

Example: Solve
$$\begin{cases} x_1'(t) = 7x_1(t) + 5x_2(t) \\ x_2'(t) = -10x_1(t) - 8x_2(t) \end{cases}$$

with $x_1(0) = 1, x_2(0) = 2.$

ODEs and
difference equations.

Solution:
$$\vec{x}'(t) = \begin{bmatrix} x_1'(t) \\ x_2'(t) \end{bmatrix} = \underbrace{\begin{bmatrix} 7 & 5 \\ -10 & -8 \end{bmatrix}}_A \vec{x}(t), \quad \vec{x}(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}.$$

$$\hookrightarrow \vec{x}(t) = e^{At} \vec{x}(0) \rightarrow \text{Find } e^{At}.$$

We first try to diagonalize A :

$$A - \lambda I = \begin{bmatrix} 7-\lambda & 5 \\ -10 & -8-\lambda \end{bmatrix} \rightarrow -(7-\lambda)(8+\lambda) + 50 = \lambda^2 + \lambda - 6 = 0$$

$$\Leftrightarrow \lambda = \frac{-1 \pm \sqrt{25}}{2} \rightarrow \begin{cases} \lambda_1 = -3 \\ \lambda_2 = 2 \end{cases} \quad (\Leftrightarrow \text{we can diagonalize})$$

Eigenvectors:

$$A - \lambda_1 I = \begin{bmatrix} 10 & 5 \\ -10 & -5 \end{bmatrix} \rightarrow \vec{u}_1 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

$$A - \lambda_2 I = \begin{bmatrix} 5 & 5 \\ -10 & -10 \end{bmatrix} \rightarrow \vec{u}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Therefore,
$$S = \begin{bmatrix} -1 & 1 \\ 2 & -1 \end{bmatrix}, \quad D = \begin{bmatrix} -3 & 0 \\ 0 & 2 \end{bmatrix}, \quad S^{-1} = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix}$$

and so

$$e^{At} = Se^{Dt}S^{-1} = \begin{bmatrix} -1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} e^{-3t} & 0 \\ 0 & e^{2t} \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix}$$

Remark: The solution can be written as follows

$$\vec{x}(t) = e^{At} \vec{x}_0 = \underbrace{\begin{bmatrix} -1 & 1 \\ 2 & -1 \end{bmatrix}}_S \underbrace{\begin{bmatrix} e^{-3t} & 0 \\ 0 & e^{2t} \end{bmatrix}}_D \underbrace{\begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix}}_{S^{-1}} \underbrace{\begin{bmatrix} 1 \\ 2 \end{bmatrix}}_{\vec{x}_0|_S} =$$

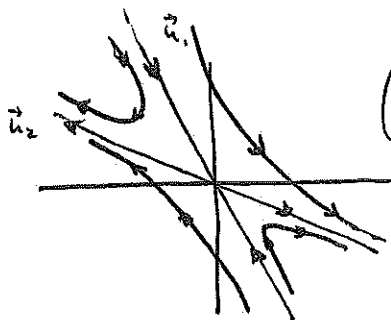
$$= \begin{bmatrix} -e^{-3t} & e^{2t} \\ 2e^{-3t} & -e^{2t} \end{bmatrix} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = 3e^{-3t} \begin{bmatrix} -1 \\ 2 \end{bmatrix} + 4e^{2t} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

\vec{u}_1 \vec{u}_2

coordinates of \vec{x}_0 in $\{\vec{u}_1, \vec{u}_2\}$

This shows us how the solution evolves in time: the \vec{u}_1 component goes to zero since its corresponding eigenvalue is negative.

Phase portrait:



-155-

(Saddle point)

↳ Always the case when

$$\lambda_1 \lambda_2 < 0$$

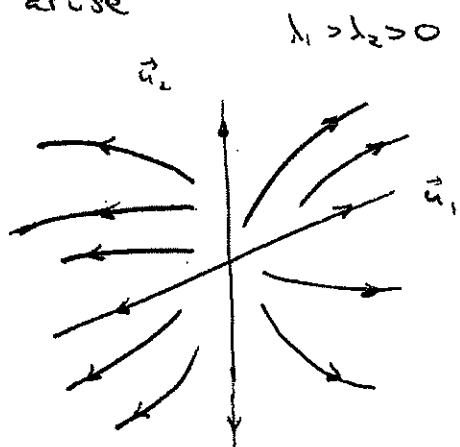
(i.e. distinct signs)

Remark: If A is diagonalizable, the solution can always be written as

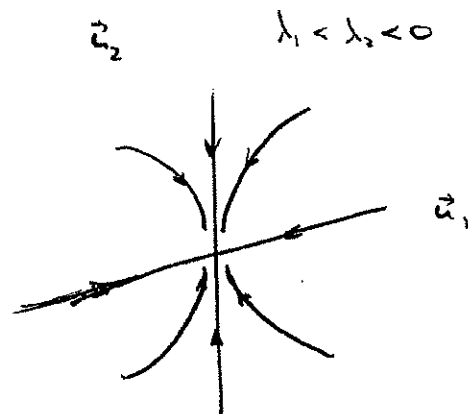
$$\vec{x}(t) = Se^{At}S^{-1}\vec{x}_0 = [\vec{u}_1 | \vec{u}_2] \begin{bmatrix} e^{\lambda_1 t} & \\ & e^{\lambda_2 t} \end{bmatrix} \vec{x}_0|_S =$$

$$= c_1 e^{\lambda_1 t} \vec{u}_1 + c_2 e^{\lambda_2 t} \vec{u}_2, \text{ where } \vec{x}_0|_S = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$$

Depending on the signs of λ_1, λ_2 , different phase portraits arise



(everything grows, but faster in \vec{u}_1 direction)



(everything goes to zero, but the \vec{u}_1 component disappears faster)

⋮

→ Exercise: 2nd order linear differential equation.

$$\left. \begin{array}{l} \text{Solve } x''(t) = -x(t) + 2x'(t) \\ \text{with } x(0) = 2, x'(0) = 1 \end{array} \right\}$$

Sol.: Right now, we only know how to solve first order linear systems.

Let's convert this 2nd order diff. eq. into a 1st order system:

$$\left. \begin{array}{l} \text{Define } x_1(t) = x(t) \\ x_2(t) = x'(t) \end{array} \right\}$$

$$\text{Then, } x_1'(t) = x'(t) = x_2(t)$$

$$x_2'(t) = x''(t) = -x(t) + 2x'(t) = -x_1(t) + 2x_2(t)$$

use the equation introduce the new variables

So, in matrix form,

$$\underbrace{\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}}_{\vec{x}'(t)} = \underbrace{\begin{bmatrix} 0 & 1 \\ -1 & 2 \end{bmatrix}}_A \underbrace{\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}}_{\vec{x}(t)} \quad \text{and thus } \vec{x}(t) = e^{At} \vec{x}_0.$$

Remark: This idea can be applied to any n -order lin. diff. equation with constant coeff. to produce ~~a~~ a 1st order lin. system (of n equations)

$$a_0 x(t) + a_1 x'(t) + \dots + a_n x^{(n)}(t) = 0$$

$$\begin{cases} \text{Define } x_1(t) = x(t) \\ x_2(t) = x'(t) \\ \vdots \\ x_n(t) = x^{(n-1)}(t) \end{cases} \Rightarrow \vec{x}'(t) = \begin{bmatrix} x_1'(t) \\ \vdots \\ x_n'(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -\frac{a_0}{a_n} & -\frac{a_1}{a_n} & \dots & -\frac{a_{n-1}}{a_n} & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{bmatrix}$$

Let's finish the exercise. We need e^{At} .

Try to diagonalise:

$$\text{Eigenvalues of } A: \det(A - \lambda I) = 0 \Leftrightarrow -\lambda(2-\lambda) + 1 = 0 \Leftrightarrow$$

$$\Leftrightarrow \lambda^2 - 2\lambda + 1 = 0 \Leftrightarrow (\lambda - 1)^2 = 0$$

$$\lambda = 1 \text{ (repeated)}$$

Eigenvectors: $(A - I)\vec{u} = \vec{0} \rightarrow \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}$ has rank 1! only one eigenvector.

|| So A cannot be diagonalised.

We have to use definition of $e^{At} = I + At + \frac{A^2 t^2}{2!} + \dots$

However,

$$A^2 = \begin{bmatrix} -1 & 2 \\ -2 & 3 \end{bmatrix}, A^3 = \begin{bmatrix} -2 & 3 \\ -3 & 4 \end{bmatrix}, \dots \text{ infinite terms?!}$$

• Nice fact: (Cayley-Hamilton Theorem)

$$(A-I)^2 = \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \left[\begin{array}{l} \text{A matrix } A \text{ always satisfies} \\ \text{its own characteristic equation:} \\ \det(A - \lambda I) = 0 \end{array} \right]$$

(here $\det(A - \lambda I) = (\lambda - 1)^2 = 0$)

So we can write $A = I + (A-I)$ and so,

$$e^{At} = e^{I t + (A-I)t} \stackrel{(*)}{=} e^{I t} e^{(A-I)t} =$$

$$= e^t I \left(I + (A-I)t + \frac{(A-I)^2 t^2}{2!} + \dots \right) =$$

$$= e^t \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix} t \right) = e^t \begin{bmatrix} 1-t & t \\ -t & 1-t \end{bmatrix}$$

$$\left(\begin{array}{l} (*) \quad e^{A+B} = e^A e^B \iff A \text{ and } B \text{ commute} \\ \text{(i.e., } AB = BA) \end{array} \right)$$

We conclude that

$$\begin{aligned}\vec{x}(t) &= e^{At} \vec{x}_0 = e^t \begin{bmatrix} 1-t & t \\ -t & 1+t \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = e^t \begin{bmatrix} 2-2t+t \\ -2t+1+t \end{bmatrix} = e^t \begin{bmatrix} 2-t \\ 1-t \end{bmatrix} \\ &= e^t \underbrace{\begin{bmatrix} 2 \\ 1 \end{bmatrix}}_{\vec{x}_0} - t e^t \underbrace{\begin{bmatrix} 1 \\ 1 \end{bmatrix}}_{\vec{u}} \\ &\quad \left(\text{of course } \vec{x}(0) = e^0 \vec{x}_0 - 0 = \vec{x}_0 \right)\end{aligned}$$

V Difference equations (sequences)

$$\left. \begin{array}{l} \vec{x}(k) = A \vec{x}(k-1) \\ \vec{x}(0) = \vec{x}_0 \end{array} \right\} \begin{array}{l} (\dots \text{ like differential equations} \\ \text{ with discrete time... bla bla...}) \end{array}$$

Solution: $\vec{x}(1) = A \vec{x}(0) = A \vec{x}_0$

$$\vec{x}(2) = A \vec{x}(1) = AA \vec{x}_0 = A^2 \vec{x}_0$$

$$\parallel \vec{x}(k) = A^k \vec{x}_0 \parallel$$

If $A = SDS^{-1}$, then

$$\vec{x}(k) = S D^k S^{-1} \vec{x}_0 = S \begin{bmatrix} \lambda_1^k & 0 \\ 0 & \lambda_2^k \end{bmatrix} S^{-1} \vec{x}_0$$

Example: Virshanca (or Fibonacci) numbers.

Sequence $0, 1, 1, 2, 3, 5, 8, \dots, N_k, N_{k+1}, N_{k+2}, \dots$

$$N_k > \frac{N_k}{N_{k-1}} >$$

Sol: The sequence is defined by $N_{k+2} = N_{k+1} + N_k$.

This is a 2nd order difference equation (as it involves two steps).

But we "already know" how to solve that problem: convert it into two 1st order equations.

Denote

$$\vec{x}(k) = \begin{bmatrix} N_{k+1} \\ N_{k+2} \end{bmatrix}, \text{ then}$$

$$\vec{x}(k) = \begin{bmatrix} N_{k+1} \\ N_{k+2} \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}}_A \underbrace{\begin{bmatrix} N_k \\ N_{k+1} \end{bmatrix}}_{\vec{x}(k-1)} \text{ with } \vec{x}(0) = \begin{bmatrix} N_1 \\ N_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

So the solution is

$$\vec{x}(k) = A^k \vec{x}_0 \rightarrow A^k \text{ ? Diagonalize:}$$

1) Eigenvalues: $\det(A - \lambda I) = 0 \Leftrightarrow$

$$\Leftrightarrow \lambda^2 - \lambda - 1 = 0 \Leftrightarrow \lambda = \frac{1 \pm \sqrt{5}}{2} \rightarrow \begin{cases} \lambda_1 = \frac{1 + \sqrt{5}}{2} \\ \lambda_2 = \frac{1 - \sqrt{5}}{2} \end{cases}$$

2) Eigenvectors:

$$(A - \lambda_1 I) = \begin{bmatrix} -\lambda_1 & 1 \\ 1 & 1 - \lambda_1 \end{bmatrix} \rightarrow \vec{u}_1 = \begin{bmatrix} 1 \\ \lambda_1 \end{bmatrix} \quad (\text{note that } 1 + \lambda_1 - \lambda_1^2 = 0)$$

Analogously, $\vec{u}_2 = \begin{bmatrix} 1 \\ \lambda_2 \end{bmatrix}$

In conclusion,

$$\vec{x}(k) = \underbrace{\begin{bmatrix} 1 & 1 \\ \lambda_1 & \lambda_2 \end{bmatrix}}_S \begin{bmatrix} \lambda_1^k & 0 \\ 0 & \lambda_2^k \end{bmatrix} (S^{-1} \vec{x}_0) \quad \text{where } S^{-1} \vec{x}_0 = \vec{x}_0|_S = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix},$$

(just a matter of computing)

thus

$$\vec{x}(k) = \begin{bmatrix} \lambda_1^k \vec{u}_1 & \lambda_2^k \vec{u}_2 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = c_1 \lambda_1^k \vec{u}_1 + c_2 \lambda_2^k \vec{u}_2.$$

Therefore,

$$N_{k+2} = c_1 \lambda_1^{k+1} + c_2 \lambda_2^{k+1} \quad (\text{2nd component of } \vec{x}(k))$$

and

$$\frac{N_{k+2}}{N_{k+1}} = \frac{c_1 \lambda_1^{k+1} + c_2 \lambda_2^{k+1}}{c_1 \lambda_1^k + c_2 \lambda_2^k} = \frac{c_1 \lambda_1 + c_2 \lambda_2 \left(\frac{\lambda_2^k}{\lambda_1^k}\right)}{c_1 + c_2 \left(\frac{\lambda_2}{\lambda_1}\right)^k} \rightarrow \lim_{k \rightarrow \infty} \frac{N_{k+2}}{N_{k+1}} = \lambda_1 = \frac{1 + \sqrt{5}}{2}$$

("golden ratio")

(because $\lambda_2 < \lambda_1$)

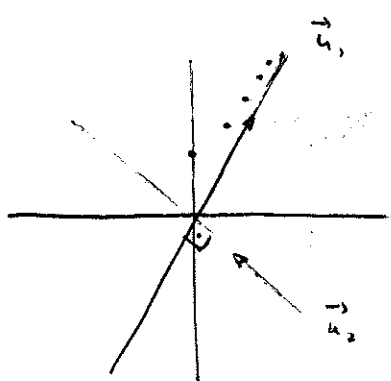
Forgetting about Virahana or Fibonacci, we could think of

$$\vec{x}(k) = \begin{bmatrix} x(k) \\ y(k) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x(k-1) \\ y(k-1) \end{bmatrix} \quad \left. \vphantom{\vec{x}(k)} \right\} \text{ as a system,}$$

$$\vec{x}(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

whose solution $\vec{x}(k) = c_1 \lambda_1^k \vec{u}_1 + c_2 \lambda_2^k \vec{u}_2$ can be drawn as "time" evolves (k increases):

$$\vec{x}(1) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad \vec{x}(2) = \begin{bmatrix} 3 \\ 5 \end{bmatrix}, \dots$$



The \vec{u}_2 component goes to zero as $|\lambda_2| < 1$.

→ Notice that $\vec{u}_1 \cdot \vec{u}_2 = 1 + \lambda_1 \lambda_2 = \frac{1+\sqrt{5}}{2} \frac{1-\sqrt{5}}{2} = 1-1=0$, that is, $\vec{u}_1 \perp \vec{u}_2$.

This is not a coincidence: $A = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$ is symmetric with real entries.