

## Lesson 5 - 6/10/2022

II Linearize  $\ddot{x} = -\sin x - \dot{x}$  at equilibria  $(\bar{x}, 0)$  with  $\bar{x} \in [0, 2\pi]$ .

First order  $\rightarrow \begin{cases} \dot{x} = v \\ \dot{v} = -\sin x - v \end{cases} \Rightarrow J(x, v) = \begin{pmatrix} 0 & 1 \\ -\cos x & -1 \end{pmatrix}$

Equilibria?  $\Leftrightarrow x(\bar{x}, \bar{v}) = (0, 0) \Leftrightarrow \begin{cases} \bar{v} = 0 \\ \sin \bar{x} = 0 \end{cases}$

$\Leftrightarrow \bar{x} = 0 \text{ and } \bar{x} = \pi.$

$\underbrace{J(0, 0)}_{\substack{\text{First} \\ \text{eq.}}} = \begin{pmatrix} 0 & 1 \\ -1 & -1 \end{pmatrix} \rightarrow \begin{cases} \dot{x} = v \\ \dot{v} = -x - v \end{cases} \rightarrow \boxed{\ddot{x} = -x - \dot{x}}$

$\underbrace{J(\pi, 0)}_{\substack{\text{Second} \\ \text{eq.}}} = \begin{pmatrix} 0 & 1 \\ 1 & -1 \end{pmatrix} \rightarrow \begin{cases} \dot{x} = v \\ \dot{v} = x - \pi - v \end{cases} \rightarrow \boxed{\ddot{x} = x - \pi - \dot{x}}$

The previous ex. is an example of linearization of a 2nd order diff. eq.

$\boxed{\ddot{x} = Y(x, \dot{x})} \Leftrightarrow \dot{v} = Y(x, v)$   
 At first order  $\begin{cases} \dot{x} = v \\ \dot{v} = Y(x, v) \end{cases}$

In this case equilibria are always of type  $(\bar{x}, 0)$  such that  $\underbrace{Y(\bar{x}, 0)}_{} = 0$   $\bar{x}$  is an eq. configuration

$\Rightarrow J(\bar{x}, 0) = \begin{pmatrix} 0 & 1 \\ \partial_x Y(\bar{x}, 0) & \partial_y Y(\bar{x}, 0) \end{pmatrix}$

$$\Rightarrow \left\{ \begin{array}{l} \dot{x} = v \\ \dot{v} = \partial_x Y(\bar{x}, 0)(x - \bar{x}) + \partial_y Y(\bar{x}, 0)v \end{array} \right. \quad \text{linearized around } (\bar{x}, 0) \text{ (eq.)}$$

2nd step: Study in details linear systems.

$$\left\{ \begin{array}{l} \dot{z} = Az \\ z(0) = z_0 \end{array} \right. \quad \begin{array}{l} z \in \mathbb{R}^n \\ A \text{ nxm matrix} \end{array}$$

We recall that the solution is given by the **matrix exponential**

[ Consider the 1-dim case:  $\left\{ \begin{array}{l} \dot{x} = ax \\ x(0) = x_0 \end{array} \right. \quad \begin{array}{l} a \in \mathbb{R} \\ x \in \mathbb{R} \end{array}$  at ]

$$\Rightarrow \varphi^t(x_0) = x_0 e^a t.$$

$$\text{Recall that } e^a = 1 + a + \frac{a^2}{2} + \frac{a^3}{3!} + \dots \quad a \in \mathbb{R}.$$

Analogously in dim m.

Def  $A = m \times m$  matrix.

$$e^A \stackrel{:=}{=} 1 + A + \frac{A^2}{2} + \frac{A^3}{3!} + \dots = \sum_{k=0}^{+\infty} \frac{A^k}{k!} \quad (*)$$

$\exp A$

We need to check that this def. is well-posed! In order to prove this fact, we recall that the set of  $m \times m$  matrices (in  $\mathbb{R}$ ) is a Banach space with the norm:

$$\|A\| := \sup_{\substack{x \in \mathbb{R}^m \\ x \neq 0 \\ (0, \dots, 0)}} \frac{|Ax|}{|x|} = \sup_{|x|=1} |Ax|$$

*n-times*

Moreover, it is a Banach algebra:  $\|AB\| \leq \|A\|\|B\|$

Lemma  $e^A$  is well defined that is the series (\*)  
is convergent.

Proof  $\|e^A\| = \left\| \sum_{k=0}^{+\infty} \frac{A^k}{k!} \right\| \leq \sum_{k=0}^{+\infty} \frac{\|A^k\|}{k!} \leq$

$$\leq \sum_{k=0}^{+\infty} \frac{\|A\|^k}{k!} = e^{\|A\|} < +\infty.$$

□

Banach algebra

Prop  $\begin{cases} \dot{z} = Az \\ z(0) = z_0 \end{cases} \rightsquigarrow \varphi^t(z_0) = e^{tA} z_0$

Proof Directly

Clearly,  $\varphi^0(z_0) = z_0$

$$\begin{aligned} \frac{d}{dt}(\varphi^t(z_0)) &= \frac{d}{dt}(e^{tA} z_0) = \\ &= \frac{d}{dt} \left( \sum_{k=0}^{+\infty} \frac{1}{k!} (tA)^k \right) z_0 = \\ &= \frac{d}{dt} \left( \underbrace{1}_{+0} + tA + \frac{(tA)^2}{2!} + \dots \right) z_0 = \\ &= \frac{d}{dt} \left( \sum_{k=1}^{+\infty} \frac{1}{k!} (tA)^k \right) z_0 = \\ &= \left[ \sum_{k=1}^{+\infty} \frac{1}{k!} t^{k-1} \underbrace{A^k}_{(k-1)!} \right] z_0 = \left[ \sum_{k=0}^{+\infty} \frac{1}{k!} t^k A^k \right] A z_0 = \\ &\quad \text{A } A^{k-1} \rightarrow = A \varphi^t(z_0) \\ \frac{d}{dt}(\varphi^t(z_0)) &= A \varphi^t(z_0) \Leftrightarrow \end{aligned}$$

this means that

$\varphi^t(z_0)$  is the unique solution of the linear v.f.  
(with starting cond.  $z_0$ ).  $\square$

Properties / exercises on the matrix exp.

1  $A, P$   $n \times n$  matrices.  $P$  invertible. Then

$$\exp(P^{-1}AP) = P^{-1} \exp(A) P$$

Proof  $\exp(P^{-1}AP) = \sum_{k=0}^{+\infty} \frac{1}{k!} \underbrace{(P^{-1}AP)^k}_{k \text{ times}} = P^{-1} \exp(A) P.$

$$(P^{-1}AP) \underbrace{(P^{-1}AP) \cdot \dots \cdot (P^{-1}AP)}_{K \text{ times}} = P^{-1} A^K P \quad \square$$

2  $A = \begin{pmatrix} 0 & \beta \\ -\beta & 0 \end{pmatrix} = \beta \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} = B$

$$\Rightarrow e^{tA} = \begin{pmatrix} \cos(\beta t) & \sin(\beta t) \\ -\sin(\beta t) & \cos(\beta t) \end{pmatrix}$$

$\downarrow$  clockwise rotation  
of angle  $\beta t$

$$(1, 0) \rightarrow \begin{array}{c} \nearrow \\ \searrow \end{array}$$



$$B = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \quad B^2 = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$B^3 = -B \quad B^4 = -B^2$$

$$B^5 = B$$

$$B \quad B^2 \quad -B \quad -B^2 \quad B \quad B^2 \dots$$

Consequently

$$e^{tA} = \sum_{k=0}^{+\infty} \frac{1}{k!} (tA)^k = \sum_{k=0}^{+\infty} \frac{1}{k!} t^k \beta^k B^k = \dots =$$

$$= \begin{pmatrix} \cos(\beta t) & \sin(\beta t) \\ -\sin(\beta t) & \cos(\beta t) \end{pmatrix}$$

Maclaurin series of

$\sin(\beta t)$  and  $\cos(\beta t)$

[3]  $A = \begin{pmatrix} 0 & 1 \\ -\omega^2 & 0 \end{pmatrix} \Rightarrow e^{tA} = \begin{pmatrix} \cos(\omega t) & \sin(\omega t)/\omega \\ -\omega \sin(\omega t) & \cos(\omega t) \end{pmatrix}$

$A$  comes from  $\ddot{x} = -\omega^2 x \rightarrow \begin{cases} \dot{x} = v \\ \dot{v} = -\omega^2 x \end{cases} \Rightarrow$

$$\Rightarrow \begin{pmatrix} \dot{x} \\ \dot{v} \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ -\omega^2 & 0 \end{pmatrix}}_A \begin{pmatrix} x \\ v \end{pmatrix}$$

We conjugate by  $P = \begin{pmatrix} 1 & 0 \\ 0 & \omega \end{pmatrix}$

$$P^{-1}AP = \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 1/\omega \end{pmatrix}}_A \underbrace{\begin{pmatrix} 0 & 1 \\ -\omega^2 & 0 \end{pmatrix}}_A \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & \omega \end{pmatrix}}_A = \begin{pmatrix} 0 & \omega \\ -\omega & 0 \end{pmatrix}$$

↓  
like in  
Ex. 2

From ex. 1, we know that

$$\exp(tP^{-1}AP) = P^{-1}\exp(tA)P \Rightarrow$$

$$\exp(tA) = P \underbrace{\exp(tP^{-1}AP)}_{\Rightarrow Ex. 2} P^{-1}$$

Therefore

$$\exp(tA) = \begin{pmatrix} 1 & 0 \\ 0 & \omega \end{pmatrix} \begin{pmatrix} \cos(\omega t) & \sin(\omega t) \\ -\sin(\omega t) & \cos(\omega t) \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1/\omega \end{pmatrix} =$$

$$= \begin{pmatrix} \cos(\omega t) & \sin(\omega t)/\omega \\ -\omega \sin(\omega t) & \cos(\omega t) \end{pmatrix}$$

### Example

- Solve the linear system  $\dot{x} = Ax$ , where  $A = \begin{pmatrix} \alpha & 0 \\ 0 & -1 \end{pmatrix}$
- Graph the phase-portrait as  $\alpha \in \mathbb{R}$ , showing qualitative differences.

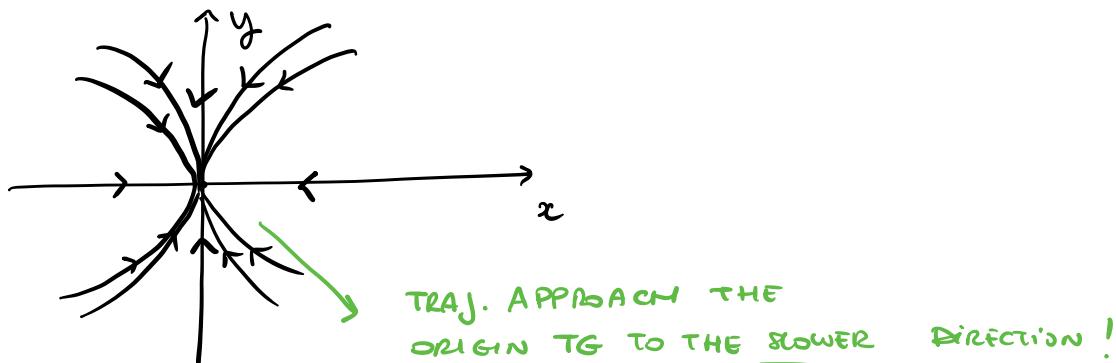
Solution

$$\begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = \begin{pmatrix} \alpha & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \Leftrightarrow \begin{cases} \dot{x} = \alpha x \\ \dot{y} = -y \end{cases}$$

→ Sys are uncoupled.

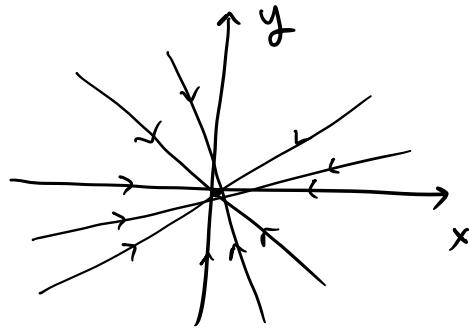
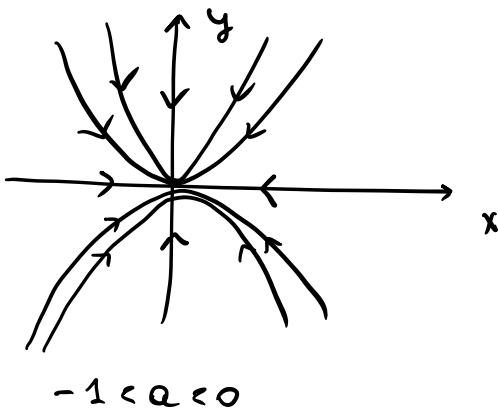
$$\begin{cases} x(t) = x_0 e^{\alpha t} \\ y(t) = y_0 e^{-t} \end{cases} \quad \begin{pmatrix} x_0 \\ y_0 \end{pmatrix} \in \mathbb{R}^2 \text{ initial point}$$

- $y(t)$  decays exponentially.
  - When  $\alpha < 0$  also  $x(t)$  decays exponentially  $\Rightarrow$  All trajectories approach the origin as  $t \rightarrow +\infty$ .
- However THE DIRECTION OF APPROACH DEPENDS ON THE SIZE OF  $\alpha < 0$  COMPARED TO  $-1$ !



$$\alpha < -1$$

↓  $x(t)$  decays more rapidly than  $y(t)$



$a < 0$   $(0,0)$  stable  
(attractor)

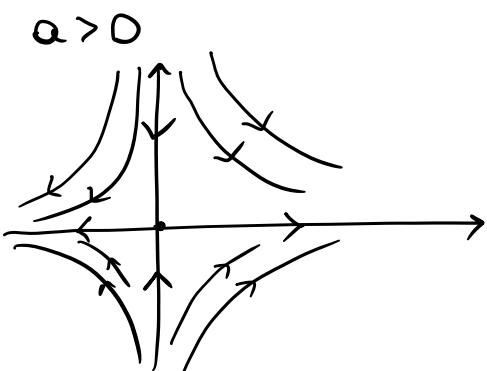
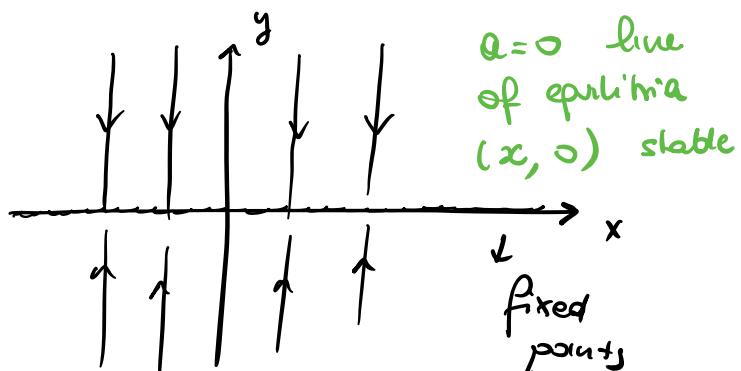
$$a = -1 \downarrow$$

$$\begin{cases} x(t) = x_0 e^{-t} \\ y(t) = y_0 e^{-t} \end{cases}$$

$$\begin{pmatrix} x(t) \\ y(t) \end{pmatrix} = e^{-t} \begin{pmatrix} x_0 \\ y_0 \end{pmatrix}$$

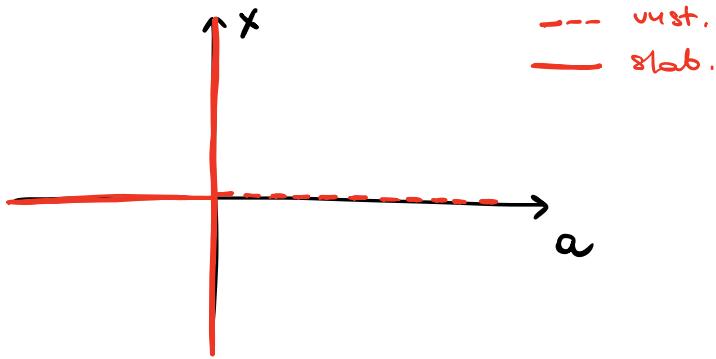
$$a = 0$$

$$\begin{cases} x(t) = x_0 \\ y(t) = y_0 e^{-t} \end{cases}$$



$a > 0 \exists ! \text{ eq. } (0,0) \text{ unstable}$

By consid. only the eq. configuration  $x \in \mathbb{R}$ , the  
bif. diagram is the following:



Classification of linear systems,  $A$   $2 \times 2$  diagonalizable

$$\begin{cases} \dot{z} = Az & z \in \mathbb{R}^2 \\ z(0) = z_0 & A, 2 \times 2 \text{ matrix diagonal.} \\ & \text{We know: } \varphi^t(z_0) = \underbrace{e^{tA}}_{\parallel} z_0 \\ & \exp(tA) \end{cases}$$

$A$  diag. (on  $\mathbb{R}$  or  $\mathbb{C}$ )  $\Leftrightarrow A$  has 2 eigenvectors  
(on  $\mathbb{R}$  or  $\mathbb{C}$ ) linearly independent.

Recall the previous example:  $x$  and  $y$  axes play a crucial role since they are INVARIANT lines for the dynamics.

For the general case  $\dot{z} = Az$ , we would like to find an analog to these axes. In particular, we seek for trajectories of this form:

$$z(t) = e^{\lambda t} v \quad v \in \mathbb{C}^2 \quad \lambda \in \mathbb{C}$$

To find conditions on  $v$  and  $\lambda$ , we impose that  $z(t)$  is a solution of  $\dot{z} = Az$ .

$$\lambda e^{\lambda t} v = A e^{\lambda t} v \quad \Leftrightarrow \quad Av = \lambda v$$

The desired invariant lines exist iff  $\sigma$  is  
an eigenvector of  $A$  with eigenvalue  $\lambda$

First case  $\lambda_1, \lambda_2 \in \mathbb{R}$ ,  $\lambda_1 \neq \lambda_2$  and  $\lambda_1, \lambda_2 \neq 0$ .  
 $\rightarrow$  eigenvalues of  $A$

This means that  $A$  has 2 real eigenvectors  $v_1, v_2 \in \mathbb{R}^2$   
linearly independent.