

# Homogeneous Linear Systems With Constant Coefficients Contd...

Department of Mathematics  
IIT Guwahati

We shall extend techniques for scalar differential equations to systems.

For example, a GS to  $x'(t) = ax(t)$ , where  $a$  is a constant, is  $x(t) = ce^{at}$ . Analogously, we shall show that a GS to the system

$$\mathbf{x}'(t) = A\mathbf{x}(t),$$

where  $A$  is a constant matrix, is

$$\mathbf{x}(t) = e^{At}\mathbf{c}.$$

**Task:** To define the matrix exponential  $e^{At}$ .

Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a linear operator. The norm on  $T$  is given by

$$\|T\| = \max_{|\mathbf{x}| \leq 1} |T(\mathbf{x})|, \quad |\mathbf{x}| = \sqrt{x_1^2 + \cdots + x_n^2}.$$

**Theorem:** Given a linear operator  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $t_0 > 0$ , the series

$$\sum_{k=0}^{\infty} \frac{T^k t^k}{k!}$$

is absolutely and uniformly convergent for all  $|t| \leq t_0$ .

**Proof.** Let  $\|T\| = a$ . Then for  $|t| \leq t_0$ ,

$$\left\| \frac{T^k t^k}{k!} \right\| \leq \frac{\|T\|^k |t|^k}{k!} \leq \frac{a^k t_0^k}{k!}.$$

But,  $\sum_{k=0}^{\infty} \frac{a^k t_0^k}{k!} = e^{at_0}$ . By the Weierstrass  $M$ -test, the series  $\sum_{k=0}^{\infty} \frac{T^k t^k}{k!}$  is absolutely and uniformly convergent for all  $|t| \leq t_0$ .

**Definition:** The exponential of the linear operator  $T$  is then defined by

$$e^T = \sum_{k=0}^{\infty} \frac{T^k}{k!}.$$

Assume that  $T$  is represented by the  $n \times n$  matrix  $A$  w.r.t. the standard basis for  $\mathbb{R}^n$  and define the exponential  $e^{At}$  as follows:

**Definition:** Let  $A$  be an  $n \times n$  matrix. Then for  $t \in \mathbb{R}$ ,

$$e^{At} = \sum_{k=0}^{\infty} \frac{A^k t^k}{k!}.$$

$e^{At}$  is an  $n \times n$  matrix which can be computed in terms of the eigenvalues and eigenvectors of  $A$ .

If  $A$  is a diagonal matrix, then the computation of  $e^{At}$  is simple.

**Example:** Let  $A = \begin{bmatrix} -1 & 0 \\ 0 & 2 \end{bmatrix}$ . Then

$$A^2 = \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}, \quad A^3 = \begin{bmatrix} -1 & 0 \\ 0 & 8 \end{bmatrix}, \quad \dots, \quad A^n = \begin{bmatrix} (-1)^n & 0 \\ 0 & 2^n \end{bmatrix}.$$

$$\begin{aligned} e^{At} &= \sum_{k=0}^{\infty} A^k \frac{t^k}{k!} \\ &= \begin{bmatrix} \sum_{k=0}^{\infty} (-1)^k \frac{t^k}{k!} & 0 \\ 0 & \sum_{k=0}^{\infty} (2)^k \frac{t^k}{k!} \end{bmatrix} \\ &= \begin{bmatrix} e^{-t} & 0 \\ 0 & e^{2t} \end{bmatrix}. \end{aligned}$$

**Theorem:** Let  $A$  and  $B$  be  $n \times n$  constant matrices and  $r, s, t \in \mathbb{R}$  (or  $\in \mathbb{C}$ ). Then

- $e^{A0} = e^0 = I$ .
- $e^{A(t+s)} = e^{At} e^{As}$ .
- $(e^{At})^{-1} = e^{-At}$ .
- $e^{(A+B)t} = e^{At} e^{Bt}$ , provided that  $AB = BA$ .
- $e^{rt} = e^{rt} I$ .

**Theorem:** If  $P$  and  $A$  are  $n \times n$  matrices and  $PAP^{-1} = B$ , then

$$e^{Bt} = Pe^{At}P^{-1}.$$

**Proof.** Using the definition of  $e^{At}$ ,

$$\begin{aligned} e^{Bt} &= \lim_{n \rightarrow \infty} \sum_{k=0}^n \frac{(PAP^{-1})^k t^k}{k!} \\ &= P \lim_{n \rightarrow \infty} \sum_{k=0}^n \frac{(At)^k}{k!} P^{-1} = Pe^{At}P^{-1}. \end{aligned}$$

**Corollary:** If  $P^{-1}AP = \text{diag}[\lambda_j]$  then  $e^{At} = P \text{diag}[e^{\lambda_j t}] P^{-1}$ .

**Corollary:** If  $A = \begin{bmatrix} a & -b \\ b & a \end{bmatrix}$  then  $e^A = e^a \begin{bmatrix} \cos b & -\sin b \\ \sin b & \cos b \end{bmatrix}$ .

**Proof.** If  $\lambda = a + ib$ , it follows by induction that

$$\begin{bmatrix} a & -b \\ b & a \end{bmatrix}^k = \begin{bmatrix} \text{Re}(\lambda^k) & -\text{Im}(\lambda^k) \\ \text{Im}(\lambda^k) & \text{Re}(\lambda^k) \end{bmatrix}.$$

Thus,

$$\begin{aligned} e^A &= \sum_{k=0}^{\infty} \begin{bmatrix} \text{Re}(\frac{\lambda^k}{k!}) & -\text{Im}(\frac{\lambda^k}{k!}) \\ \text{Im}(\frac{\lambda^k}{k!}) & \text{Re}(\frac{\lambda^k}{k!}) \end{bmatrix} \\ &= \begin{bmatrix} \text{Re}(e^\lambda) & -\text{Im}(e^\lambda) \\ \text{Im}(e^\lambda) & \text{Re}(e^\lambda) \end{bmatrix} = e^a \begin{bmatrix} \cos b & -\sin b \\ \sin b & \cos b \end{bmatrix}. \end{aligned}$$

If  $a = 0$ , then  $e^A$  is simply a rotation through  $b$  radians.

**Lemma:** Let  $A$  be a square matrix, then

$$\frac{d}{dt}e^{At} = Ae^{At}.$$

**Proof.** We have

$$\begin{aligned}\frac{d}{dt}e^{At} &= \lim_{h \rightarrow 0} \frac{e^{A(t+h)} - e^{At}}{h} \\ &= \lim_{h \rightarrow 0} e^{At} \frac{(e^{Ah} - I)}{h} \\ &= e^{At} \lim_{h \rightarrow 0} \lim_{k \rightarrow \infty} \left( A + \frac{A^2 h}{2!} + \cdots + \frac{A^k h^{k-1}}{k!} \right) = Ae^{At}.\end{aligned}$$

In the above, we have used the fact that the series defining  $e^{Ah}$  converges uniformly for  $|h| \leq 1$ .



Note that

$$\frac{d}{dt}(e^{At}) = Ae^{At}$$

$\implies e^{At}$  is a solution to the matrix differential equation  $\mathbf{x}'(t) = A\mathbf{x}(t)$ .

Since  $e^{At}$  is invertible it follows that the columns of  $e^{At}$  form a fundamental solution set for  $\mathbf{x}'(t) = A\mathbf{x}(t)$ .

**Theorem:** If  $A$  is an  $n \times n$  constant matrix, then the columns of  $e^{At}$  form a fundamental solution set for

$$\mathbf{x}'(t) = A\mathbf{x}(t).$$

Therefore,  $e^{At}$  is a fundamental matrix for the system, and a GS is

$$\mathbf{x}(t) = e^{At}\mathbf{c} = \Phi(t)\mathbf{c}.$$

**Theorem: (The fundamental theorem for linear systems)**

Let  $A$  be an  $n \times n$  matrix. Then for a given  $\mathbf{x}_0 \in \mathbb{R}^n$ , the IVP  $\mathbf{x}' = A\mathbf{x}$ ;  $\mathbf{x}(0) = \mathbf{x}_0$  has a unique solution given by  $\mathbf{x}(t) = e^{At}\mathbf{x}_0$ .

**Proof.** If  $\mathbf{x}(t) = e^{At}\mathbf{x}_0$ , then  $\mathbf{x}'(t) = \frac{d}{dt}e^{At}\mathbf{x}_0 = A\mathbf{x}(t)$ ,  $t \in \mathbb{R}$ . Also,  $\mathbf{x}(0) = I\mathbf{x}_0 = \mathbf{x}_0$ .

**Uniqueness:** Let  $\mathbf{x}(t)$  be any solution of the IVP. Set  $\mathbf{y}(t) = e^{-At}\mathbf{x}(t)$ . Then

$$\begin{aligned} \mathbf{y}'(t) &= -Ae^{-At}\mathbf{x}(t) + e^{-At}\mathbf{x}'(t) \\ &= -Ae^{-At}\mathbf{x}(t) + e^{-At}A\mathbf{x}(t) = \mathbf{0}, \text{ for all } t \in \mathbb{R}. \\ \Rightarrow \mathbf{y}(t) &\text{ is a constant.} \end{aligned}$$

Further,  $\mathbf{y}(0) = \mathbf{x}_0$ . Thus, any solution of the IVP is given by

$$\mathbf{x}(t) = e^{At}\mathbf{y}(t) = e^{At}\mathbf{x}_0.$$

## When $A$ has complex eigenvalues

**Theorem:** Let  $A$  be a real matrix of size  $2n \times 2n$ . If  $A$  has  $2n$  **distinct complex eigenvalues**  $\lambda_j = a_j + ib_j$  and  $\bar{\lambda}_j = a_j - ib_j$  and corresponding complex eigenvectors  $\mathbf{w}_j = \mathbf{u}_j + i\mathbf{v}_j$  and  $\bar{\mathbf{w}}_j = \mathbf{u}_j - i\mathbf{v}_j$ ,  $j = 1, \dots, n$ , then  $\{\mathbf{u}_1, \mathbf{v}_1, \dots, \mathbf{u}_n, \mathbf{v}_n\}$  is a basis for  $\mathbb{R}^{2n}$ , the matrix

$$P = [\mathbf{v}_1 \ \mathbf{u}_1 \ \mathbf{v}_2 \ \mathbf{u}_2 \ \cdots \ \mathbf{v}_n \ \mathbf{u}_n]$$

is invertible and

$$P^{-1}AP = \text{diag} \begin{bmatrix} a_j & -b_j \\ b_j & a_j \end{bmatrix}$$

a real  $2n \times 2n$  matrix with  $2 \times 2$  blocks along the diagonal.

**Remark:** Instead of  $P$  if we use

$$Q = [\mathbf{u}_1 \ \mathbf{v}_1 \ \mathbf{u}_2 \ \mathbf{v}_2 \ \cdots \ \mathbf{u}_n \ \mathbf{v}_n]$$

then

$$Q^{-1}AQ = \text{diag} \begin{bmatrix} a_j & b_j \\ -b_j & a_j \end{bmatrix}.$$

Using the above result, the solution of the IVP  $\mathbf{x}' = A\mathbf{x}$ ,  $\mathbf{x}(0) = \mathbf{x}_0$  is given by

$$\mathbf{x}(t) = P \text{diag} e^{a_j t} \begin{bmatrix} \cos(b_j t) & -\sin(b_j t) \\ \sin(b_j t) & \cos(b_j t) \end{bmatrix} P^{-1} \mathbf{x}_0$$

**Example:** Solve the IVP  $\mathbf{x}' = A\mathbf{x}$ ,  $\mathbf{x}(0) = \mathbf{x}_0$  with

$$A = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 3 & -2 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

$A$  has complex eigenvalues  $\lambda_1 = 1 + i$   $\lambda_2 = 2 + i$  (as well as  $\bar{\lambda}_1 = 1 - i$   $\bar{\lambda}_2 = 2 - i$ ). A corresponding pair of complex eigenvectors is

$$\mathbf{w}_1 = \mathbf{u}_1 + i\mathbf{v}_1 = [i \ 1 \ 0 \ 0]^T \quad \text{and} \quad \mathbf{w}_2 = \mathbf{u}_2 + i\mathbf{v}_2 = [0 \ 0 \ 1 + i \ 1]^T.$$

The matrix

$$P = [\mathbf{v}_1 \ \mathbf{u}_1 \ \mathbf{v}_2 \ \mathbf{u}_2] = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$P^{-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \text{ and } P^{-1}AP = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 2 & -1 \\ 0 & 0 & 1 & 2 \end{bmatrix}$$

The solution to IVP is

$$\begin{aligned} \mathbf{x}(t) &= P \begin{bmatrix} e^t \cos t & -e^t \sin t & 0 & 0 \\ e^t \sin t & e^t \cos t & 0 & 0 \\ 0 & 0 & e^{2t} \cos t & -e^{2t} \sin t \\ 0 & 0 & e^{2t} \sin t & e^{2t} \cos t \end{bmatrix} P^{-1} \mathbf{x}_0 \\ &= \begin{bmatrix} e^t \cos t & -e^t \sin t & 0 & 0 \\ e^t \sin t & e^t \cos t & 0 & 0 \\ 0 & 0 & e^{2t}(\cos t + \sin t) & -e^{2t} \sin t \\ 0 & 0 & e^{2t} \sin t & e^{2t}(\cos t - \sin t) \end{bmatrix} \mathbf{x}_0 \end{aligned}$$

## When $A$ has both real and complex eigenvalues.

**Theorem:** If  $A$  has distinct real eigenvalues  $\lambda_j$  and corresponding eigenvectors  $\mathbf{v}_j$ ,  $j = 1, \dots, k$  and distinct complex eigenvalues  $\lambda_j = a_j + ib_j$  and  $\bar{\lambda}_j = a_j - ib_j$  and corresponding eigenvectors  $\mathbf{w}_j = \mathbf{u}_j + i\mathbf{v}_j$  and  $\bar{\mathbf{w}}_j = \mathbf{u}_j - i\mathbf{v}_j$ ,  $j = k + 1, \dots, n$ , then the matrix

$$P = [\mathbf{v}_1 \cdots \mathbf{v}_k \mathbf{v}_{k+1} \mathbf{u}_{k+1} \cdots \mathbf{v}_n \mathbf{u}_n]$$

is invertible and

$$P^{-1}AP = \text{diag}[\lambda_1, \dots, \lambda_k, B_{k+1}, \dots, B_n],$$

where  $B_j = \begin{bmatrix} a_j & -b_j \\ b_j & a_j \end{bmatrix}$  for  $j = k + 1, \dots, n$ .

**Example:** Solve the IVP  $\mathbf{x}' = A\mathbf{x}$ ,  $\mathbf{x}(0) = \mathbf{x}_0$  with

$$A = \begin{bmatrix} -3 & 0 & 0 \\ 0 & 3 & -2 \\ 0 & 1 & 1 \end{bmatrix}.$$

The eigenvalues are  $\lambda_1 = -3$ ,  $\lambda_2 = 2 + i$  ( $\bar{\lambda}_2 = 2 - i$ ).

The corresponding eigenvectors

$$\mathbf{v}_1 = [1 \ 0 \ 0]^T \text{ and } \mathbf{w}_2 = \mathbf{u}_2 + i\mathbf{v}_2 = [0 \ 1 + i \ 1]^T$$

$$P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}, \quad P^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix}.$$



$$P^{-1}AP = \begin{bmatrix} -3 & 0 & 0 \\ 0 & 2 & -1 \\ 0 & 1 & 2 \end{bmatrix}$$

The solution of IVP is

$$\begin{aligned} \mathbf{x}(t) &= P \begin{bmatrix} e^{-3t} & 0 & 0 \\ 0 & e^{2t} \cos t & -e^{2t} \sin t \\ 0 & e^{2t} \sin t & e^{2t} \cos t \end{bmatrix} P^{-1} \mathbf{x}_0 \\ &= \begin{bmatrix} e^{-3t} & 0 & 0 \\ 0 & e^{2t}(\cos t + \sin t) & -2e^{2t} \sin t \\ 0 & e^{2t} \sin t & e^{2t}(\cos t - \sin t) \end{bmatrix} \mathbf{x}_0. \end{aligned}$$

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