_x g, \quad d = \partial_y g$$

Steady states \bar{x} and \bar{y} are also constants, with derivatives zero:

$$\frac{\mathrm{d}x}{\mathrm{d}t} = \frac{\mathrm{d}x}{\mathrm{d}t} - \frac{\mathrm{d}\bar{x}}{\mathrm{d}t} = \frac{\mathrm{d}(x - \bar{x})}{\mathrm{d}t} \quad \text{and} \quad \frac{\mathrm{d}y}{\mathrm{d}t} = \frac{\mathrm{d}y}{\mathrm{d}t} - \frac{\mathrm{d}\bar{y}}{\mathrm{d}t} = \frac{\mathrm{d}(y - \bar{y})}{\mathrm{d}t}$$

Hence

$$\begin{cases} \mathrm{d}(x-\bar{x})/\mathrm{d}t = a(x-\bar{x}) + b(y-\bar{y}) \\ \mathrm{d}(y-\bar{y})/\mathrm{d}t = c(x-\bar{x}) + d(y-\bar{y}) \end{cases}$$

Changing variables to the distances $h_x = x - \bar{x}$ and $h_y = y - \bar{y}$: $\begin{cases} dh_x/dt = ah_x + bh_y \\ dh_y/dt = ch_x + dh_y \end{cases}$

having the solution

$$\begin{pmatrix} h_x(t) \\ h_y(t) \end{pmatrix} = C_1 \begin{pmatrix} x_1 \\ y_1 \end{pmatrix} e^{\lambda_1 t} + C_2 \begin{pmatrix} x_2 \\ y_2 \end{pmatrix} e^{\lambda_2 t}$$

where $\lambda_{1,2}$ and $(x_i \ y_i)$ are the eigenvalues and corresponding eigenvectors of the Jacobi matrix

$$J = \begin{pmatrix} \partial_x f & \partial_y f \\ \partial_x g & \partial_y g \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

Knowing the two eigenvalues of

$$J = \begin{pmatrix} \partial_x f \ \partial_y f \\ \partial_x g \ \partial_y g \end{pmatrix} = \begin{pmatrix} a \ b \\ c \ d \end{pmatrix}$$

the steady state will be stable when $\lambda_1 < 0$ and $\lambda_2 < 0$.

If so the return time is defined by the largest eigenvalue:

$$T_R = \frac{-1}{\max(\lambda_1, \lambda_2)}$$

Example:

$$\frac{\mathrm{d}x}{\mathrm{d}t} = f(x, y) = a - bx - cxy \quad \text{and} \quad \frac{\mathrm{d}y}{\mathrm{d}t} = g(x, y) = dxy - ey ,$$

with $\bar{x} = \frac{a}{b}$ when $y = 0$, and $\bar{x} = \frac{e}{d}$ and $\bar{y} = \frac{ad}{ce} - \frac{b}{c}$

$$J = \begin{pmatrix} \partial_x f \ \partial_y f \\ \partial_x g \ \partial_y g \end{pmatrix} = \begin{pmatrix} -b - c\bar{y} & -c\bar{x} \\ d\bar{y} & d\bar{x} - e \end{pmatrix}$$

$$J = \begin{pmatrix} \partial_x f \ \partial_y f \\ \partial_x g \ \partial_y g \end{pmatrix} = \begin{pmatrix} -b - c\bar{y} & -c\bar{x} \\ d\bar{y} & d\bar{x} - e \end{pmatrix}$$

Fill in $\bar{x} = \frac{a}{b}$ and $\bar{y} = 0$,
$$J_1 = \begin{pmatrix} -b & -\frac{ca}{b} \\ 0 & \frac{da}{b} - e \end{pmatrix}$$

Since this matrix is in a diagonal form we know that the diagonal elements provide the eigenvalues, i.e., $\lambda_1 = -b$ and $\lambda_2 = \frac{da}{b} - e$.

Hence this state is stable whenever $\lambda_2 < 0$, i.e., $\frac{a}{b} < \frac{e}{d}$.

$$J = \begin{pmatrix} \partial_x f \ \partial_y f \\ \partial_x g \ \partial_y g \end{pmatrix} = \begin{pmatrix} -b - c\bar{y} & -c\bar{x} \\ d\bar{y} & d\bar{x} - e \end{pmatrix}$$

Now consider $\bar{x} = \frac{e}{d}$ and $\bar{y} = \frac{ad}{ce} - \frac{b}{c}$ and first fill in \bar{x} :
$$J_2 = \begin{pmatrix} -b - c\bar{y} & -\frac{ce}{d} \\ d\bar{y} & 0 \end{pmatrix}$$

When $\bar{y} > 0$ the signs of this matrix are given by

$$J_3 = \begin{pmatrix} -\alpha & -\beta \\ \gamma & 0 \end{pmatrix}$$
 with $\operatorname{tr} J_3 = -\alpha < 0$ and $\det J_3 = \beta \gamma > 0$,

such that

$$\lambda_{1,2} = \frac{\operatorname{tr} \pm \sqrt{\operatorname{tr}^2 - 4 \operatorname{det}}}{2} = \frac{-\alpha \pm \sqrt{\alpha^2 - 4\beta\gamma}}{2} < 0 ,$$

Since $\lambda_{1,2} < 0$ the non-trivial steady state is stable.

Having

$$J = \begin{pmatrix} \partial_x f & \partial_y f \\ \partial_x g & \partial_y g \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

we know that

$$\lambda_{1,2} = \frac{\operatorname{tr} \pm \sqrt{D}}{2}$$
 where $D = \operatorname{tr}^2 - 4 \operatorname{det}$

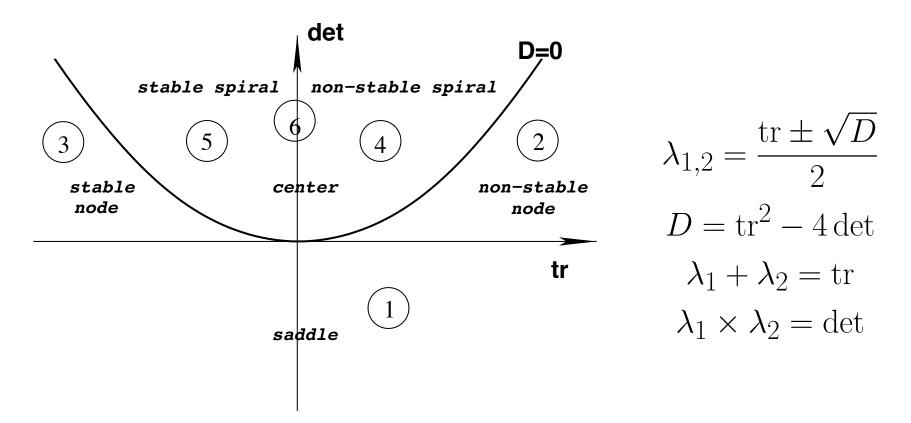
Observing that

$$\lambda_1 + \lambda_2 = \operatorname{tr}[J] \quad \text{and} \quad \lambda_1 \times \lambda_2 = \operatorname{det}[J] ,$$

the latter because

$$\frac{1}{4}(\mathrm{tr} + \sqrt{D})(\mathrm{tr} - \sqrt{D}) = \frac{1}{4}(\mathrm{tr}^2 - D) = \frac{1}{4}(\mathrm{tr}^2 - \mathrm{tr}^2 + 4 \det) = \det$$

we can classify steady states by just the trace and determinant of
their Jacobi matrix.



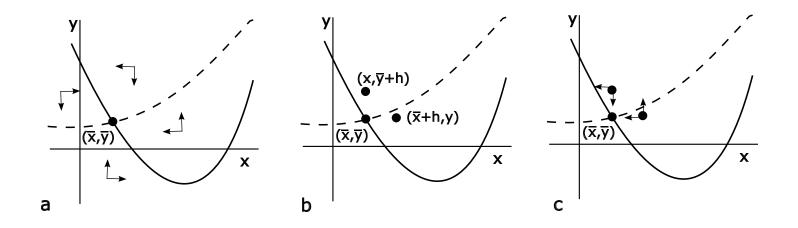
1. if det < 0 then D > 0: $\lambda_{1,2}$ are real with unequal sign: saddle 2. if det > 0, tr > 0 and D > 0 then $\lambda_{1,2} > 0$: unstable node. 3. if det > 0, tr < 0 and D > 0 then $\lambda_{1,2} < 0$: stable node. 4. if det > 0, tr > 0 and D < 0 then $\lambda_{1,2} > 0$: unstable spiral. 5. if det > 0, tr < 0 and D < 0 then $\lambda_{1,2} > 0$: stable spiral.

Graphical Jacobian: use the signs only

$$J = \begin{pmatrix} \partial_x f \simeq \frac{f(\bar{x} + h, \bar{y})}{h} & \partial_y f \simeq \frac{f(\bar{x}, \bar{y} + h)}{h} \\ \partial_x g \simeq \frac{g(\bar{x} + h, \bar{y})}{h} & \partial_y g \simeq \frac{g(\bar{x}, \bar{y} + h)}{h} \end{pmatrix} = \begin{pmatrix} \alpha & \beta \\ \gamma & \delta \end{pmatrix}$$

with
$$\operatorname{tr}[J] = \alpha + \delta$$
 and $\operatorname{det}[J] = \alpha \delta - \beta \gamma$.

If tr < 0 and det > 0 the state will be stable.



$$J = \begin{pmatrix} \partial_x f \simeq \frac{f(\bar{x} + h, \bar{y})}{h} & \partial_y f \simeq \frac{f(\bar{x}, \bar{y} + h)}{h} \\ \partial_x g \simeq \frac{g(\bar{x} + h, \bar{y})}{h} & \partial_y g \simeq \frac{g(\bar{x}, \bar{y} + h)}{h} \end{pmatrix} = \begin{pmatrix} - & - \\ + & - \end{pmatrix}$$