

1. Existence and Uniqueness of sol'n.

• let $x_1(t), \dots, x_n(t)$ be the unknown function. $\vec{x}(t) = \begin{pmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{pmatrix}$

• let
$$\begin{cases} \dot{x}_1(t) = F_1(t, x_1, \dots, x_n) \\ \dot{x}_2(t) = F_2(t, x_1, \dots, x_n) \\ \vdots \\ \dot{x}_n(t) = F_n(t, x_1, \dots, x_n) \end{cases}$$

be the equations.

Thm: Suppose $\{F_i(t, \vec{x})\}$ have (partial) continuous derivatives.

in t and x_1, \dots, x_n , then for any initial condition at $t = t_0$, given by $(x_1(t_0), \dots, x_n(t_0))$, we have a small ^{open} segment $(t_0 - \epsilon, t_0 + \epsilon)$ containing t_0 , such that the sol'n $\vec{x}(t)$ exists for $t \in (t_0 - \epsilon, t_0 + \epsilon)$ that satisfies the eqn & initial condition. Such sol'n is unique.

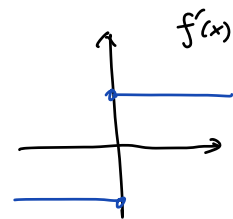
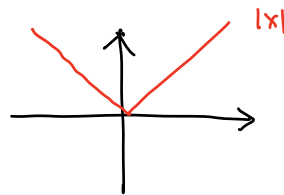
Ex: a function that does not have continuous derivative:

• $f(x) = \frac{1}{x}$ $x \in (-\infty, +\infty)$.

function is not continuous at $x=0$

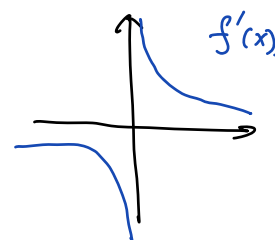
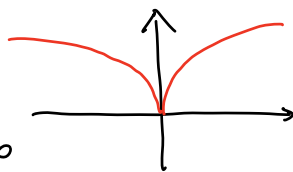
$\Rightarrow f'(x)$ does not exist at $x=0$

• $f(x) = |x|$



• $f(x) = |x|^{\frac{1}{2}}$

$$f'(x) = \begin{cases} \frac{1}{2} \frac{1}{x} & x > 0 \\ -\frac{1}{2} \frac{1}{x} & x < 0. \end{cases}$$



• Ex : $\begin{cases} \dot{x}(t) = [x(t)]^{\frac{2}{3}} \\ x(0) = 0 \end{cases}$

diff eq.

initial condition

we have 2 sol'n.

$$x(t) = 0 \quad \forall t.$$

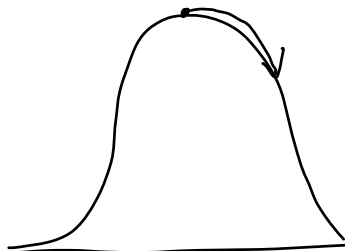
or

$$x(t) = \left(\frac{t}{3}\right)^3 \Rightarrow \dot{x}(t) = \left(\frac{t}{3}\right)^2$$

We don't have uniqueness here, since

$F(t, x) = x^{\frac{2}{3}}$ doesn't have continuous derivative at $x=0$.

Picture:



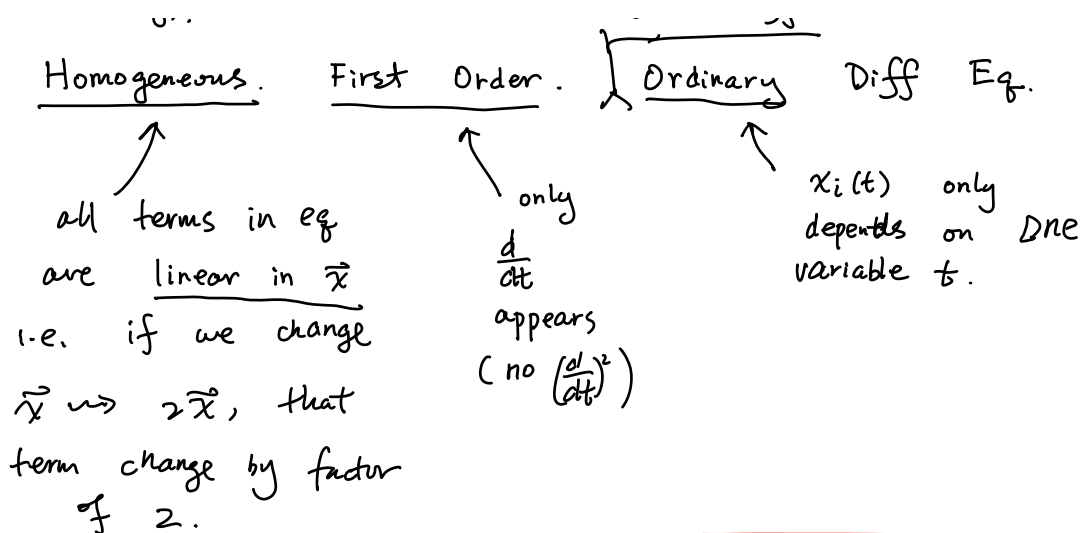
roll down the hill
or stay there.

(2)

$$\frac{d}{dt} \begin{pmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{pmatrix} = \underbrace{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & & & \vdots \\ \vdots & & & \vdots \\ \vdots & & & \vdots \end{bmatrix}}_A \begin{pmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{pmatrix}$$

$\{a_{ij}\}$ are constants.

const coeff \swarrow coeff is indep of t .

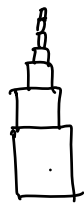


$$\frac{d}{dt} \vec{x}(t) = A \cdot \vec{x}(t) \Rightarrow \vec{x}(t) = e^{A \cdot t} \cdot \vec{x}(0)$$

Here, since A is an $n \times n$ matrix with constant coeff.

$$e^x = 1 + x + \frac{x^2}{2 \cdot 1} + \frac{x^3}{3 \cdot 2 \cdot 1} + \frac{x^4}{4!} + \dots$$

(this series converges for all $x \in \mathbb{C}$.)



if each term has size that shrinks fast enough, then the total size is finite.

$$e^{A \cdot t} = I_n + At + \frac{(At)^2}{2!} + \frac{(At)^3}{3!} + \dots$$

- each term make sense $\because \frac{(At)^n}{n!} = \frac{t^n}{n!} \cdot \underbrace{A^n}_{\text{well defined}}$
- summation converges. (need proof).

$$\frac{d}{dt}(e^{A \cdot t}) = 0 + A + A^2 \cdot t + A^3 \cdot \frac{t^2}{2!} + \dots$$

$$= A \cdot (I + At + A^2 \frac{t^2}{2!} + \dots)$$

$$= A \cdot e^{At}$$

$$\left[\frac{d}{dt} e^{\lambda t} = \lambda \cdot e^{\lambda t} \right]$$

$$\frac{d}{dt} (e^{At}) = A \cdot e^{At} \quad \text{then}$$

↑
size $n \times n$ matrix
depends on t

$$\frac{d}{dt} \left(\underbrace{e^{At}}_{\vec{x}(t)} \cdot \vec{x}_0 \right) = A \cdot \underbrace{e^{At}}_{\vec{x}(t)} \cdot \vec{x}_0$$

$$\Rightarrow \frac{d}{dt} \vec{x}(t) = A \cdot \vec{x}(t)$$

notation:

e^{At} is called the fundamental sol'n. since, each column vector in e^{At} is a sol'n, the general sol'n $\vec{x}(t) = e^{At} \vec{x}_0$ is a linear comb. of the columns

In practice, to do exponentiation of a matrix A , we want to diagonalize A first.

Recall Jordan Decomposition Thm:

Given a ^{square} matrix A over \mathbb{C} of size n , there exist an invertible matrix C , s.t.

$$A = C \cdot J \cdot C^{-1}$$

where J is block diagonal, with each block $\sim \begin{pmatrix} \lambda & & 0 \\ & \ddots & \\ 0 & & \lambda \end{pmatrix}$

$$A^2 = (C \cdot J \cdot C^{-1}) \cdot (C \cdot J \cdot C^{-1})$$

$$= C \cdot J \cdot J \cdot C^{-1} = C \cdot J^2 \cdot C^{-1}$$

$$\vdots$$

$$A^n = (C \cdot J \cdot C^{-1}) \underbrace{(C \cdot J \cdot C^{-1})}_{\dots} \dots \underbrace{(C \cdot J \cdot C^{-1})}_{\dots}$$

$$= C \cdot J^n \cdot C^{-1}$$

$$e^{At} = 1 + At + A^2 \cdot \frac{t^2}{2!} + \dots$$

$$= C \left[1 + Jt + J^2 \cdot \frac{t^2}{2!} + \dots \right] \cdot C^{-1}$$

$$= C \cdot e^{Jt} \cdot C^{-1}$$

$$\left(\begin{array}{c|c} B_1 & 0 \\ \hline 0 & B_2 \end{array} \right)^2 = \begin{pmatrix} B_1 & 0 \\ 0 & B_2 \end{pmatrix} \begin{pmatrix} B_1 & 0 \\ 0 & B_2 \end{pmatrix} = \begin{pmatrix} B_1^2 & 0 \\ 0 & B_2^2 \end{pmatrix}$$

Suffice to consider the case where A is of Jordan form.

Suffice to consider one single Jordan block:

<u>Jordan block size</u>	<u>J</u>	J^n	e^{Jt}
1	(λ)	λ^n	$e^{\lambda t}$
2	$\begin{pmatrix} \lambda & 1 \\ 0 & \lambda \end{pmatrix}$	$\begin{pmatrix} \lambda^n & n \cdot \lambda^{n-1} \\ 0 & \lambda^n \end{pmatrix}$	$e^{\lambda t} \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix}$

$$\begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix} = \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} + \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$

$$= \lambda \cdot I_2 + N_2$$

\uparrow regular nilpotent matrix of size 2.

we know $I \cdot N = N \cdot I = N$.

$$(N_2)^2 = 0.$$

$$\begin{aligned} (\lambda I + N)^2 &= (\lambda + N)^2 = \lambda^2 + 2\lambda N + N^2 \\ &= \lambda^2 + 2\lambda N \\ &= \begin{pmatrix} \lambda^2 & 0 \\ 0 & \lambda^2 \end{pmatrix} + 2\lambda \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} \lambda^2 & 2\lambda \\ 0 & \lambda^2 \end{pmatrix} \end{aligned}$$

$$\begin{aligned} (\lambda I + N)^k &= (\lambda + N)^k = \lambda^k + \binom{k}{1} \lambda^{k-1} \cdot N + \underbrace{\binom{k}{2} \lambda^{k-2} \cdot N^2 + \dots}_{=0 \quad \because N^2=0} \\ &= \begin{pmatrix} \lambda^k & k \cdot \lambda^{k-1} \\ 0 & \lambda^k \end{pmatrix} \end{aligned}$$

$$e^{(\lambda+N)t} = I + (\lambda+N)t + \frac{(\lambda+N)^2 \cdot t^2}{2} + \dots$$

$$e^{(\lambda t \cdot I) + (t \cdot N)} = e^{\lambda t \cdot I} \cdot e^{t \cdot N}$$

If A, B are 2 square matrices, and if $AB = BA$,

then $e^{A+B} = e^A \cdot e^B$

\therefore if x, y are numbers,

$$\begin{aligned} e^x \cdot e^y &= \left(1 + x + \frac{x^2}{2!} + \dots \right) \left(1 + y + \frac{y^2}{2!} + \dots \right) \\ &= 1 + (x+y) + \frac{1}{2!} (x^2 + y^2 + 2xy) + \dots \\ &= e^{x+y} \end{aligned}$$

the only property we use, is $xy = yx$.

Hence the same argument works for two commuting matrices. A, B .

$$e^{tN} = 1 + tN + \underbrace{\frac{t^2}{2}N^2 + \dots}_{=0} = 1 + tN = \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix}$$

$$e^{(\lambda+N)t} = e^{\lambda t} \cdot \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix}.$$

For $\overset{\text{size } m}{J} = \lambda I_m + N_m = \left(\begin{array}{ccc} \lambda & & \\ & \ddots & \\ & & \lambda \end{array} \right) \Bigg\}^m$

$$J^k = \begin{pmatrix} \lambda^k & & 0 \\ & \ddots & \\ 0 & & \lambda^k \end{pmatrix} + k \cdot \lambda^{k-1} \cdot N + \binom{k}{2} \cdot \lambda^{k-2} \cdot N^2 + \dots$$

$$N_4 = \begin{pmatrix} 0 & 1 & & \\ & 0 & 1 & \\ & & 0 & 1 \\ & & & 0 \end{pmatrix}, \quad (N_4)^2 = \begin{pmatrix} 0 & 0 & 1 & \\ & 0 & 0 & 1 \\ & & 0 & 0 \\ & & & 0 \end{pmatrix}, \quad N_4^3 = \begin{pmatrix} 0 & 0 & 0 & 1 \\ & 0 & 0 & 0 \\ & & 0 & 0 \\ & & & 0 \end{pmatrix}$$

$$, N_4^4 = 0.$$



$$= \begin{pmatrix} \lambda^k & k \cdot \lambda^{k-1} & \binom{k}{2} \lambda^{k-2} & \dots \\ & \lambda^k & k \cdot \lambda^{k-1} & \binom{k}{2} \lambda^{k-2} \\ & & \ddots & \ddots \\ & & & \lambda^k \end{pmatrix}$$

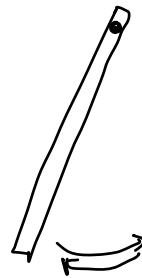
[ODE]. 3.6.2

Ex: (damped pendulum).

Eg

$$\ddot{x} = \underbrace{-kx}_{\text{Hook's Law.}} - \underbrace{c \cdot \dot{x}}_{\text{resistance}}$$

$$k, c > 0.$$



- First, we turn this to a 1st order diff eq.

let $x_2(t) = \dot{x}_1(t)$, then

$$x_2(t) = \dot{x}_1(t)$$

$$\begin{aligned} \frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} x_2 \\ \ddot{x}_1 \end{pmatrix} = \begin{pmatrix} x_2 \\ -kx_1 - c \cdot x_2 \end{pmatrix} \\ &= \begin{pmatrix} 0 & 1 \\ -k & -c \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \end{aligned}$$

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -k & -c \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = \underbrace{e^{\begin{pmatrix} 0 & 1 \\ -k & -c \end{pmatrix} t}}_{\substack{\text{evolution} \\ \text{matrix} \\ \text{or propagator.}}} \cdot \begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix}$$

$$A = \begin{pmatrix} 0 & 1 \\ -k & -c \end{pmatrix}$$

Diagonalize A , by

$$A = C \cdot J \cdot C^{-1}$$

$$0 = \det(\lambda I - A) = \det \begin{pmatrix} \lambda & -1 \\ k & \lambda + c \end{pmatrix} = \lambda(\lambda + c) + k = \lambda^2 + c\lambda + k$$

$$\ddot{x} + c \cdot \dot{x} + kx = 0$$

try $x = e^{\lambda t}$, get

$$\text{Solve } \lambda^2 + c\lambda + k = 0, \quad \lambda^2 + c \cdot \lambda + \left(\frac{c}{2}\right)^2 = \left(\frac{c}{2}\right)^2 - k$$

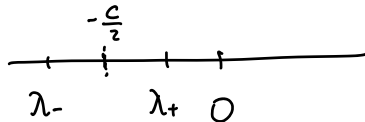
$$\left(\lambda + \frac{c}{2}\right)^2 = \left(\frac{c}{2}\right)^2 - k$$

$$\lambda + \frac{c}{2} = \pm \sqrt{\left(\frac{c}{2}\right)^2 - k}$$

$$\lambda_{\pm} = -\frac{c}{2} \pm \sqrt{\left(\frac{c}{2}\right)^2 - k}$$

case 1: λ_{\pm} are distinct real sol'n.

$$\Leftrightarrow k < \left(\frac{c}{2}\right)^2 \quad \left(\text{intuitively this means friction is large} \right)$$



$$A = C \begin{pmatrix} \lambda_+ & 0 \\ 0 & \lambda_- \end{pmatrix} C^{-1}$$

$$e^{At} = C \cdot \begin{pmatrix} e^{\lambda_+ t} & 0 \\ 0 & e^{\lambda_- t} \end{pmatrix} C^{-1}$$

$$\vec{x}(t) = C \cdot \begin{pmatrix} e^{\lambda_+ t} & 0 \\ 0 & e^{\lambda_- t} \end{pmatrix} C^{-1} \begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix}$$

case 2: λ_{\pm} are a pair of complex conjugate sol'n

$$\Leftrightarrow k > \left(\frac{c}{2}\right)^2$$

$$\lambda_{\pm} = -\rho \pm i\omega$$

$$\omega = \sqrt{k - \left(\frac{c}{2}\right)^2}$$

$$e^{(-\rho + i\omega)t} = e^{-\rho t} \left[e^{i\omega t} \right]$$

$$= e^{-\rho t} \left[\cos(\omega t) + i \cdot \sin(\omega t) \right]$$

$$\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = C \begin{pmatrix} e^{\lambda_+ t} & 0 \\ 0 & e^{\lambda_- t} \end{pmatrix} C^{-1} \begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix}$$

$$\Rightarrow x(t) = x_1(t) = \underbrace{c_1 e^{-\rho t} \sin(\omega t)}_{\text{Im}} + \underbrace{c_2 e^{-\rho t} \cos(\omega t)}_{\text{Re}}$$

$$= \alpha e^{\lambda_+ t} + \bar{\alpha} e^{\lambda_- t}$$

case 3 : $\lambda_+ = \lambda_- = -\frac{c}{2}$. $K = \left(\frac{c}{2}\right)^2$.

~~$J = \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix}$~~ or $J = \begin{pmatrix} \lambda & 1 \\ 0 & \lambda \end{pmatrix}$

$\Rightarrow C \cdot J \cdot C^{-1} = J \neq A = \begin{pmatrix} 0 & 1 \\ -k & -c \end{pmatrix}$

$A = C \cdot J \cdot C^{-1}$

$e^{Jt} = e^{[\lambda + \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}]t} = e^{\lambda t} \cdot e^{\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}t} = e^{\lambda t} \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix}$

$\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = C \cdot e^{Jt} \cdot C^{-1} \begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix}$

$x_1(t) = \underline{c_1} \cdot e^{\lambda t} + \underline{c_2} \cdot e^{\lambda t} \cdot t$

$\ddot{x} + a_1 \dot{x} + a_2 x + a_3 \cdot x = 0$

introduce new functions

$\begin{cases} x_1(t) = x(t) \\ x_2(t) = \dot{x}_1(t) = \dot{x}(t) \\ x_3(t) = \dot{x}_2(t) = \ddot{x}(t) \end{cases}$

$\dot{x}_3 = \ddot{x}(t) = -a_1 \dot{x} - a_2 x - a_3 x = -a_1 x_3 - a_2 x_2 - a_3 x_1$

~~$x_4(t) = \dot{x}_3(t) = \ddot{x}(t) = \dots$~~

$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_2 \\ x_3 \\ -a_1 x_3 - a_2 x_2 - a_3 x_1 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -a_3 & -a_2 & -a_1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$

3.6.3

higher order

$\det(\lambda - A) = (\lambda - \lambda_1)^{m_1} \dots (\lambda - \lambda_r)^{m_r}$

$x(t)$ is a linear combination of e

$J = \begin{pmatrix} \lambda_1 & \dots & | & \\ \dots & \dots & | & \\ \hline & & \lambda_2 & \dots & \lambda_r \end{pmatrix}$