

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: $\Rightarrow \vec{x}' = G(\vec{x})$

classification

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



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

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

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

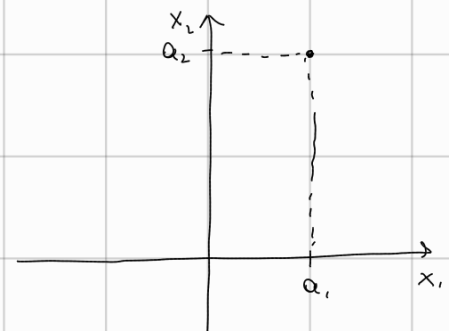
(i.e. $\det(A) \neq 0$), so we have just 1 eq. point $\vec{x}_0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$.

Remark: solutions of $x' = Ax$ 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 any 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

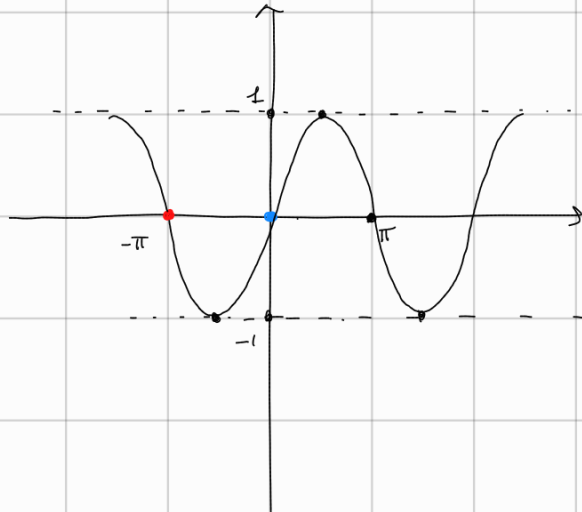


Examples: ① $x(t) = \begin{pmatrix} t \\ \sin(t) \end{pmatrix}$

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

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

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

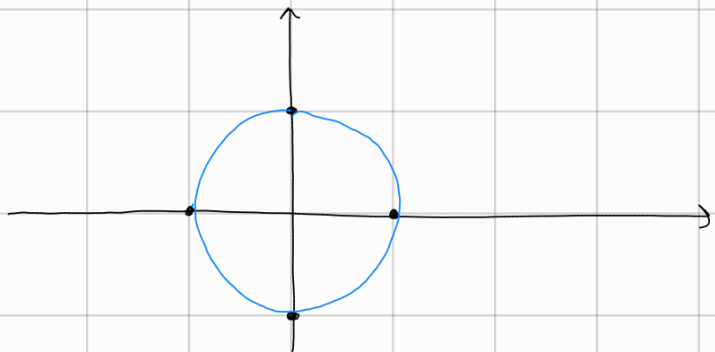


② $x(t) = \begin{pmatrix} \sin(t) \\ \cos(t) \end{pmatrix}$

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

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

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



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

This curve is called the trajectory of the solution.

The plane x_1, x_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 $x' = Ax$

① 2 different real eigenvalues λ_1, λ_2

② ± real eigenvalues with $\mu = 2$

③ complex conjugate eigenvalues.

Remark: since $\det(A) \neq 0$, $\lambda = 0$ is never an eigenvalue!

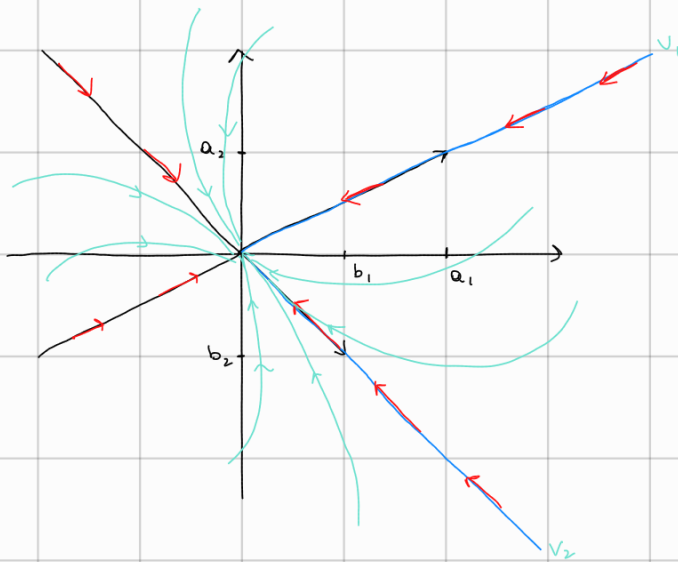
(A) Since $\lambda_1 \neq \lambda_2$: the solution is $x(t) = C_1 e^{\lambda_1 t} v_1 + C_2 e^{\lambda_2 t} v_2$

(A1) $\lambda_1 < \lambda_2 < 0$
(so $|\lambda_1| > |\lambda_2|$)

(A2) $\lambda_2 < 0 < \lambda_1$
($\lambda_1 > \lambda_2$)

(A3) $0 < \lambda_2 < \lambda_1$
($\lambda_1 > \lambda_2$)

(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 2: draw the solution

$$e^{\lambda_1 t} v_1 = \begin{pmatrix} e^{\lambda_1 t} a_1 \\ e^{\lambda_1 t} a_2 \end{pmatrix}$$

notice: always a positive multiple of v_1

$$e^{\lambda_2 t} v_2 = \begin{pmatrix} e^{\lambda_2 t} b_1 \\ e^{\lambda_2 t} b_2 \end{pmatrix}$$

always a positive multiple of v_2

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

\Rightarrow arrows pointing $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$

steps: fill in: regarden the values of c_1 & $c_2 \rightarrow c_1 e^{\lambda_1 t} v_1 + c_2 e^{\lambda_2 t} v_2 \rightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ as $t \rightarrow +\infty$

\Rightarrow all the trajectories are approaching $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$

Moreover since $\lambda_1 - \lambda_2 < 0$: $e^{\lambda_2 t} [\underbrace{c_1 v_1 e^{(\lambda_1 - \lambda_2)t}}_{\uparrow} + c_2 v_2]$
 for $t \rightarrow +\infty$, this term can be forgotten w.r.t. $c_2 v_2$.

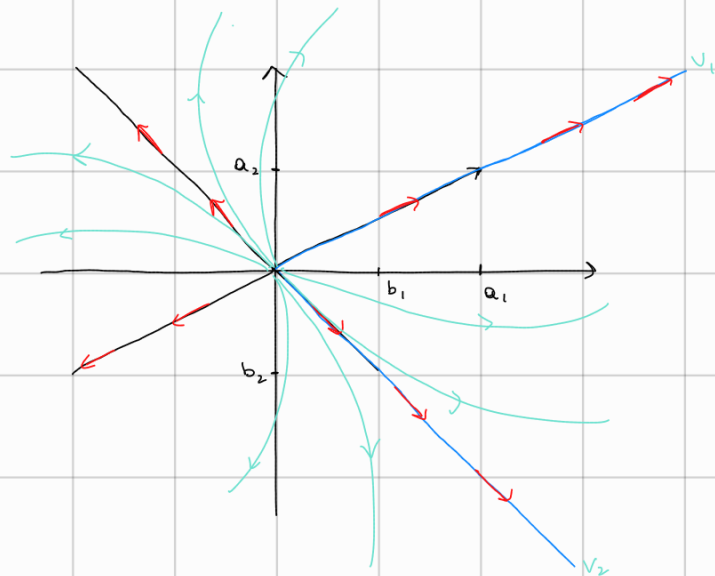
So not only the trajectories are approaching 0, they approach $e^{\lambda_2 t} c_2 v_2$ as well!

DEFINITION: in this case $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$ is a NODE / NODAL SINK.

(A3) Same picture as before but

reversed arrows:

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



Rule: $\lambda_1 > \lambda_2$!

DEF: NODE / NODAL SOURCE \Rightarrow

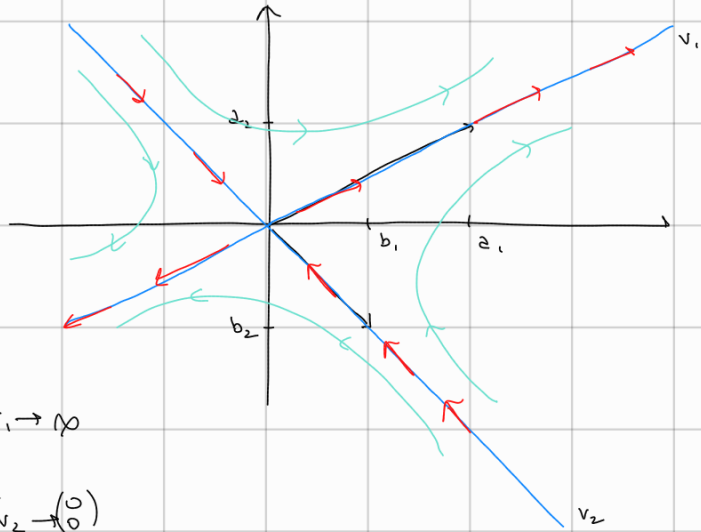
A2

Same steps: (a) eigenvectors:

(b) solutions $\pm e^{\lambda_1 t} v_1$
 $\pm e^{\lambda_2 t} v_2$

(c) Signs $\lambda_1 > 0 : t \rightarrow +\infty \Rightarrow e^{\lambda_1 t} v_1 \rightarrow \infty$

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



* fill-in the gaps

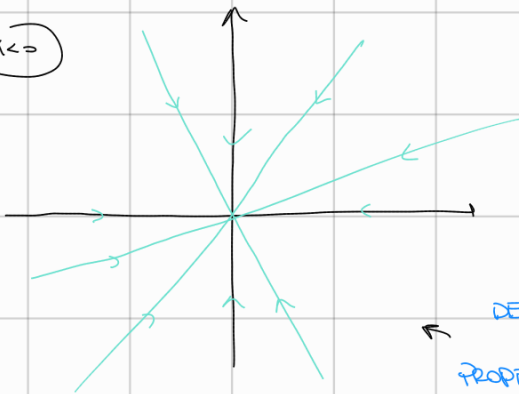
DEF: SADDLE POINT ↑↑

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

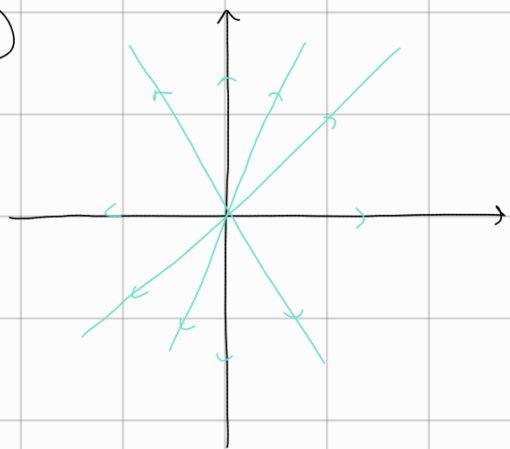
(B) $\lambda_1 = \lambda_2$: 2 lin. ind. eigenvectors. the general solution is $\underbrace{(C_1 v_1 + C_2 v_2)}_v e^{\lambda t}$

\Rightarrow fixed C_1, C_2 : trajectory is ALWAYS a straight line.

$\lambda < 0$



$\lambda > 0$



DEF: PROPER NODE / STAR POINT

(B2) $\lambda_1 = \lambda_2$: 1 eigenvector $C_1 v e^{\lambda t} + C_2 (w + tv) e^{\lambda t}$ where $(A - \lambda Id)w = v$.
 $e^{\lambda t} [C_1 v + C_2 w + C_2 tv]$

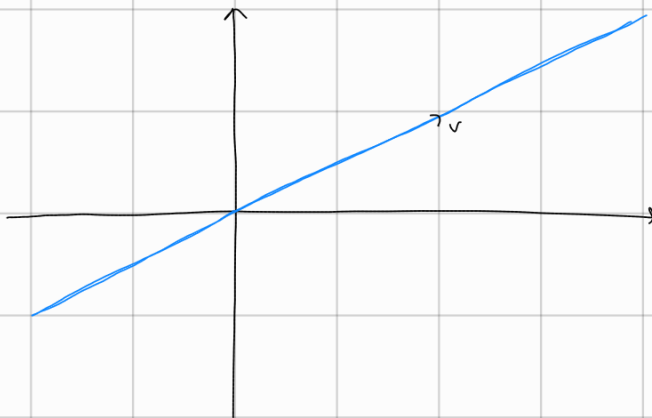
* $e^{\lambda t}$ determines the limit of the solution for $t \rightarrow +\infty$

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

* in the term $c_1 v + c_2 w + c_3 t v$: if $c_2 = 0$: $c_1 v \Rightarrow$ this corresponds to the solutions given

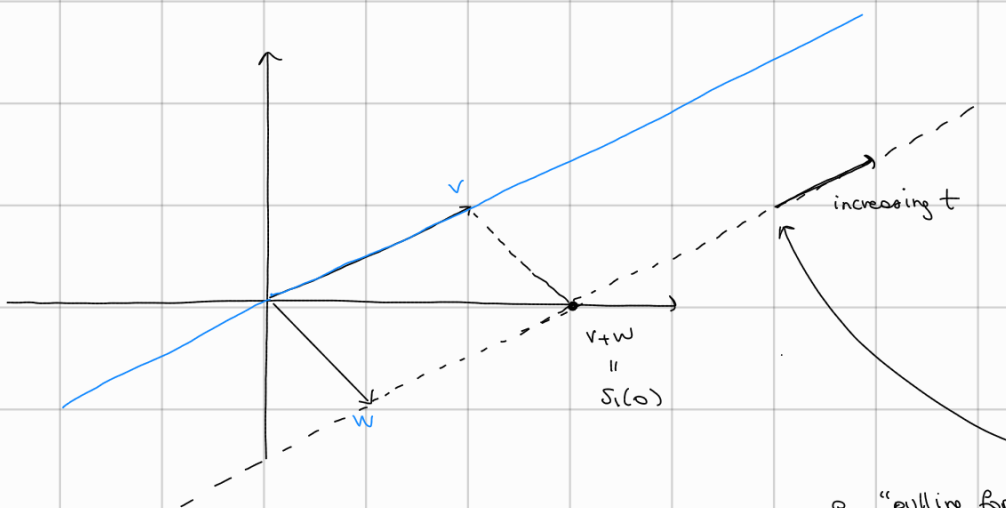
by eigenvector $\cdot e^{\lambda t}$

\Rightarrow



if $c_2 \neq 0$: for $t \rightarrow \pm\infty$ the dominant term $[c_1 v + c_2 w + c_3 t v]$ is $c_3 t v$.

Draw w on the phase plane. Pose $c_1 = c_2 = 1$ & let's draw $e^{\lambda t} [v + w + t v] = s_1(t)$



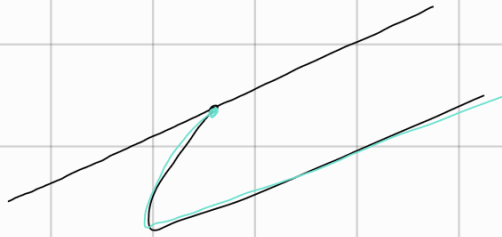
the dotted line is $(v+w) + tv$

you should imagine this line as

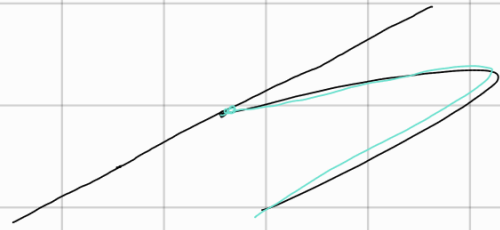
a "pulling force" deforming $(v+w)e^{\lambda t}$

[straight line] using $v t e^{\lambda t}$

we expect:
 \Rightarrow
 either

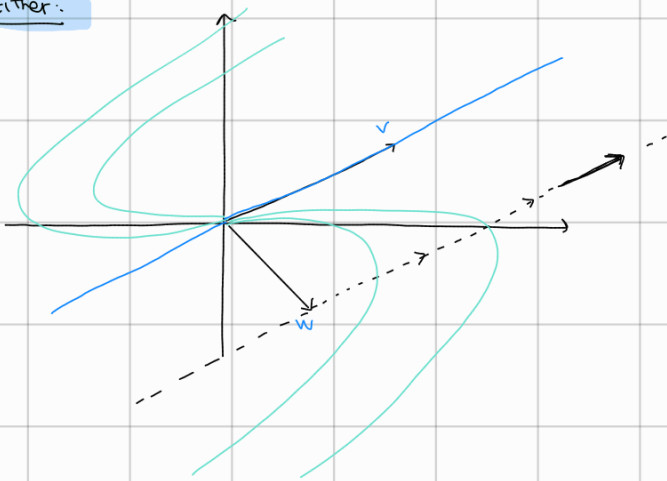


or

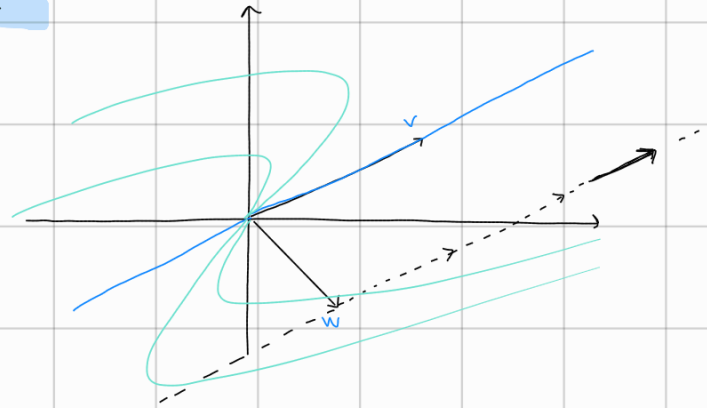


On the other side: the behavior is the same but in the opposite direction

Either:



Or:

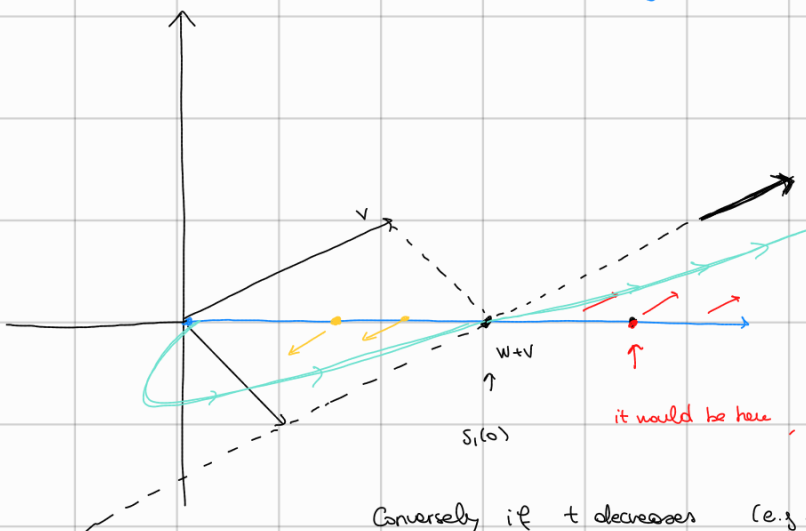


DEF: IMPROPER / DEGENERATE NODE

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)

$$e^{\lambda t} [v+w + tv] = \underbrace{(v+w)}_{\substack{\text{blue line} \\ \text{below} \\ \downarrow}} e^{\lambda t} + v \cdot t \cdot e^{\lambda t}$$



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

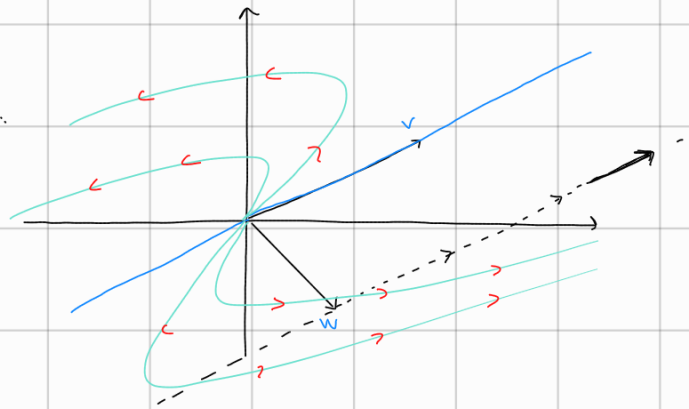
the point moves outward \rightarrow

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)

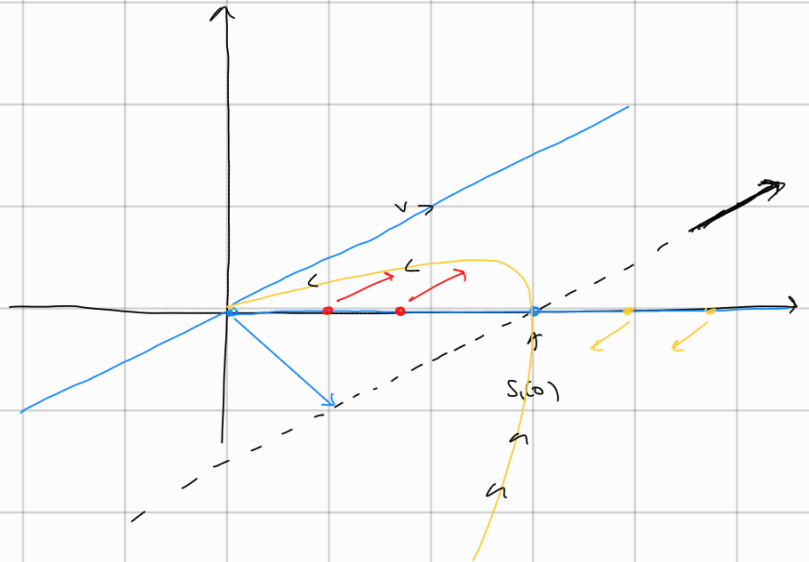
=> the right picture is then:



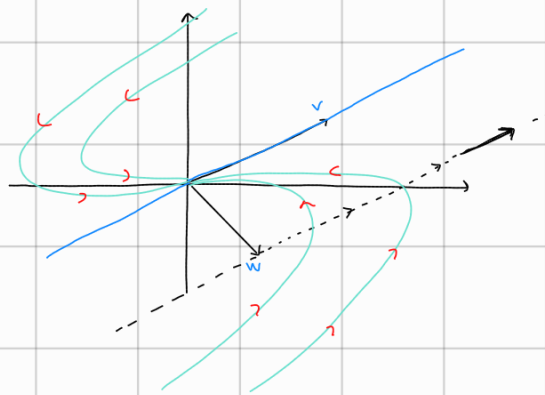
If $\lambda < 0$, the process is reversed:

t increasing = moving inward

t decreasing = moving outward



So, the right picture is:

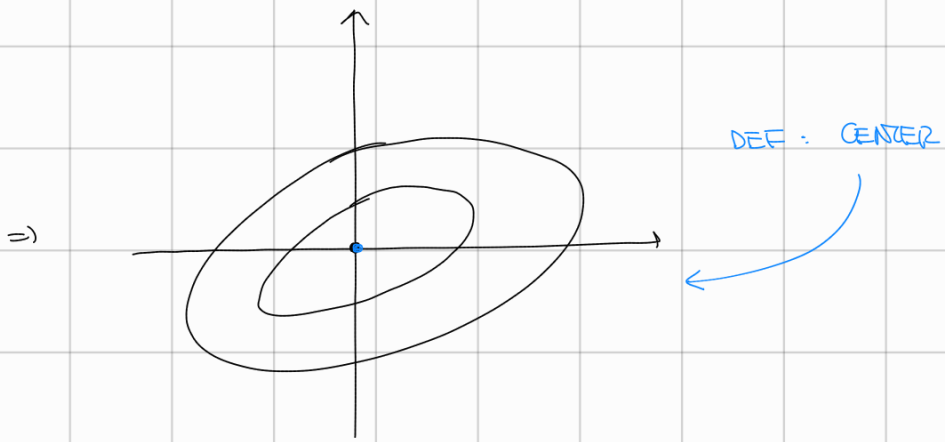


(C) $\lambda = \alpha + i\beta$, $\bar{\lambda} = \alpha - i\beta$. α pick $\beta > 0$.

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

(C1) $\alpha = 0$: general solution $s_1 = \cos(\beta t) \cdot \vec{x} - \sin(\beta t) \vec{y}$ ← periodic of period $\frac{2\pi}{\beta}$
 $s_2 = \cos(\beta t) \vec{y} + \sin(\beta t) \vec{x}$ ←

=> we get closed orbits! And they are around zero.



clockwise / counterclockwise:
you need to draw a solution.

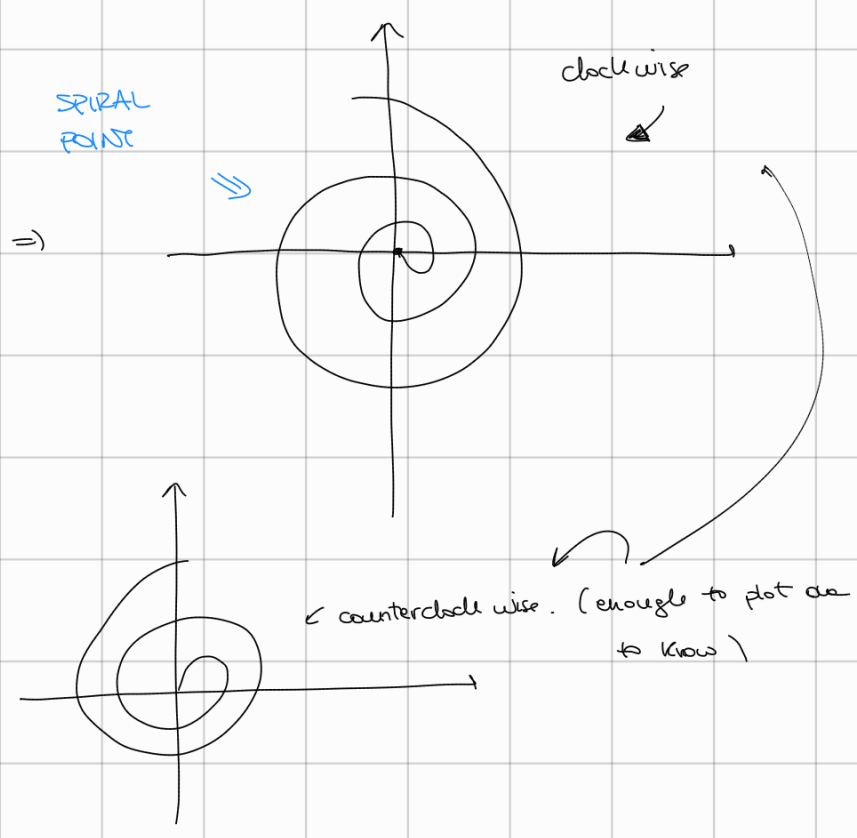
(C2) $\alpha \neq 0$

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

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

no real 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 $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$



SPIRAL SINK \Downarrow

direction:

outward if $\alpha > 0$

inward if $\alpha < 0$

SPIRAL SINK \Downarrow

• solutions escape \Rightarrow unstable

• solutions collapse \Rightarrow asymptotically stable

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

STABILITY CLASSIFICATION:

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

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

Q1: Qualitative discussion + classify eq. points.

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

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

$-1 < 0 < 2$
$\uparrow \quad \uparrow$
$\lambda_2 \quad \lambda_1$

We are in the λ_2 -case: saddle point = unstable

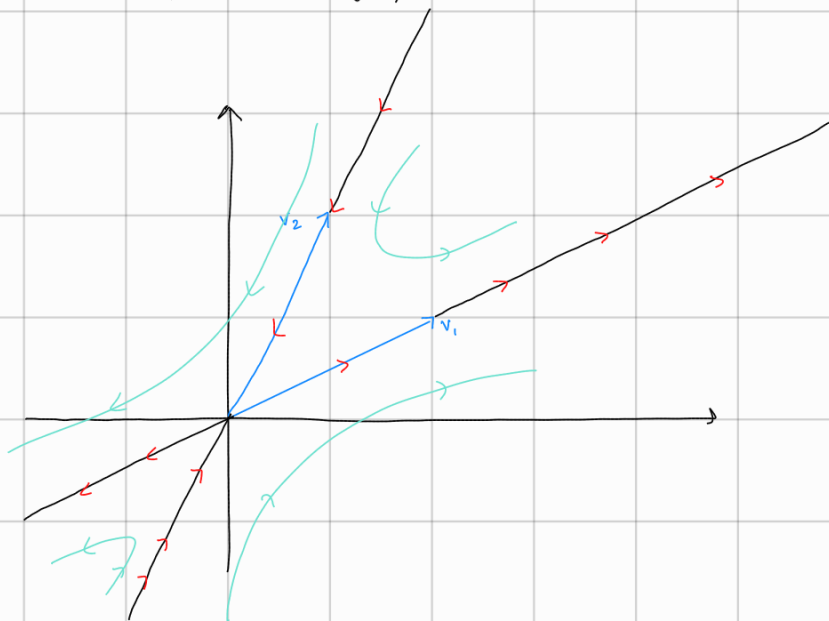
Sketch: we need to find the eigenvectors.

$\lambda_1: \lambda = 2$

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

$\lambda_2: \lambda = -1$

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



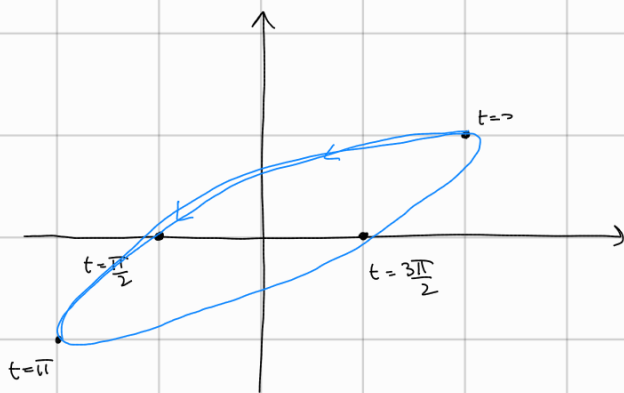
(III) $x' = \begin{pmatrix} 2 & -5 \\ 1 & -2 \end{pmatrix} x$ $\det = -4 + 5 = 1$

Eigenvalues: $\begin{pmatrix} 2-\lambda & -5 \\ 1 & -2-\lambda \end{pmatrix} \Rightarrow \lambda^2 - 4 + 5 = \lambda^2 + 1 = 0 \Rightarrow \lambda = i, -i$ $\alpha = 0, \beta = 1$

\Rightarrow CENTER (stable)

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

$$s_1 = \cos(t) \begin{pmatrix} 2 \\ 1 \end{pmatrix} - \sin(t) \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2\cos(t) - \sin(t) \\ \cos(t) \end{pmatrix}$$



$$t=0 \quad \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

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

$$t=\pi \quad \begin{pmatrix} -2 \\ -1 \end{pmatrix}$$

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

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

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

DEF: as before the set of points x . $G(x) = 0$ are called **equilibrium / critical points**

Equivalently: $G_1(x_1, x_2) = G_2(x_1, x_2) = 0 \Rightarrow$ 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 $\begin{pmatrix} a \\ a_2 \end{pmatrix}$

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

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

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

\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: $\det(J_G \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}) \neq 0!$ ↑ This is the first case.

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

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

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

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

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

$\Rightarrow \begin{cases} x_1 = x_2 \\ x_1 = 0 \end{cases}$	or	$\begin{cases} x_1 = x_2 \\ x_2 = -2 \end{cases}$	or	$\begin{cases} x_1 + x_2 = 1 \\ x_1 = 0 \end{cases}$	or	$\begin{cases} x_1 + x_2 = 1 \\ x_2 = -2 \end{cases}$
\Downarrow		\Downarrow		\Downarrow		\Downarrow
$p_1 = (0, 0)$		$p_2 = (-2, -2)$		$p_3 = (0, 1)$		$p_4 = (3, -2)$

2nd) Find linear approximation: \otimes Check continuity assumption. G_1 & G_2 are polynomials \Rightarrow

they are cont. differentiable.

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

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

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

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

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

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

$-2 < 0 < 1 \Rightarrow$ saddle point

$$(2) \begin{pmatrix} -5-\lambda & 5 \\ 0 & -2-\lambda \end{pmatrix} \Rightarrow (-5-\lambda)(-2-\lambda) = \lambda^2 + 7\lambda + 10 = (\lambda+5)(\lambda+2) = 0$$

$-5 < -2 < 0 \Rightarrow$ nodal sink

$$(3) \begin{pmatrix} -1-\lambda & -1 \\ 3 & -\lambda \end{pmatrix} \Rightarrow \lambda(\lambda+1) + 3 = \lambda^2 + \lambda + 3 = 0 \Rightarrow \lambda_{1,2} = \frac{-1 \pm \sqrt{1-12}}{2} = \frac{-1 \pm \sqrt{-11}}{2}$$

$\alpha = -\frac{1}{2}, \beta = \frac{\sqrt{11}}{2} \Rightarrow$ spiral sink

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

Classification for stability for non-linear

if the linearized ODE has:

we say the non linear

as eq. point is:

Nodal sink

asym. stable

Nodal source

unstable

Saddle point

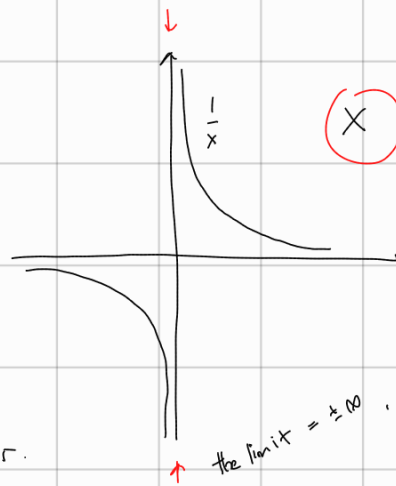
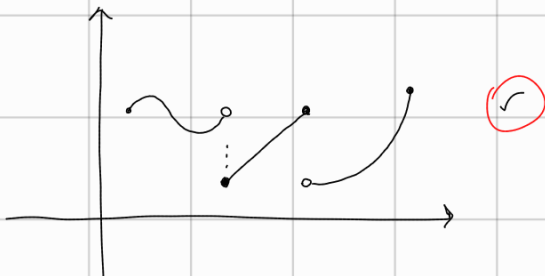
unstable

DEF: $f(x)$ is **piecewise continuous** on $[a, b]$ if $\exists a < t_1 < \dots < t_{n-1} < b$ s.t.

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

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

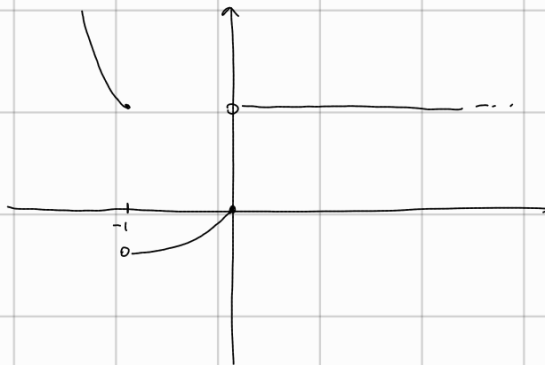
Example:



Remark: any continuous function is piecewise continuous.

In particular $f(x)$ piecewise continuous can be written as $f(x) = \begin{cases} f_1(x) & x \in [a, t_1) \\ \vdots & \vdots \\ f_n(x) & x \in (t_{n-1}, b] \end{cases}$

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



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

$$\mathcal{L}(f)(s) = \int_0^{\infty} e^{-st} f(t) dt$$

$s =$ new variable - it does NOT depend on t .

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

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

(it is not well defined otherwise)

for what we have seen before

however, we have nice properties:

① Laplace transform is linear, $\mathcal{L}(af + bg) = a\mathcal{L}(f) + b\mathcal{L}(g)$

② Relationship with derivatives:

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

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

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

b) [Assume f & f' continuous + f'' piecewise continuous + $\exists k, \epsilon, M: |f(t)|, |f'(t)| \leq k e^{\epsilon t} \forall t \geq M$]

$\Rightarrow \mathcal{L}(f''(t)) \exists \forall s > a$ &

$$\mathcal{L}(f''(t)) = s^2 \mathcal{L}(f) - s f(0) - f'(0)$$

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

[again under technical assumptions]

How do we want to use it?

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

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

$$\mathcal{L}(y'') - \mathcal{L}(y') - 2\mathcal{L}(y) = 0 \quad \Rightarrow \quad \frac{s^2 \mathcal{L}(y) - s y(0) - y'(0)}{1} - \frac{-s \mathcal{L}(y) - y(0)}{1} - 2 \mathcal{L}(y) = 0.$$

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

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

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

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

[thm: if \exists , it is unique]

$$\Downarrow$$

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

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