

## First Order Linear Homogeneous Systems with Constants Coefficients

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Find  $\mathbf{v} \neq \mathbf{0}$  such that  $e^{\mathbf{A}t}\mathbf{v}$  can be “*calculated easily*”.

- If  $\mathbf{A}\mathbf{v} = \mathbf{0}$ , then  $e^{\mathbf{A}t}\mathbf{v} = \mathbf{v}$
- If not, consider  $(\mathbf{A} - \lambda\mathbf{I})\mathbf{v}$

## Eigenvalue Method

**Definition** Let  $\mathbf{A}$  be a square matrix of order  $n$  (entries can be complex)

A number  $\lambda$  (can be complex) is said to be an *eigenvalue* of  $\mathbf{A}$  if there exists a non-zero vector  $\mathbf{v} \in \mathbb{C}^n$  such that

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*Reason*  $\lambda$  is an eigenvalue of  $\mathbf{A}$  iff  $(\mathbf{A} - \lambda\mathbf{I})\mathbf{v} = \mathbf{0}$  has non-trivial solution.

$$e^{At} = e^{(A-rI)t+rtI}$$

$$\begin{aligned} e^{\mathbf{A}t} &= e^{(\mathbf{A}-r\mathbf{I})t+rt\mathbf{I}} \\ &= e^{(\mathbf{A}-r\mathbf{I})t} \cdot e^{rt\mathbf{I}} \end{aligned}$$

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**Linear Algebra Result** If  $r \in \mathbb{R}$  is an eigenvalue of  $\mathbf{A}$  where  $\mathbf{A} \in \mathcal{M}_n$ , then there is an associated eigenvector  $\mathbf{v} \in \mathbb{R}^n$

## Theorem

$\mathbf{x}(t) = \mathbf{v}e^{rt}$  is non-trivial solution to  $\mathbf{x}' = \mathbf{A}\mathbf{x} \iff r \in \mathbb{R}$  is an eigenvalue of  $\mathbf{A}$  and  $\mathbf{v} \in \mathbb{R}^n$  an associated eigenvector

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( $\Leftarrow$ ) Done.  $\mathbf{x}(t) = \mathbf{v}e^{rt}$  is an  $\mathbb{R}^n$ -valued function.

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Non-trivial means  $\mathbf{v} \neq \mathbf{0}$

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General solution  $\mathbf{x}(t) = c_1\mathbf{v}^{(1)}e^{r_1t} + \dots + c_n\mathbf{v}^{(n)}e^{r_nt}$

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General solution  $\mathbf{x}(t) = c_1 \begin{bmatrix} -\cos 3t + 3 \sin 3t \\ 5 \cos 3t \end{bmatrix} + c_2 \begin{bmatrix} -3 \cos 3t - \sin 3t \\ 5 \sin 3t \end{bmatrix}$



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**Fact**  $G.M. \leq A.M.$

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$$e^{\mathbf{A}t}\mathbf{u} = e^{rt} \left( \mathbf{I} + t(\mathbf{A} - r\mathbf{I}) + \frac{t^2}{2}(\mathbf{A} - r\mathbf{I})^2 + \frac{t^3}{3!}(\mathbf{A} - r\mathbf{I})^3 + \dots \right) \mathbf{u}$$

- ◇ If  $(\mathbf{A} - r\mathbf{I})\mathbf{u} = \mathbf{0}$ , then  $\mathbf{u}$  is an eigenvector. ( $\mathbf{u}$  is a multiple of  $\mathbf{v}$ , no use)
- ◇ If  $(\mathbf{A} - r\mathbf{I})\mathbf{u} \neq \mathbf{0}$  but  $(\mathbf{A} - r\mathbf{I})^2\mathbf{u} = \mathbf{0}$ , then
  - $(\mathbf{A} - r\mathbf{I})\mathbf{u}$  is an eigenvector.  $\because (\mathbf{A} - r\mathbf{I})(\mathbf{A} - r\mathbf{I})\mathbf{u} = \mathbf{0}$
  - up to scalar multiples, may assume

$$(\mathbf{A} - r\mathbf{I})\mathbf{u} = \mathbf{v}$$

**Linear algebra result** Above equation has solution  $\mathbf{u} \in \mathbb{R}^n$



$$\begin{aligned} e^{\mathbf{A}t}\mathbf{u} &= e^{rt} \left( \mathbf{I} + t(\mathbf{A} - r\mathbf{I}) + \frac{t^2}{2}(\mathbf{A} - r\mathbf{I})^2 + \frac{t^3}{3!}(\mathbf{A} - r\mathbf{I})^3 + \dots \right) \mathbf{u} \\ &= e^{rt}(\mathbf{u} + t\mathbf{v}) \end{aligned}$$

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Second solution  $\mathbf{x}^{(2)}(t) = \mathbf{v}te^{rt} + \mathbf{u}e^{rt}$

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### Summary for Subcase (2,1)

- Solve for  $\mathbf{v}$  and  $\mathbf{u}$ 

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- Two independent solutions
 
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**Definition** Let  $\lambda$  be an **eigenvalue** of a square matrix  $\mathbf{A}$ .

Suppose  $\mathbf{u}$  is a *non-zero vector* such that

$$(\mathbf{A} - \lambda\mathbf{I})^k \mathbf{u} = \mathbf{0}$$

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**Remark** Eigenvector: special case where  $k = 1$

**Example** Find two linearly independent solutions to  $\mathbf{x}' = \begin{bmatrix} -6 & 5 \\ -5 & 4 \end{bmatrix} \mathbf{x}$

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Second linearly independent solution  $\mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-t} + \begin{bmatrix} 0 \\ \frac{1}{5} \end{bmatrix} e^{-t}$