

## Review of linear homogeneous ODEs (review, Chapter 2)

$$(*) \quad \dot{X} = A(t)X$$

$$X \in \mathbb{R}^n$$

$$A(t) = n \times n \text{ (continuous) matrix}$$

$X=0$  is a soln. (homogeneous)

if  $X_1(t)$  &  $X_2(t)$  satisfy  $(*)$ , then so does  $c_1 X_1(t) + c_2 X_2(t)$  for any  $c_1, c_2 \in \mathbb{R}$  (linear superposition principle)

If we can find  $n$  linearly independent solns.  $X_1(t), X_2(t), \dots, X_n(t)$  then we have found the general soln. to  $(*)$ , which can be written as

$$X = \sum_{k=1}^n c_k X_k(t)$$

Special case  $A = \text{constant matrix}$

### • Approach 1

compute eigenvalues of  $A$  as roots of the characteristic eqn.  $\text{Det}(A - \lambda I) = 0$   
if none are repeated, then we know we can write the general soln. as

$$X(t) = c_1 e^{\lambda_1 t} v_1 + c_2 e^{\lambda_2 t} v_2 + \dots + c_n e^{\lambda_n t} v_n$$

where  $v_1, \dots, v_n$  are linearly independent eigenvectors  $Av_j = \lambda_j v_j$

i.p. 
$$X(t) = \begin{bmatrix} e^{\lambda_1 t} v_1 & e^{\lambda_2 t} v_2 & \dots & e^{\lambda_n t} v_n \\ | & | & & | \end{bmatrix} \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}$$

initial condition  $X(0) = x_0$  determines

ie. 
$$\begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} \quad X_0 = \begin{bmatrix} | & | & \dots & | \\ v_1 & v_2 & \dots & v_n \\ | & | & & | \end{bmatrix} \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = \begin{bmatrix} | & \dots & | \\ v_1 & \dots & v_n \\ | & & | \end{bmatrix}^{-1} x_0$$

↑  
guaranteed to be invertible since eigenvectors are linearly independent

$$\Phi(t) = \begin{bmatrix} | & | & \dots & | \\ v_1 e^{\lambda_1 t} & v_2 e^{\lambda_2 t} & \dots & v_n e^{\lambda_n t} \\ | & | & & | \end{bmatrix}$$

= "Fundamental Matrix Soln," @

$\Phi(t)$  satisfies:

$$\dot{\Phi} = A \Phi$$

[nice convention to add to this:  $\Phi(0) = \text{Id}$ .  
then soln. to  $\dot{X} = AX$ ,  $X(0) = X_0$  is  
 $X = \Phi(t) X_0$ ]

side note:

if any eigenvalue  $\lambda$  is complex, ~~so that~~ <sup>e.g.</sup>  
 $\lambda = \alpha + i\beta \Rightarrow \lambda^* = \alpha - i\beta$  is also an eigenvalue  
 $Av = \lambda v$ ,  $Av^* = \lambda^* v^*$

then we can construct a real soln. from these as

$$c e^{\lambda t} v + c^* e^{\lambda^* t} v^*$$

& can rewrite as

$$e^{\alpha t} (a \cos(\beta t) + b \sin(\beta t))$$

or

$$e^{\alpha t} (A \cos(\beta t + \phi))$$

What if there is a repeated eigenvalue?  
then it could be that

geometric multiplicity < algebraic multiplicity  
dim. of eigenspace

$\Rightarrow$  not enough eigenvectors to construct  $n$  linearly

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independent solns.

Additional solns(s) obtained in terms of  
"generalized eigenvectors" & have form  
 $te^{\lambda t}$ ,  $t^2e^{\lambda t}$ , etc.

example:

$v$  is eigenvector of  $A$  w/ eigenvalue  $\lambda$   
(multiplicity 2)

$$(A - \lambda I)v = 0$$

$w$  is generalized eigenvector

$$\left. \begin{array}{l} (A - \lambda I)^2 w = 0 \\ (A - \lambda I)w \neq 0 \end{array} \right\} \begin{array}{l} (A - \lambda I)w = v \\ \downarrow \\ Aw = v + \lambda w \end{array}$$

$$\begin{aligned} x_1 &= e^{\lambda t} v \\ x_2 &= e^{\lambda t} (tv + w) \end{aligned}$$

$$\dot{x}_2 = \lambda e^{\lambda t} (tv + w) + e^{\lambda t} v = e^{\lambda t} (\lambda tv + \lambda w + v)$$

$$Ax_2 = e^{\lambda t} (t\lambda v + v + \lambda w) \quad \boxed{\checkmark}$$

if multiplicity 3:  $(A - \lambda I)^3 u = 0$

$$x_3 = e^{\lambda t} \left( \frac{t^2}{2} v + tw + u \right)$$

$\uparrow (A - \lambda I)^2 w = 0$        $\downarrow (A - \lambda I)^3 u = 0$

Approach 2.

$$\frac{dX}{dt} = AX, \quad X(0) = X_0$$

has soln.  $X(t) = e^{At} X_0$

matrix

$$\begin{aligned} \Rightarrow e^{At} &= I + At + \frac{1}{2} A^2 t^2 + \frac{1}{3!} A^3 t^3 + \dots \\ &= \sum_{k=0}^{\infty} \frac{A^k t^k}{k!} \quad (0! = 1) \end{aligned}$$

converges for all  $t$ .

Check:  $X(0) = e^{A \cdot 0} X_0 = I X_0 = X_0 \quad \checkmark$

$$\begin{aligned} \frac{dX}{dt} &= \lim_{h \rightarrow 0} \frac{X(t+h) - X(t)}{h} \\ &= \lim_{h \rightarrow 0} \frac{e^{A(t+h)} X_0 - e^{At} X_0}{h} \\ &= \lim_{h \rightarrow 0} \left( \frac{e^{At} e^{Ah} - e^{At}}{h} \right) X_0 \\ &= \lim_{h \rightarrow 0} \left( \frac{e^{Ah} - I}{h} \right) \underbrace{e^{At} X_0}_{X(t)} = A X(t) \\ &\quad \lim_{h \rightarrow 0} \left( \frac{Ah + \frac{1}{2} h^2 A^2 + \dots}{h} \right) = A \end{aligned}$$

Important :

I needed

$$e^{A(t+h)} = e^{At} e^{Ah} = e^{Ah} e^{At}$$

~~$e^A e^B$~~

$$e^A e^B \stackrel{?}{=} e^{(A+B)}$$

not necessarily true:

Baker-Campbell-Hausdorff  
Formula

$$e^A e^B = e^C$$

$$C = A+B + \frac{1}{2} [A, B] + \frac{1}{12} [A, [A, B]] + \dots$$

where

$$[A, B] = AB - BA$$

$$e^A e^B = e^{A+B}$$

only if  $[A, B] = 0$ , i.e.  
 $A$  &  $B$  commute

in our case

$$e^{A(t+h)} = e^{At} e^{Ah}$$

because  $[At, Ah] = th[A, A] = 0$

How do we evaluate  $e^{At}$ ?

simple case of complete set of eigenvectors  
( $v_1, v_2, \dots, v_n$ ) with eigenvalues ( $\lambda_1, \lambda_2, \dots, \lambda_n$ )

~~$A$~~

$\Rightarrow A$  can be diagonalized.

$$A = P \Lambda P^{-1}, \quad (P^{-1} A P = \Lambda)$$

$$\Lambda = \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \ddots \\ & & & \lambda_n \end{bmatrix} \quad P = \begin{bmatrix} | & | & & | \\ v_1 & v_2 & \dots & v_n \\ | & | & & | \end{bmatrix}$$

$\curvearrowright$  invertible since  
 $v_j$  are linearly  
independent

$$e^{At} = I + At + \frac{1}{2} A^2 t^2 + \frac{1}{3!} A^3 t^3 + \dots$$

$$= I + P \Lambda P^{-1} t + \frac{1}{2} (P \Lambda P^{-1}) (P \Lambda P^{-1}) t^2 + \dots$$

$$= P \left( I + \Lambda t + \frac{1}{2} \Lambda^2 t^2 + \frac{1}{3!} \Lambda^3 t^3 + \dots \right) P^{-1}$$

$$\left( \begin{array}{ccc} 1 + \lambda_1 t + \frac{1}{2} \lambda_1^2 t^2 + \dots & & \\ & \ddots & \\ & & 1 + \lambda_n t + \frac{1}{2} \lambda_n^2 t^2 + \dots \end{array} \right)$$

$$\left( \begin{array}{ccc} e^{\lambda_1 t} & & \\ & e^{\lambda_2 t} & \\ & & \ddots \\ & & & e^{\lambda_n t} \end{array} \right)$$

$$e^{At} = P \left( \begin{array}{ccc} e^{\lambda_1 t} & & \\ & e^{\lambda_2 t} & \\ & & \ddots \\ & & & e^{\lambda_n t} \end{array} \right) P^{-1}$$

But what if  $A$  is deficient & we can't diagonalize it?

e.g.  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  has only one eigenvector  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$

Thm (Proved in Chapter 2)

The matrix  $A$  on a complex vector space  $E$  has a unique decomposition,  $A = S + N$ , where  $S$  is semi-simple,  $N$  is nilpotent and  $[S, N] = 0$

semi-simple:  $P^{-1}SP = \Lambda = \text{diagonal}$

nilpotent:  $N^k = 0, N^{k-1} \neq 0$  nilpotency  $k$   
 ( $k=1$  is diagonalizable case)

$[S, N] = 0$  :  $SN = NS$

ex.  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ ,  $S = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = I$ ,  $N = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$   
 $N^2 = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \checkmark$

$e^{tA} = e^{tS} e^{tN}$  since  $[S, N] = 0$

$\nearrow \left( I + tN + \frac{1}{2}t^2N^2 + \dots \right) \swarrow$  terminates at  $k$ .



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

$$e^{tA} = e^{tS} e^{tN}$$

$$\begin{pmatrix} e^t & 0 \\ 0 & e^t \end{pmatrix} \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix} = I + tN$$

→ soln to  $\dot{X} = AX$  is  $X = e^{At} X_0$

$$X = \begin{pmatrix} e^t & te^t \\ 0 & e^t \end{pmatrix} X_0 = \begin{pmatrix} c_1 e^t + c_2 t e^t \\ c_2 e^t \end{pmatrix}$$

soln involving eigenvector

2<sup>nd</sup> independent soln. involving generalized eigenvector

§ homework problem:  $\dot{X} = A(t)X$   
 $A(t) = A(t+T), X \in \mathbb{R}^2$

eigenvalues of  $A(t)$  don't help in general

$$[A(t), A(s)] \neq 0 \text{ in general...}$$