

Qualitative discussion PART 2 § 9.1 - 9.2 + Laplace Transform § 6.1 - 6.2

Remark: many ODE cannot be solved by analytic methods \Rightarrow it is important to consider qualitative info.

Goal: know the behavior of solutions without solving the ODE.

We have discussed: autonomous + scalar + 1st-order ODE: $y' = G(y)$



Now: 2x2 system! $\Rightarrow \vec{x}' = G(\vec{x})$

① classification

① First case: $G(\vec{x}) = A \cdot \vec{x}$, A constant matrix. Namely ODE: linear + homogeneous + const. coeff.'s

\top

(Remark: in this case, we do know the solutions)

② Eq. points: $G(\vec{x}) = \vec{0} \Rightarrow A \cdot \vec{x} = \vec{0} \Rightarrow \text{Eq. points} = \ker(A)$.

For simplicity we will consider only the case where $\ker(A) = \{\vec{0}\}$, namely A is invertible

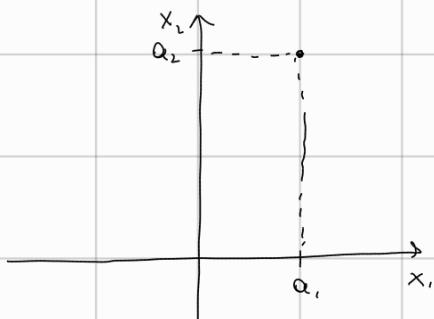
(i.e. $\det(A) \neq 0$), so we have just 1 eq. point $\vec{x}_0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$.

Remark: solutions of $\vec{x}' = A\vec{x}$ are of the form $\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$: so they can be represented in x_1, x_2 -plane

as follows:

for every time $\bar{t} \rightarrow$ you get 2-coordinates $\begin{pmatrix} x_1(\bar{t}) \\ x_2(\bar{t}) \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$

which is a point in the x_1, x_2 -plane \rightarrow

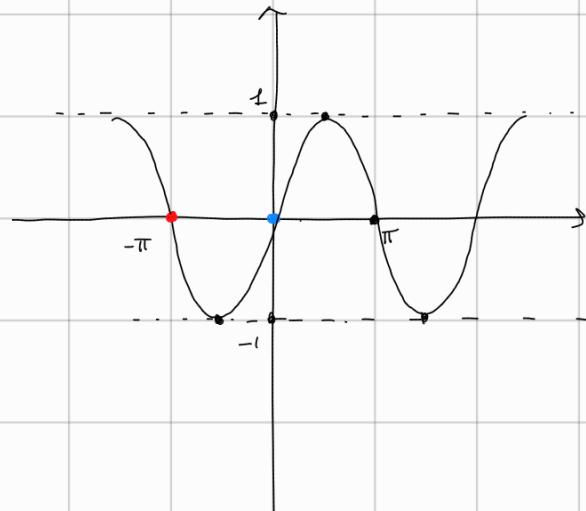


Example: ① $x(t) = \begin{pmatrix} t \\ \sin(t) \end{pmatrix}$

$$t=0: \begin{pmatrix} 0 \\ \sin(0) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \bullet$$

$$t=-\pi: \begin{pmatrix} -\pi \\ \sin(-\pi) \end{pmatrix} = \begin{pmatrix} -\pi \\ 0 \end{pmatrix} \Rightarrow \bullet$$

$$t = -\frac{\pi}{2}: \begin{pmatrix} -\frac{\pi}{2} \\ -1 \end{pmatrix} \Rightarrow$$

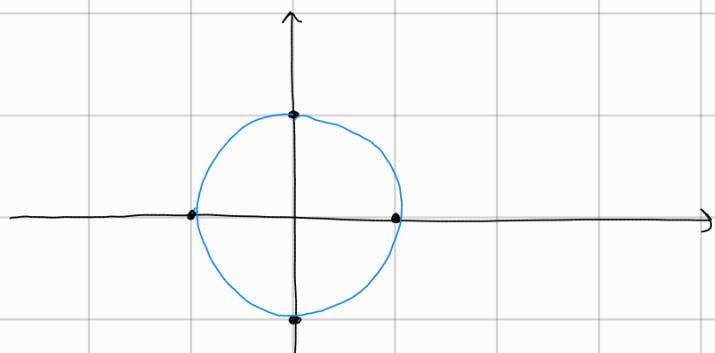


② $x(t) = \begin{pmatrix} \sin(\omega t) \\ \cos(\omega t) \end{pmatrix}$

$$t=0: \begin{pmatrix} \sin(0) \\ \cos(0) \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$t=\frac{\pi}{2}: \begin{pmatrix} \sin(\pi/2) \\ \cos(\pi/2) \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$t=\pi: \begin{pmatrix} 0 \\ -1 \end{pmatrix} \quad t = \frac{3\pi}{2} = \begin{pmatrix} -1 \\ 0 \end{pmatrix}$$



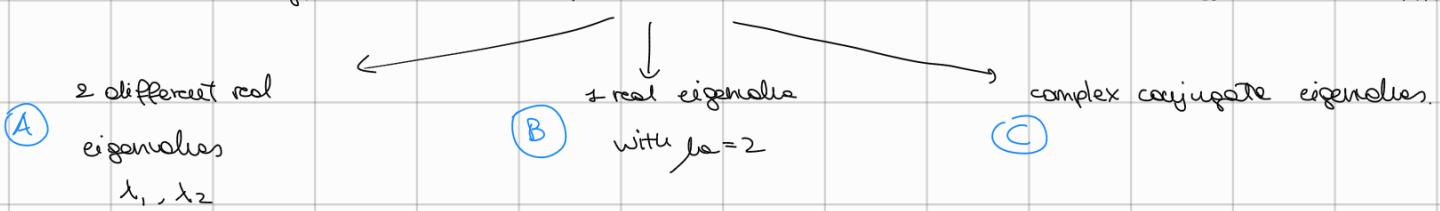
DEFINITION: $\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$ is a parametric representation (t is the parameter) of a curve in x_1x_2 -plane

This curve is called the trajectory of the solution.

The plane x_1x_2 is called the phase plane.

A "representative set" of trajectories is called phase portrait.

Let's see what a "representative set" is. Recall that we have 3 cases when we deal with $x' = Ax$



Rule, since $\det(A) \neq 0$, $\lambda \Rightarrow$ is never an eigenvalue!

(A) Since $\lambda_1 \neq \lambda_2$

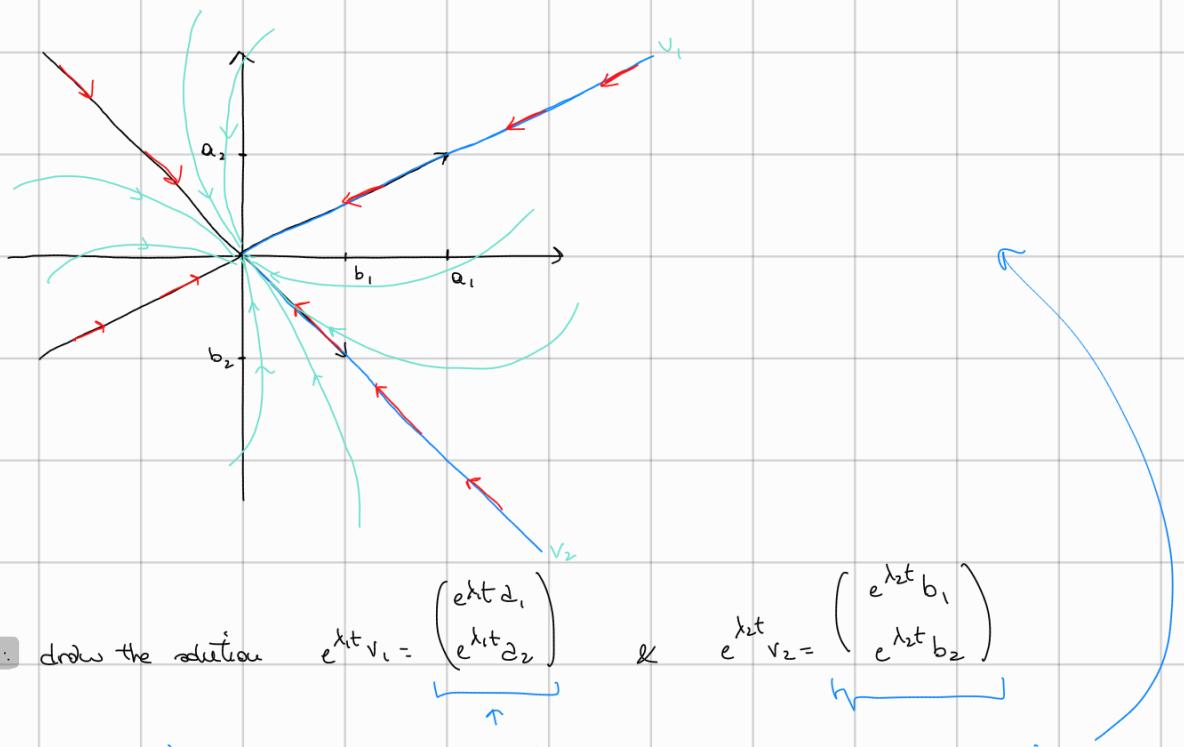
: the solution is $x(t) = C_1 e^{\lambda_1 t} v_1 + C_2 e^{\lambda_2 t} v_2$

(A1) $\boxed{\lambda_1 < \lambda_2 < 0}$
 $(\lambda_1 > |\lambda_2|)$

(A2) $\boxed{0 < \lambda_2 < \lambda_1}$
 $(\lambda_1 > \lambda_2)$

(A3) $\boxed{0 < \lambda_2 < \lambda_1}$
 $(\lambda_1 > \lambda_2)$

(A1): step 1: draw in the phase plane the eigenvectors $v_1 = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$ $v_2 = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$



step 3: draw $-e^{\lambda_1 t} v_1$, $-e^{\lambda_2 t} v_2$

step 4: put the "right arrows / directions": since both λ_1 & λ_2 are negative

then $t \rightarrow +\infty$ $e^{\lambda_1 t} \rightarrow 0$ & $e^{\lambda_2 t} \rightarrow 0$

\Rightarrow the solution $e^{\lambda_1 t} v_1 \rightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ & $e^{\lambda_2 t} v_2 \rightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

=> arrows pointing $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$

step 5: fill in: regardless the values of c_1 & $c_2 \rightarrow c_1 e^{\lambda_1 t} v_1 + c_2 e^{\lambda_2 t} v_2 \rightarrow (0)$ as $t \rightarrow +\infty$

\Rightarrow all the trajectories are approaching (0)

Moreover since $\lambda_1 - \lambda_2 < 0$:

$$e^{\lambda_2 t} [c_1 v_1 e^{(\lambda_1 - \lambda_2)t} + c_2 v_2]$$

↑

for $t \rightarrow +\infty$, this term can be forgotten w.r.t. $c_2 v_2$.

So not only the trajectories are approaching 0 , they approach $e^{\lambda_2 t} c_2 v_2$ as well!

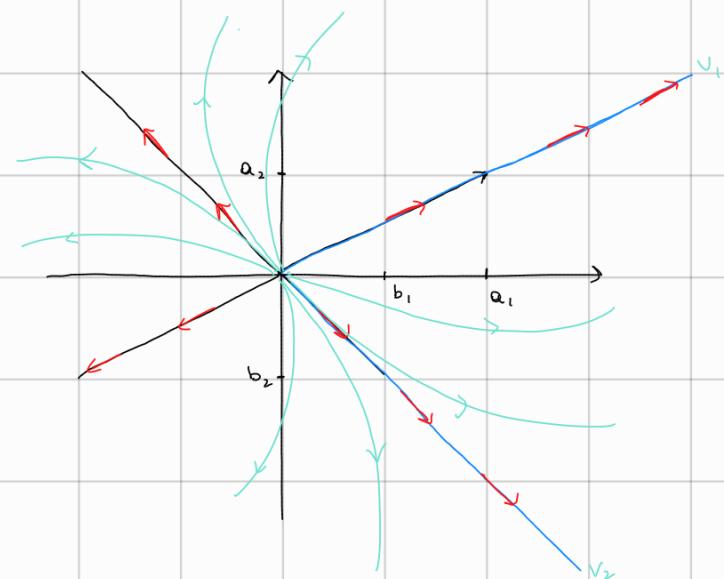
DEFINITION: in this case (0) is a NODE / NODAL SINK.

(A3) Same picture as before but

reversed arrows.

$$t \rightarrow +\infty \Rightarrow e^{\lambda_1 t} v_1, e^{\lambda_2 t} \rightarrow 0$$

Rule: $\boxed{\lambda_1 > \lambda_2}$:



NODE /

DEF: NODAL SOURCE \Rightarrow

A2

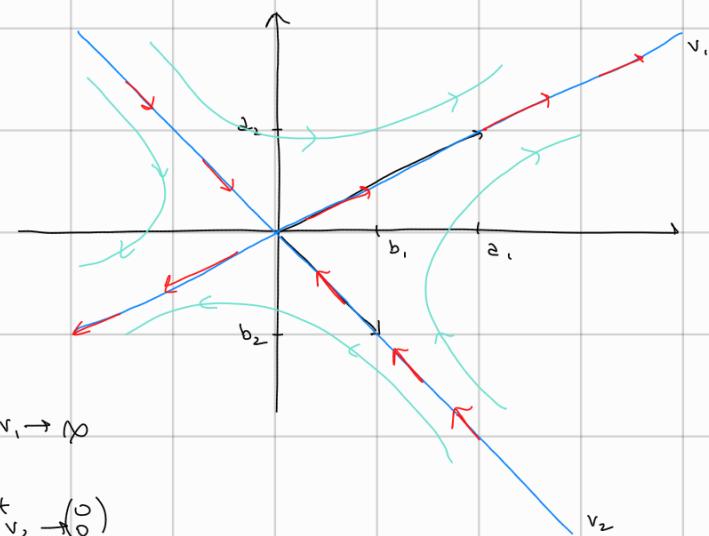
Same steps: ④ eigenvectors:

④ solutions $\pm e^{\lambda_1 t} v_1$

$\pm e^{\lambda_2 t} v_2$

④ signs $\lambda_1 > 0 : t \rightarrow +\infty \Rightarrow e^{\lambda_1 t} v_1 \rightarrow \infty$

$\lambda_2 < 0 : t \rightarrow +\infty \Rightarrow e^{\lambda_2 t} v_2 \rightarrow 0$



④ fill-in the gaps

DEF.. SADDLE POINT ↑



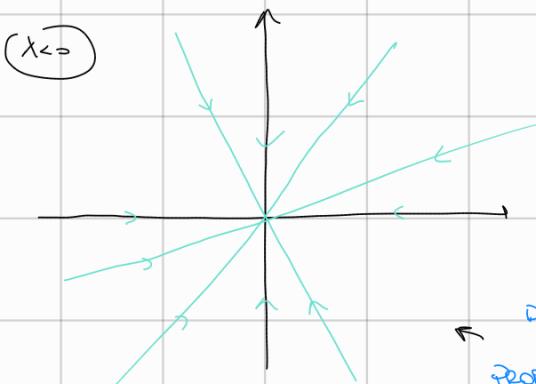
(in particular all solution $\rightarrow \infty$ but $c_2 e^{\lambda_2 t} v_2 \rightarrow 0$)

B1

$\lambda_1 = \lambda_2$, 2 lin. ind. eigenvectors. the general solution is

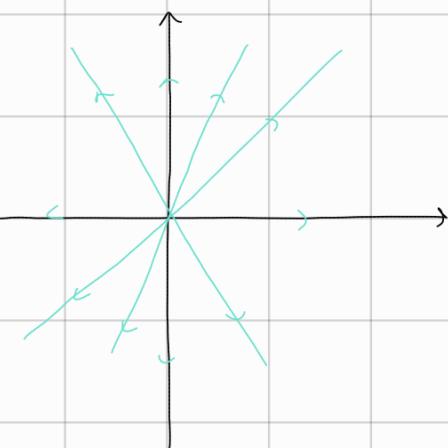
$$\underbrace{(c_1 v_1 + c_2 v_2)}_{v} e^{\lambda_1 t}$$

\Rightarrow fixed c_1, c_2 : trajectory is ALWAYS a straight line.



($\lambda > 0$)

DEF:
PROPER NODE
/ STAR POINT



B2

$\lambda > \mu$: 1 eigenvector $c_1 v e^{\lambda t} + c_2 (w + tv) e^{\lambda t}$

"

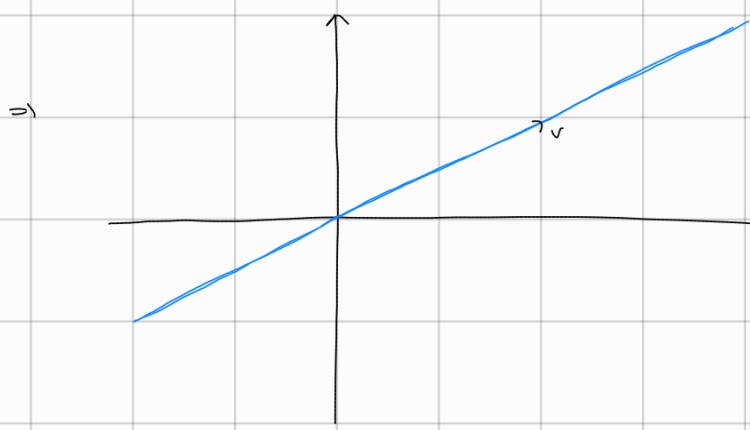
$$e^{\lambda t} [c_1 v + c_2 w + c_2 t v]$$

where $(A - \lambda \text{Id}) w = v$.

(*) $e^{\lambda t}$ determines the limit of the solution for $t \rightarrow +\infty$

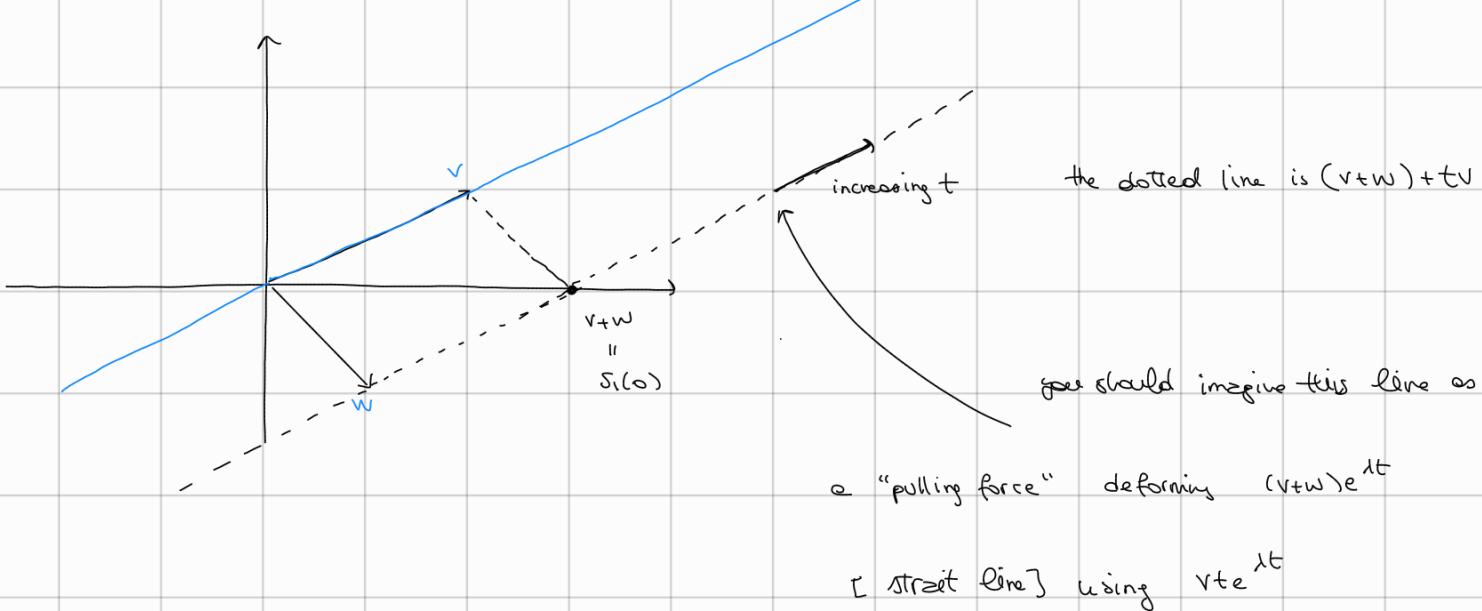
$\Rightarrow \lambda > 0 : \text{solution} \rightarrow \infty, \quad \lambda < 0 : \text{solution} \rightarrow 0.$

(*) in the term $c_1v + c_2w + c_3tv$: $c_2 = 0$: $c_1v \Rightarrow$ this corresponds to the solutions given by eigenvector $e^{\lambda t}$.



if $c_2 \neq 0$, for $t \rightarrow \pm\infty$ the dominant term $[c_1v + c_2w + c_3tv]$ is c_3tv .

Draw w on the phase plane. Now $c_1=c_2=1$ & let's draw $e^{\lambda t} [v+w+tv] = s_i(t)$



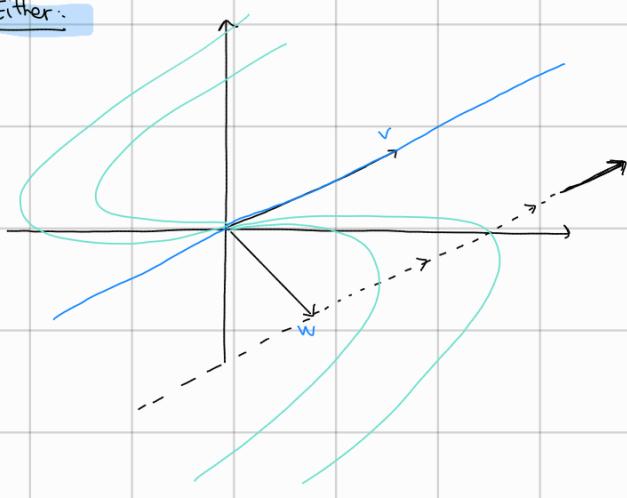
we expect:



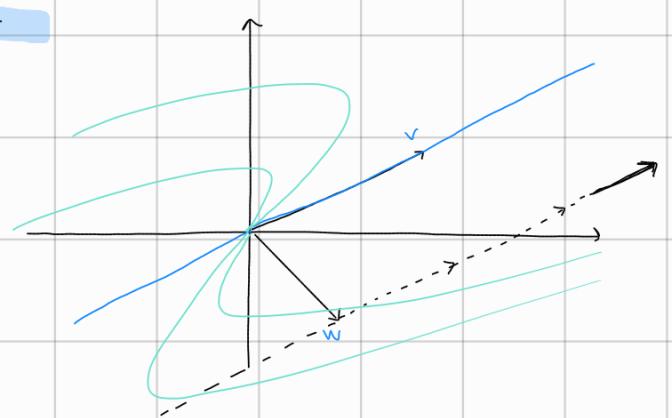
or

On the other side - the behavior is the same but in the opposite direction

Either:



Or:



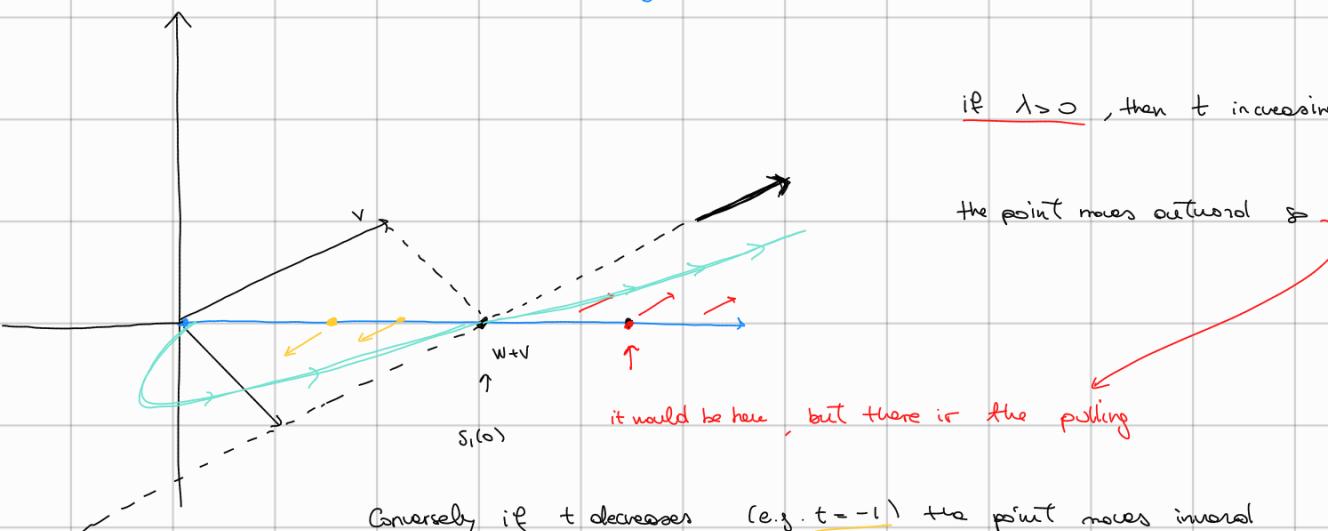
DEF: IMPROPER/DEGENERATE NODE

How to determine the right pulling?

- * Consider the line $(v+w) + tv$ and notice the direction along which t increases (it is given by v)

$$e^{\lambda t} [v+w+tv] = (v+w)e^{\lambda t} + v \cdot t \cdot e^{\lambda t}$$

blue line below



if $\lambda > 0$, then t increasing (e.g. $t=1$)

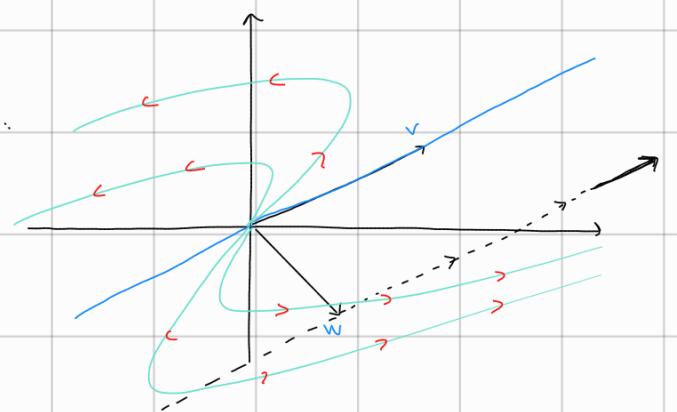
the point moves outward \Rightarrow

it would be here, but there is the pulling

Conversely if t decreases (e.g. $t=-1$) the point moves inward

t is decreasing = pulling in the opposite direction

\Rightarrow the right picture is then:

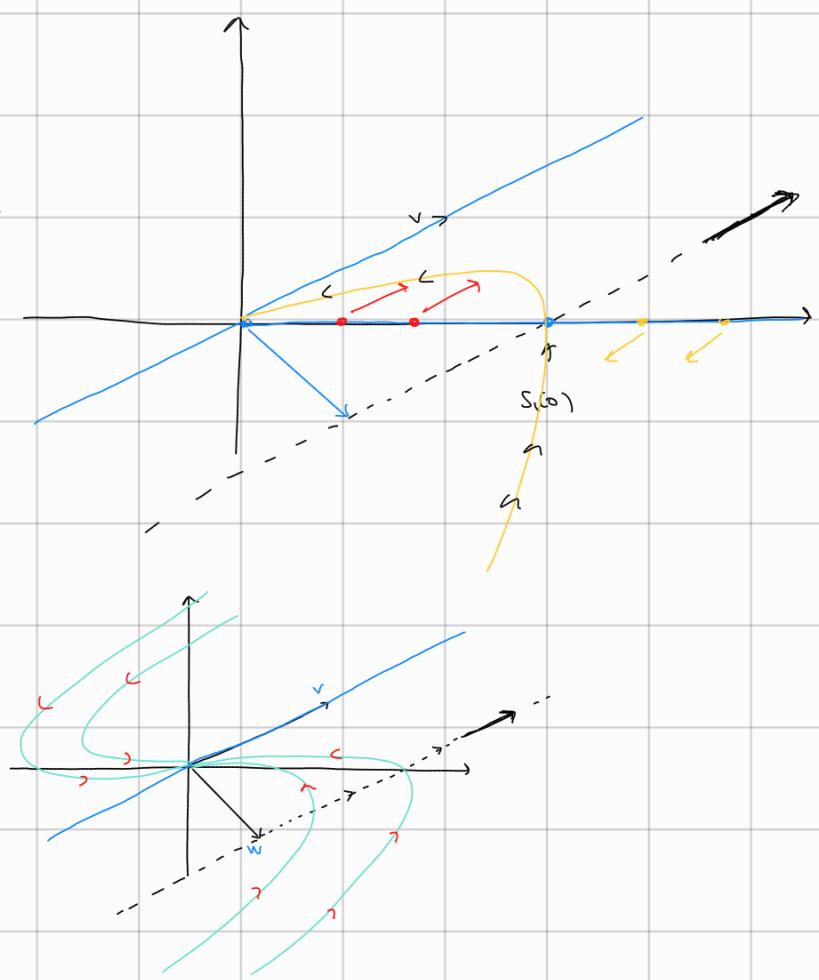


If $\lambda < 0$, the process is reversed:

t increasing \Rightarrow moving inward

t decreasing \Rightarrow moving outward

So, the right picture is:



(c)

$$\lambda = \alpha + i\beta, \bar{\lambda} = \alpha - i\beta. \text{ & pick } \beta > 0.$$

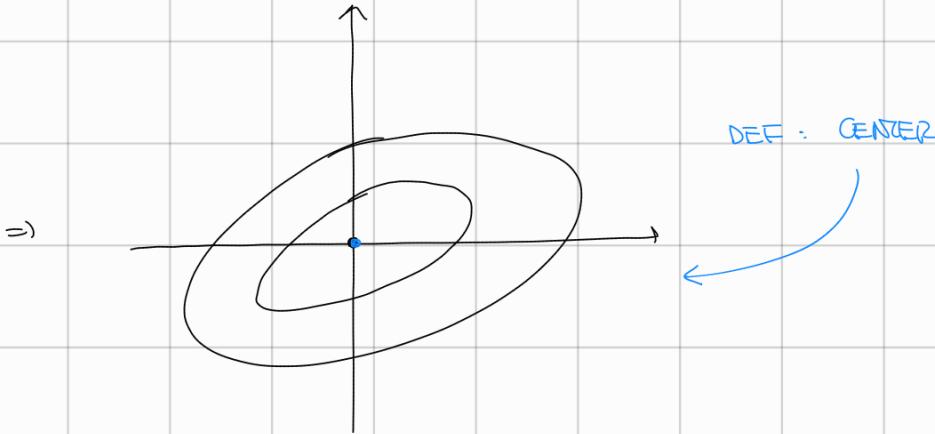
$$v = \vec{x} + i\vec{y}$$

(1) $d = 0$

: general solution $s_1 = \cos(\beta t) \cdot \vec{x} - \sin(\beta t) \vec{y}$ \leftarrow periodic of period $\frac{2\pi}{\beta}$

$$s_2 = \cos(\beta t) \vec{y} + \sin(\beta t) \vec{x}$$

\Rightarrow we got closed orbits! And they are around zero.



clockwise / counterclockwise:
you need to draw one solution.

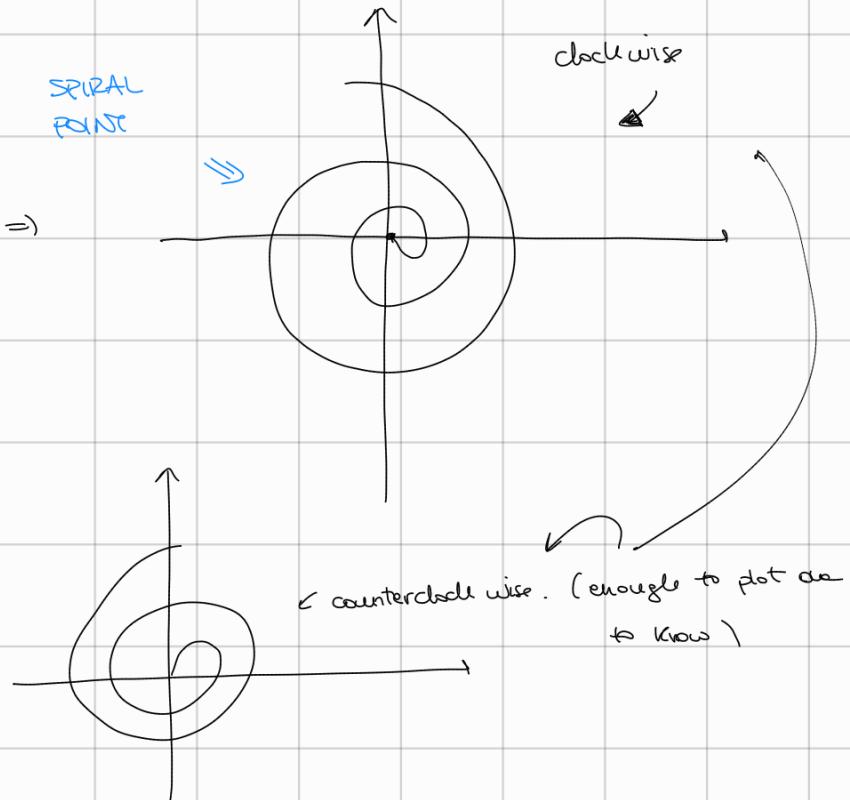
(C2) $\boxed{d \neq 0}$

$$s_1: e^{dt} [\cos(\beta t) \vec{x} - \sin(\beta t) \vec{y}]$$

$$s_2: e^{dt} [\cos(\beta t) \vec{y} + \sin(\beta t) \vec{x}]$$

no here a periodic component \rightarrow we are revolving around zero

However: no more closed orbits! Indeed there is the exponential component which makes the trajectories collapse to (0)



SPRAL
SINK

counter-clock wise. (enough to plot one to know)

• solutions escape \Rightarrow unstable

STABILITY CLASSIFICATION:

• solutions collapse \Rightarrow asymptotically stable

• solutions do not escape but do not collapse \Rightarrow stable

Node sink

asym. stable

Node source

unstable

Saddle point

unstable

Proper node $\lambda > 0$

unstable

Proper node $\lambda < 0$

asym. stable

Improper node $\lambda = 0$

unstable

Improper node $\lambda < 0$

asym. stable

Center

stable

Spiral sink

asym. stable

Spiral source

unstable

(I)

Examples: $x' = \begin{pmatrix} 3 & -2 \\ 2 & -2 \end{pmatrix} x$. $\det(A) = -6 + 4 = -2 \neq 0$.

Q1: Qualitative discussion + classify eq. points.

Find eigenvalues: $\det \begin{pmatrix} 3-\lambda & -2 \\ 2 & -2-\lambda \end{pmatrix} = (-2-\lambda)(3-\lambda) + 4 = -6 - 3\lambda + 2\lambda + \lambda^2 + 4 = \lambda^2 - \lambda - 2 =$

$$= (\lambda-2)(\lambda+1) = \lambda=2, -1$$

$\uparrow \quad \uparrow$

$\lambda_2 \quad \lambda_1$

We are in the A2-case: saddle point = unstable.

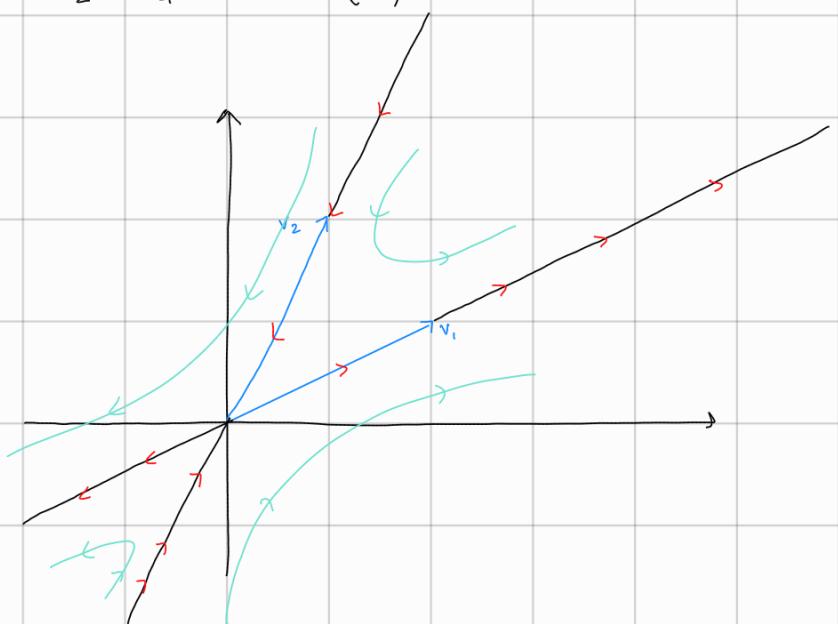
Sketch: we need to find the eigenvectors.

$$\lambda_1: (\lambda=2)$$

$$\begin{array}{cc|c} 1 & -2 & 0 \\ 2 & -4 & 0 \end{array} \Rightarrow \begin{pmatrix} 2 \\ 1 \end{pmatrix} = v_1$$

$$\lambda_2: (\lambda=-1)$$

$$\begin{array}{cc|c} 4 & -2 & 0 \\ 2 & -1 & 0 \end{array} \Rightarrow \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$



$$(II) x^1 = \begin{pmatrix} 2 & -5 \\ 1 & -2 \end{pmatrix} x$$

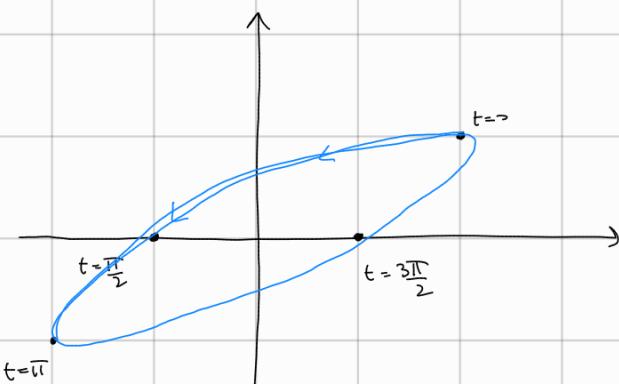
$$\det = -4 + 5 = 1$$

$$\text{Eigenvalues: } \begin{pmatrix} 2-\lambda & -5 \\ 1 & -2-\lambda \end{pmatrix} = 1 \quad \lambda^2 - 4 + 5 = \lambda^2 + 1 \quad \Rightarrow \lambda = i, -i \quad d \Rightarrow, \beta = 1.$$

\Rightarrow CENTER (stable)

$$\text{Eigenvectors: } \begin{pmatrix} 2-i & -5 \\ 1 & -2-i \end{pmatrix} : \quad x_1 = (2+i)x_2 \quad \begin{pmatrix} 2+i \\ 1 \end{pmatrix} = v \quad \Rightarrow \quad x = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, y = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$s_2 = \cos(t) \begin{pmatrix} 2 \\ 1 \end{pmatrix} - \sin(t) \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \cos(t) & -\sin(t) \\ \cos(t) & 0 \end{pmatrix}$$



$$\begin{aligned}
 & t=0 \quad \begin{pmatrix} 2 \\ 1 \end{pmatrix} \\
 & t=\frac{\pi}{2} \quad \begin{pmatrix} -\sin(\pi/2) \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \\
 & t=\pi \quad \begin{pmatrix} -2 \\ -1 \end{pmatrix} \\
 & t=\frac{3\pi}{2} \quad \begin{pmatrix} 2\cos(\frac{3\pi}{2}) - \sin(\frac{3\pi}{2}) \\ \cos(\frac{3\pi}{2}) \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}
 \end{aligned}$$

General case: no more linear: $x' = G(x) \Rightarrow \begin{pmatrix} x_1' \\ x_2' \end{pmatrix} = \begin{pmatrix} G_1(x_1, x_2) \\ G_2(x_1, x_2) \end{pmatrix}$

Example: $\begin{pmatrix} x_1' \\ x_2' \end{pmatrix} = \begin{pmatrix} x_1^2 + x_2^2 \\ 2x_1 x_2 \end{pmatrix}$

DEF: as before the set of points $x \dots G(x) = \dots$ are called equilibrium / critical points

Equivalently: $G_1(x_1, x_2) = G_2(x_1, x_2) = \dots \Rightarrow$ they generate equilibrium solutions.

Stability: same definition as before

- stay close .. stable
- escape : unstable
- collapse : asym. stable .

Aim: Use our discussion for linear ODEs for studying non-linear ones in proximity of eq. points.

Key step: G_1, G_2 continuously differentiable \Rightarrow the non-linear system can be approximated

using Taylor expansion of G

Recall: the first term of a Taylor expansion for a function $G: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ at $\begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$

$$G\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = G\begin{pmatrix} a_1 \\ a_2 \end{pmatrix} + J_G\begin{pmatrix} a_1 \\ a_2 \end{pmatrix} \begin{pmatrix} x_1 - a_1 \\ x_2 - a_2 \end{pmatrix} + \dots$$

↑
Jacobian of G : $\begin{pmatrix} \frac{\partial G_1}{\partial x_1} \Big|_{(a_1)} & \frac{\partial G_1}{\partial x_2} \Big|_{(a_1)} \\ \frac{\partial G_2}{\partial x_1} \Big|_{(a_2)} & \frac{\partial G_2}{\partial x_2} \Big|_{(a_2)} \end{pmatrix}$

Rule: if $\begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$ is picked to be one of the eq. points \Rightarrow by definition $G\begin{pmatrix} a_1 \\ a_2 \end{pmatrix} = \vec{0}$.

\Rightarrow in order to study $\begin{pmatrix} x_1' \\ x_2' \end{pmatrix} = G\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \Rightarrow$ study $\begin{pmatrix} (x_1 - a_1)' \\ (x_2 - a_2)' \end{pmatrix} = J_G\begin{pmatrix} a_1 \\ a_2 \end{pmatrix} \begin{pmatrix} x_1 - a_1 \\ x_2 - a_2 \end{pmatrix}$

Assumption: $\det(J_G(a_2)) \neq 0$! This is the first case.

$$x_1' = -(x_1 - x_2)(1 - x_2 - x_1) = G_1$$

Example: $x_2' = x_1(2 + x_2) = G_2$.

$$G_1 \Leftrightarrow x_1 = x_2 \text{ or } x_1 + x_2 = 1$$

1st) Find eq. points: $G_1 = 0 = G_2 \Rightarrow$

$$G_2 \Leftrightarrow x_1 = 0 \text{ or } x_2 = -2$$

$$\Rightarrow \begin{cases} x_1 = x_2 \\ x_1 = 0 \end{cases} \quad \text{or} \quad \begin{cases} x_1 = x_2 \\ x_2 = -2 \end{cases} \quad \text{or} \quad \begin{cases} x_1 + x_2 = 1 \\ x_1 = 0 \end{cases} \quad \text{or} \quad \begin{cases} x_1 + x_2 = 1 \\ x_2 = -2 \end{cases}$$

↓ ↓ ↓ ↓

$$P_1 = (0, 0) \quad P_2 = (-2, -2) \quad P_3 = (0, 1) \quad P_4 = (3, -2)$$

2nd) Find linear approximation: \circledast Under continuity assumption, G_1 & G_2 are polynomials \Rightarrow

they are cont. differentiable.

(2) Find Jacobian: $\frac{\partial G_1}{\partial x_1} = \frac{\partial}{\partial x_1} -(x_1 - x_2)(1 - x_2 - x_1) = -(1 - x_2 - x_1) + (x_1 - x_2) = -1 + \cancel{x_2} + x_1 + \cancel{x_1} - \cancel{x_2}$

\downarrow

$$= 2x_1 - 1$$

$$\frac{\partial G_1}{\partial x_2} = (1 - x_2 - x_1) + (x_1 - x_2) = 1 - x_2 - \cancel{x_1} + \cancel{x_1} - x_2 = 1 - 2x_2$$

$$\frac{\partial G_2}{\partial x_1} = \frac{\partial}{\partial x_1} x_1(2 + x_2) = 2 + x_2 ; \quad \frac{\partial G_2}{\partial x_2} = x_1$$

$$J_G(x_1, x_2) = \begin{pmatrix} 2x_1 - 1 & 1 - 2x_2 \\ x_2 + 2 & x_1 \end{pmatrix}.$$

$$J_G(p_1) = \begin{pmatrix} -1 & 1 \\ 2 & 0 \end{pmatrix}; \quad J_G(p_2) = \begin{pmatrix} -5 & 5 \\ 0 & -2 \end{pmatrix}; \quad J_G(p_3) = \begin{pmatrix} -1 & -1 \\ 3 & 0 \end{pmatrix}; \quad J_G(p_4) = \begin{pmatrix} 5 & 5 \\ 0 & 3 \end{pmatrix}$$

$$(1) \begin{pmatrix} -1-\lambda & 1 \\ 2 & -\lambda \end{pmatrix} \Rightarrow -\lambda(-1-\lambda) - 2 = 0 \Rightarrow \lambda + \lambda^2 - 2 = 0 \Rightarrow (\lambda+2)(\lambda-1) = 0;$$

$-2 < \lambda < 1 \Rightarrow$ saddle point.

$$(2) \begin{pmatrix} -5-\lambda & 5 \\ 0 & -2-\lambda \end{pmatrix} \Rightarrow (-5-\lambda)(-2-\lambda) = 0 \Rightarrow \lambda = -5 \text{ or } -2; \quad -5 < -2 < 0 \Rightarrow \text{ nodal sink}$$

$$(3) \begin{pmatrix} -1-\lambda & -1 \\ 3 & -\lambda \end{pmatrix} \Rightarrow \lambda(\lambda+1) + 3 = 0 \Rightarrow \lambda^2 + \lambda + 3 = 0 \Rightarrow \lambda_{1,2} = \frac{-1 \pm \sqrt{1-12}}{2} = \frac{-1 \pm \sqrt{11}}{2}; \quad \alpha = -\frac{1}{2}, \quad \beta = \frac{\sqrt{11}}{2} \Rightarrow \text{ spiral sink}.$$

$$(4) \begin{pmatrix} 5-\lambda & 5 \\ 0 & 3-\lambda \end{pmatrix} \Rightarrow \lambda = 3, 5 \Rightarrow \text{ nodal source}$$

Clarification for stability for non-linear

if the linearized ODE has:

we say the non linear

or eq. point is:

Nodal sink

asym. stable

Nodal source

unstable

Saddle point

unstable

Proper node $\lambda > 0$

unstable

Proper node $\lambda < 0$

asym. stable

Same as before apart from

Improper node $\lambda = 0$

unstable

CENTER \rightarrow indeterminate.

Improper node $\lambda < 0$

asym. stable

We say the same if $\det(\mathcal{J}_G(p)) = 0$

Spiral sink

asym. stable

Spiral source

unstable

Laplace Transform: Recall that

$$\int_0^{+\infty} f(t) dt = \lim_{T \rightarrow +\infty} \int_0^T f(t) dt \quad (\text{if this limit exists})$$

improper integral

Notation: if the limit \exists & it is finite we say that the improper integral converges

$\dots \dots \exists$ or $\dots \infty \dots \dots \dots$ diverges.

Example, $f(t) = e^{ct}$.

$c \in \mathbb{R}$

$$\int_0^T e^{ct} dt = \begin{cases} \frac{e^{ct}}{c} \Big|_0^T & \text{if } c \neq 0 \\ t \Big|_0^T & \text{if } c = 0 \end{cases} \Rightarrow \begin{cases} \frac{e^{cT} - 1}{c} & \text{if } c \neq 0 \\ T & \text{if } c = 0 \end{cases}$$

$$\lim_{T \rightarrow +\infty} T = +\infty \quad X$$

$$\lim_{T \rightarrow +\infty} \frac{e^{cT} - 1}{c} = \begin{cases} +\infty & c > 0 \\ -\frac{1}{c} & c < 0 \end{cases} \quad X$$

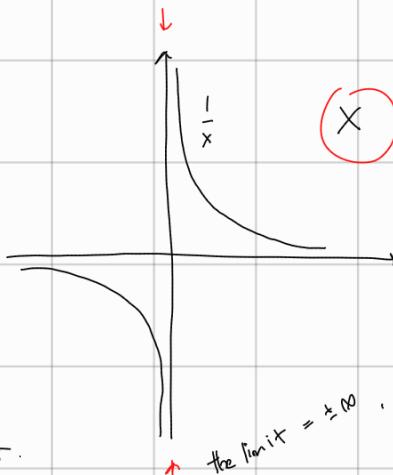
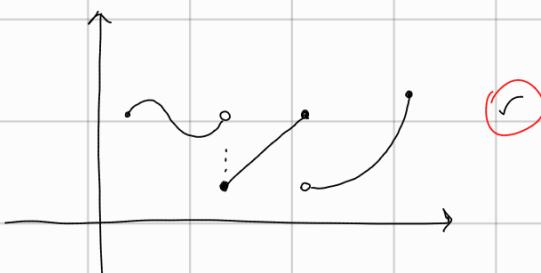
$\Rightarrow \int_0^{+\infty} e^{ct} dt \text{ converges iff. } c < 0$

DEF: $f(t)$ is piecewise continuous on $[t_0, b]$ if \exists $\exists t_0 < \dots < t_n < b$ s.t

1. $f(t)$ continuous on $[t_0, t_1], (t_1, t_2), (t_2, t_3), \dots, (t_{n-1}, b]$

2. $\lim_{t \rightarrow t_i^+} f(t), \lim_{t \rightarrow t_i^-} f(t)$ both exist & finite

Example:

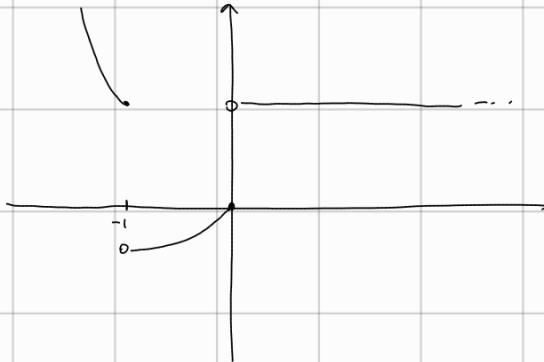


Remark: any continuous function is piecewise continuous.

In particular $f(t)$ piecewise continuous can be written as

$$f(t) = \begin{cases} f_1(t) & t \in [t_0, t_1] \\ \vdots & \vdots \\ f_n(t) & t \in (t_{n-1}, b] \end{cases}$$

Example: $f(x) = \begin{cases} x^2 & x \leq -1 \\ \sin(x) & -1 < x \leq 0 \\ 1 & x > 0 \end{cases}$



DEF: Take $f(t)$ piecewise continuous : define the laplace transform of f $\mathcal{L}(f)$ to be

$$\mathcal{L}(f)(s) = \int_0^\infty e^{-st} f(t) dt \quad s = \text{new variable - it does NOT depend on } t.$$

Punk: $\mathcal{L}(f)$ may not be well defined for all s .

Example: $f(t) = e^{at} \Rightarrow L(f)(s) = \int_0^{\infty} e^{-st} \cdot e^{at} dt = \int_0^{\infty} e^{(a-s)t} dt = \frac{1}{s-a}$ for $s > a$

(it is not well defined otherwise)

for what we have seen before

However, we have nice properties:

① Laplace transform is linear, $L(af + bg) = aL(f) + bL(g)$

② Relationship with derivatives:

a) Under some technical conditions: f continuous + f' piecewise continuous + $\exists k, \varepsilon, M$ s.t.

$$|f(t)| \leq K e^{at} \quad \forall t \geq M$$

$$\Rightarrow L(f'(t)) \exists \quad \forall s > a \quad \text{and} \quad L(f'(t)) = s \cdot L(f) - f(0)$$

b) Assume f & f' continuous + f'' piecewise continuous + $\exists k, \varepsilon, M$: $|f(t)|, |f'(t)| \leq K e^{at} \quad \forall t \geq M$

$$\Rightarrow L(f''(t)) \exists \quad \forall s > a - k$$

$$L(f''(t)) = s^2 L(f) - s f(0) - f'(0)$$

c) For general n -th derivative: $L(f^{(n)}(t)) = s^n L(f) - s^{n-1} f(0) - \dots - s f^{(n-2)}(0) - f^{(n-1)}(0)$.

[again under technical assumptions]

How do we want to use it?

Pick our IVP . ex. $y'' - y' - 2y = 0$, $y(0) = 1$, $y'(0) = 0$.

Apply L on both sides assuming that you can. $\Rightarrow L(y'' - y' - 2y) = L(0) = 0$.

$$\mathcal{L}(y'') - \mathcal{L}(y') - 2\mathcal{L}(y) = \dots$$

$$\underline{\mathcal{L}(y'')} - \underline{\mathcal{L}(y')} - \underline{\mathcal{L}(y)} = \underline{-2\mathcal{L}(y)} = 0.$$

↑ ↑ ↑ ↑ ↑
 1 0 1 1

$$\Rightarrow \mathcal{L}(y)(s^2 - s - 2) = s - 1 \Rightarrow \mathcal{L}(y) = \frac{s-1}{s^2 - s - 2} = \frac{s-1}{(s-2)(s+1)} = \frac{\frac{1}{3}}{\underbrace{s-2}_{\uparrow}} + \frac{\frac{2}{3}}{s+1}$$

$$\mathcal{L}(e^{2t}) \quad \mathcal{L}(e^{-t})$$

Pro: find $\mathcal{L}(y)$ of a solution is easy

Cons: find \mathcal{L}^{-1} - Laplace inverse of a function can be tricky.

$$y = \frac{e^{2t}}{3} + \frac{2}{3}e^{-t}$$

[Thm: if \exists , it is unique]

However! There is a table you can use - 317 of the book.