

Lecture 11

- 5.5 Fundamental Matrices
- 5.6 Repeated Eigenvalues

5.1 - 5.4 Recap

$$\left\{ \begin{array}{l} x'_1 = p_{11}(t)x_1 + p_{12}(t)x_2 + \dots + p_{1n}(t)x_n + g_1(t) \\ x'_2 = p_{21}(t)x_1 + p_{22}(t)x_2 + \dots + p_{2n}(t)x_n + g_2(t) \\ \vdots \\ x'_n = p_{n1}(t)x_1 + p_{n2}(t)x_2 + \dots + p_{nn}(t)x_n + g_n(t) \end{array} \right.$$

is a system of n linear ODEs and n variables: x_1, x_2, \dots, x_n .

5.1 - 5.4 Recap

$$\begin{cases} x'_1 = p_{11}(t)x_1 + p_{12}(t)x_2 + \dots + p_{1n}(t)x_n + g_1(t) \\ x'_2 = p_{21}(t)x_1 + p_{22}(t)x_2 + \dots + p_{2n}(t)x_n + g_2(t) \\ \vdots \\ x'_n = p_{n1}(t)x_1 + p_{n2}(t)x_2 + \dots + p_{nn}(t)x_n + g_n(t) \end{cases}$$

is a system of n linear ODEs and n variables: x_1, x_2, \dots, x_n .
 If we write

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \quad \mathbf{x}' = \begin{bmatrix} x'_1 \\ x'_2 \\ \vdots \\ x'_n \end{bmatrix}, \quad P = \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1n} \\ p_{21} & p_{22} & \dots & p_{2n} \\ \vdots & \vdots & & \vdots \\ p_{n1} & p_{n2} & \dots & p_{nn} \end{bmatrix}, \text{ and } \mathbf{g} = \begin{bmatrix} g_1 \\ g_2 \\ \vdots \\ g_n \end{bmatrix}$$

5.1 - 5.4 Recap

$$\left\{ \begin{array}{l} x'_1 = p_{11}(t)x_1 + p_{12}(t)x_2 + \dots + p_{1n}(t)x_n + g_1(t) \\ x'_2 = p_{21}(t)x_1 + p_{22}(t)x_2 + \dots + p_{2n}(t)x_n + g_2(t) \\ \vdots \\ x'_n = p_{n1}(t)x_1 + p_{n2}(t)x_2 + \dots + p_{nn}(t)x_n + g_n(t) \end{array} \right.$$

is a system of n linear ODEs and n variables: x_1, x_2, \dots, x_n .

If we write

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \quad \mathbf{x}' = \begin{bmatrix} x'_1 \\ x'_2 \\ \vdots \\ x'_n \end{bmatrix}, \quad P = \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1n} \\ p_{21} & p_{22} & \dots & p_{2n} \\ \vdots & \vdots & & \vdots \\ p_{n1} & p_{n2} & \dots & p_{nn} \end{bmatrix}, \text{ and } \mathbf{g} = \begin{bmatrix} g_1 \\ g_2 \\ \vdots \\ g_n \end{bmatrix}$$

then we can write this system as

$$\mathbf{x}' = P(t)\mathbf{x} + \mathbf{g}(t).$$

5.1 - 5.4 Recap



In Chapter 3 when we had multiple solutions, we wrote them as $y_1(t)$, $y_2(t)$, But we are already using x_1 , x_2 , . . . to denote coordinates. So we need a new type of notation.

5.1 - 5.4 Recap



In Chapter 3 when we had multiple solutions, we wrote them as $y_1(t)$, $y_2(t)$, \dots . But we are already using x_1 , x_2 , \dots to denote coordinates. So we need a new type of notation.

Notation

We use $\mathbf{x}^{(1)}(t)$, $\mathbf{x}^{(2)}(t)$, \dots to denote different vector solutions.

5.1 - 5.4 Recap



Recall from Chapter 3 that if $y_1(t)$ and $y_2(t)$ are both solutions to

$$ay'' + by' + cy = 0,$$

then

$$c_1y_1 + c_2y_2$$

is also a solution.

5.1 - 5.4 Recap



Theorem

If $\mathbf{x}^{(1)}(t)$ and $\mathbf{x}^{(2)}(t)$ are solutions to $\mathbf{x}' = P(t)\mathbf{x}$, then

$$c_1\mathbf{x}^{(1)} + c_2\mathbf{x}^{(2)}$$

is also a solution for any $c_1, c_2 \in \mathbb{R}$.

5.1 - 5.4 Recap



(Suppose that $P(t)$ is an $n \times n$ matrix.)

Theorem

If $\mathbf{x}^{(1)}(t), \mathbf{x}^{(2)}(t), \dots, \mathbf{x}^{(n)}(t)$ are linearly independent solutions to $\mathbf{x}' = P(t)\mathbf{x}$, then every solution to this system can be written as

$$\mathbf{x}(t) = c_1\mathbf{x}^{(1)} + c_2\mathbf{x}^{(2)} + \dots + c_n\mathbf{x}^{(n)}$$

in exactly one way.

5.1 - 5.4 Recap



Definition

In this case, we say that $\mathbf{x}^{(1)}(t), \mathbf{x}^{(2)}(t), \dots, \mathbf{x}^{(n)}(t)$ form a *fundamental set of solutions* to $\mathbf{x}' = P(t)\mathbf{x}$.

5.1 - 5.4 Recap

Definition

In this case, we say that $\mathbf{x}^{(1)}(t), \mathbf{x}^{(2)}(t), \dots, \mathbf{x}^{(n)}(t)$ form a *fundamental set of solutions* to $\mathbf{x}' = P(t)\mathbf{x}$.

Definition

In this case,

$$\mathbf{x}(t) = c_1\mathbf{x}^{(1)} + c_2\mathbf{x}^{(2)} + \dots + c_n\mathbf{x}^{(n)}$$

is called the *general solution* to $\mathbf{x}' = P(t)\mathbf{x}$.

5.1 - 5.4 Recap



Consider

$$\mathbf{x}' = A\mathbf{x}$$

where $A \in \mathbb{R}^{n \times n}$.

Remark

The idea is:

- 1 Find the eigenvalues;
- 2 Find the eigenvectors; then
- 3 Write $\mathbf{x}^{(j)}(t) = \xi^{(j)} e^{r_j t}$.

5.1 - 5.4 Recap

Example

Solve

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

The eigenvalues are $r_1 = 3$ and $r_2 = -1$.

The eigenvectors are $\xi^{(1)} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $\xi^{(2)} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$.

This gives us two solutions:

$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ 2 \end{bmatrix} e^{3t} \quad \text{and} \quad \mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ -2 \end{bmatrix} e^{-t}.$$

5.1 - 5.4 Recap

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Therefore the general solution is

$$\boxed{\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} e^{3t} + c_2 \begin{bmatrix} 1 \\ -2 \end{bmatrix} e^{-t}.}$$

5.1 - 5.4 Recap

Example

Solve

$$\begin{cases} \mathbf{x}' = \begin{bmatrix} 8 & -1 \\ 6 & 1 \end{bmatrix} \mathbf{x} \\ \mathbf{x}(0) = \begin{bmatrix} 1 \\ -2 \end{bmatrix}. \end{cases}$$

5.1 - 5.4 Recap

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$$\begin{cases} \mathbf{x}' = \begin{bmatrix} 8 & -1 \\ 6 & 1 \end{bmatrix} \mathbf{x} \\ \mathbf{x}(0) = \begin{bmatrix} 1 \\ -2 \end{bmatrix}. \end{cases}$$

The eigenvalues are $r_1 = 7$ and $r_2 = 2$.

5.1 - 5.4 Recap

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The eigenvalues are $r_1 = 7$ and $r_2 = 2$. The corresponding eigenvectors are $\xi^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and $\xi^{(2)} = \begin{bmatrix} 1 \\ 6 \end{bmatrix}$.

5.1 - 5.4 Recap



Example

Solve

$$\begin{cases} \mathbf{x}' = \begin{bmatrix} 8 & -1 \\ 6 & 1 \end{bmatrix} \mathbf{x} \\ \mathbf{x}(0) = \begin{bmatrix} 1 \\ -2 \end{bmatrix}. \end{cases}$$

The eigenvalues are $r_1 = 7$ and $r_2 = 2$. The corresponding eigenvectors are $\xi^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and $\xi^{(2)} = \begin{bmatrix} 1 \\ 6 \end{bmatrix}$. Therefore the general solution to the ODE is

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.$$

5.1 - 5.4 Recap



$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.$$

Setting $t = 0$, we have

$$\begin{bmatrix} 1 \\ -2 \end{bmatrix} = \mathbf{x}(0)$$

5.1 - 5.4 Recap

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.$$

Setting $t = 0$, we have

$$\begin{bmatrix} 1 \\ -2 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} = \begin{bmatrix} c_1 + c_2 \\ c_1 + 6c_2 \end{bmatrix}$$

5.1 - 5.4 Recap

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.$$

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5.1 - 5.4 Recap



$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.$$

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$$\begin{bmatrix} 1 \\ -2 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} = \begin{bmatrix} c_1 + c_2 \\ c_1 + 6c_2 \end{bmatrix} \implies \begin{cases} c_1 + c_2 = 1 \\ c_1 + 6c_2 = -2 \end{cases}$$
$$\implies \begin{cases} c_1 = \frac{8}{5} \\ c_2 = -\frac{3}{5}. \end{cases}$$

5.1 - 5.4 Recap



$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.$$

Setting $t = 0$, we have

$$\begin{bmatrix} 1 \\ -2 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} = \begin{bmatrix} c_1 + c_2 \\ c_1 + 6c_2 \end{bmatrix} \implies \begin{cases} c_1 + c_2 = 1 \\ c_1 + 6c_2 = -2 \end{cases}$$
$$\implies \begin{cases} c_1 = \frac{8}{5} \\ c_2 = -\frac{3}{5}. \end{cases}$$

Therefore the solution to the IVP is

$$\boxed{\mathbf{x}(t) = \frac{8}{5} \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} - \frac{3}{5} \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.}$$

5.1 - 5.4 Recap



Remark

$$\det(A - rI) = 0$$

There are three possibilities for the eigenvalues of A .

- 1 All the eigenvalues are real and different;
- 2 Some eigenvalues occur in complex conjugate pairs;
- 3 Some eigenvalues are repeated.

5.1 - 5.4 Recap



If all the eigenvalues are real and different, then the eigenvectors are linearly independent:

5.1 - 5.4 Recap



If all the eigenvalues are real and different, then the eigenvectors are linearly independent:

So $W(\mathbf{x}^{(1)}, \dots, \mathbf{x}^{(n)})(t) \neq 0$ and $\mathbf{x}^{(1)}(t), \dots, \mathbf{x}^{(n)}(t)$ form a fundamental set of solutions.

If some eigenvalues are repeated, *but there are n linearly independent eigenvectors*, then this is also true: $\mathbf{x}^{(1)}(t), \dots, \mathbf{x}^{(n)}(t)$ form a fundamental set of solutions.

5.1 - 5.4 Recap



Example

Solve

$$\mathbf{x}' = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix} \mathbf{x}.$$

5.1 - 5.4 Recap

The eigenvalues and eigenvectors are

$$r_1 = 2$$

$$r_2 = -1$$

$$r_3 = -1$$

$$\xi^{(1)} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\xi^{(2)} = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

$$\xi^{(3)} = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$$

5.1 - 5.4 Recap



The eigenvalues and eigenvectors are

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$$r_3 = -1$$

$$\xi^{(1)} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\xi^{(2)} = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

$$\xi^{(3)} = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$$

which gives us the following three solutions

$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} e^{2t} \quad \mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} e^{-t} \quad \mathbf{x}^{(3)}(t) = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} e^{-t}.$$

5.1 - 5.4 Recap



$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} e^{2t} \quad \mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} e^{-t} \quad \mathbf{x}^{(3)}(t) = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} e^{-t}.$$

You can check that the Wronskian of $\mathbf{x}^{(1)}$, $\mathbf{x}^{(2)}$ and $\mathbf{x}^{(3)}$ is non-zero. Therefore $\mathbf{x}^{(1)}$, $\mathbf{x}^{(2)}$ and $\mathbf{x}^{(3)}$ form a fundamental set of solutions.

5.1 - 5.4 Recap



$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} e^{2t} \quad \mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} e^{-t} \quad \mathbf{x}^{(3)}(t) = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} e^{-t}.$$

You can check that the Wronskian of $\mathbf{x}^{(1)}$, $\mathbf{x}^{(2)}$ and $\mathbf{x}^{(3)}$ is non-zero. Therefore $\mathbf{x}^{(1)}$, $\mathbf{x}^{(2)}$ and $\mathbf{x}^{(3)}$ form a fundamental set of solutions. The general solution to the ODE is

$$\boxed{\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} e^{2t} + c_2 \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} e^{-t} + c_3 \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} e^{-t}.}$$

5.1 - 5.4 Recap



Remark

So if we have repeated eigenvalues with n linearly independent eigenvectors then there is no problem.

Later today we will study what to do if we do not have enough eigenvectors.

5.1 - 5.4 Recap



Consider

$$\mathbf{x}' = A\mathbf{x}$$

where $A \in \mathbb{R}^{n \times n}$.

$$\begin{bmatrix} x'_1 \\ x'_2 \\ \vdots \\ x'_n \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

5.1 - 5.4 Recap

Consider

$$\mathbf{x}' = A\mathbf{x}$$

where $A \in \mathbb{R}^{n \times n}$.

$$\begin{cases} x'_1 = a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ x'_2 = a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \\ x'_n = a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n \end{cases}$$

5.1 - 5.4 Recap



Now let's talk about complex eigenvalues.

Any complex eigenvalues of A must occur in complex conjugate pairs: If $r_1 = \lambda + i\mu$ is an eigenvalue of A , then $r_2 = \bar{r}_1 = \lambda - i\mu$ is also an eigenvalue of A .

Moreover, if $\xi^{(1)}$ is an eigenvector of A corresponding to r_1 , then $\xi^{(2)} = \overline{\xi^{(1)}}$ is an eigenvector of A corresponding to $r_2 = \bar{r}_1$.

5.1 - 5.4 Recap

Two solutions of $\mathbf{x}' = A\mathbf{x}$ are

$$\mathbf{x}^{(1)}(t) = \boldsymbol{\xi}^{(1)} e^{r_1 t} \quad \text{and} \quad \mathbf{x}^{(2)}(t) = \overline{\boldsymbol{\xi}^{(1)}} e^{\bar{r}_1 t}.$$

5.1 - 5.4 Recap

Two solutions of $\mathbf{x}' = A\mathbf{x}$ are

$$\mathbf{x}^{(1)}(t) = \boldsymbol{\xi}^{(1)} e^{r_1 t} \quad \text{and} \quad \mathbf{x}^{(2)}(t) = \overline{\boldsymbol{\xi}^{(1)}} e^{\bar{r}_1 t}.$$

But $\mathbf{x}^{(1)}, \mathbf{x}^{(2)} : \mathbb{R} \rightarrow \mathbb{C}^n$ and we want solutions : $\mathbb{R} \rightarrow \mathbb{R}^n$.

5.1 - 5.4 Recap



If $r_1 = \lambda + i\mu$, and $\xi^{(1)} = \mathbf{a} + i\mathbf{b}$ ($\lambda, \mu \in \mathbb{R}$, $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$), then

$$\mathbf{x}^{(1)}(t) = (\mathbf{a} + i\mathbf{b})e^{(\lambda+i\mu)t}$$

5.1 - 5.4 Recap



If $r_1 = \lambda + i\mu$, and $\xi^{(1)} = \mathbf{a} + i\mathbf{b}$ ($\lambda, \mu \in \mathbb{R}$, $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$), then

$$\begin{aligned}\mathbf{x}^{(1)}(t) &= (\mathbf{a} + i\mathbf{b})e^{(\lambda+i\mu)t} \\ &= (\mathbf{a} + i\mathbf{b})e^{\lambda t} (\cos \mu t + i \sin \mu t)\end{aligned}$$

5.1 - 5.4 Recap



If $r_1 = \lambda + i\mu$, and $\xi^{(1)} = \mathbf{a} + i\mathbf{b}$ ($\lambda, \mu \in \mathbb{R}$, $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$), then

$$\begin{aligned}\mathbf{x}^{(1)}(t) &= (\mathbf{a} + i\mathbf{b})e^{(\lambda+i\mu)t} \\ &= (\mathbf{a} + i\mathbf{b})e^{\lambda t} (\cos \mu t + i \sin \mu t) \\ &= e^{\lambda t} (\mathbf{a} \cos \mu t - \mathbf{b} \sin \mu t) + i e^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t)\end{aligned}$$

5.1 - 5.4 Recap



If $r_1 = \lambda + i\mu$, and $\boldsymbol{\xi}^{(1)} = \mathbf{a} + i\mathbf{b}$ ($\lambda, \mu \in \mathbb{R}$, $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$), then

$$\begin{aligned}\mathbf{x}^{(1)}(t) &= (\mathbf{a} + i\mathbf{b})e^{(\lambda+i\mu)t} \\ &= (\mathbf{a} + i\mathbf{b})e^{\lambda t} (\cos \mu t + i \sin \mu t) \\ &= e^{\lambda t} (\mathbf{a} \cos \mu t - \mathbf{b} \sin \mu t) + i e^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t) \\ &= \mathbf{u}(t) + i\mathbf{v}(t).\end{aligned}$$

5.1 - 5.4 Recap



Remark

- The functions $\mathbf{u}(t)$ and $\mathbf{v}(t)$ solve the ODE.

5.1 - 5.4 Recap



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- The functions $\mathbf{u}(t)$ and $\mathbf{v}(t)$ will be linearly independent.

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Remark

- The functions $\mathbf{u}(t)$ and $\mathbf{v}(t)$ solve the ODE.
- The functions $\mathbf{u}(t)$ and $\mathbf{v}(t)$ will be linearly independent.
- $\text{span}\{\mathbf{u}(t), \mathbf{v}(t)\} = \text{span}\{\mathbf{x}^{(1)}, \mathbf{x}^{(2)}\}$.

So we can include $\mathbf{u}(t)$ and $\mathbf{v}(t)$ in our fundamental set of solutions instead of $\mathbf{x}^{(1)}(t)$ and $\mathbf{x}^{(2)}(t)$.

5.1 - 5.4 Recap

Example

Solve

$$\mathbf{x}' = \begin{bmatrix} -\frac{1}{2} & 1 \\ -1 & -\frac{1}{2} \end{bmatrix} \mathbf{x}.$$

5.1 - 5.4 Recap

Example

Solve

$$\mathbf{x}' = \begin{bmatrix} -\frac{1}{2} & 1 \\ -1 & -\frac{1}{2} \end{bmatrix} \mathbf{x}.$$

We calculate that

$$0 = \det(A - rI) = \begin{vmatrix} -\frac{1}{2} - r & 1 \\ -1 & -\frac{1}{2} - r \end{vmatrix} = r^2 + r + \frac{5}{4}$$

and

$$r_{1,2} = \frac{-1 \pm \sqrt{1-5}}{2} = \frac{-1 \pm 2i}{2} = -\frac{1}{2} \pm i.$$

5.1 - 5.4 Recap

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$$r_{1,2} = \frac{-1 \pm \sqrt{1-5}}{2} = \frac{-1 \pm 2i}{2} = -\frac{1}{2} \pm i.$$

So we have $r_1 = -\frac{1}{2} + i$ and $r_2 = -\frac{1}{2} - i$.

5.1 - 5.4 Recap

Example

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We calculate that

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and

$$r_{1,2} = \frac{-1 \pm \sqrt{1-5}}{2} = \frac{-1 \pm 2i}{2} = -\frac{1}{2} \pm i.$$

So we have $r_1 = -\frac{1}{2} + i$ and $r_2 = -\frac{1}{2} - i$. We will use r_1 . We do not need r_2 .

5.1 - 5.4 Recap

Since

$$0 = (A - r_1 I) \boldsymbol{\xi}^{(1)} = \begin{bmatrix} -i & 1 \\ -1 & -i \end{bmatrix} \begin{bmatrix} \xi_1 \\ \xi_2 \end{bmatrix}$$

5.1 - 5.4 Recap

Since

$$0 = (A - r_1 I) \boldsymbol{\xi}^{(1)} = \begin{bmatrix} -i & 1 \\ -1 & -i \end{bmatrix} \begin{bmatrix} \xi_1 \\ \xi_2 \end{bmatrix} \implies \begin{cases} -i\xi_1 + \xi_2 = 0 \\ -\xi_1 - i\xi_2 = 0 \end{cases}$$

we can choose

$$\boldsymbol{\xi}^{(1)} = \begin{bmatrix} 1 \\ i \end{bmatrix}.$$

5.1 - 5.4 Recap



Since

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we can choose

$$\boldsymbol{\xi}^{(1)} = \begin{bmatrix} 1 \\ i \end{bmatrix}.$$

Note that we also have

$$\boldsymbol{\xi}^{(2)} = \overline{\boldsymbol{\xi}^{(1)}} = \begin{bmatrix} \overline{1} \\ \overline{i} \end{bmatrix} = \begin{bmatrix} 1 \\ -i \end{bmatrix},$$

but we don't need $\boldsymbol{\xi}^{(2)}$.

5.1 - 5.4 Recap



Next we look at $\mathbf{x}^{(1)}(t)$:

5.1 - 5.4 Recap



Next we look at $\mathbf{x}^{(1)}(t)$:

$$\mathbf{x}^{(1)}(t) = \boldsymbol{\xi}^{(1)} e^{r_1 t} = \begin{bmatrix} 1 \\ i \end{bmatrix} e^{-\frac{t}{2}} (\cos t + i \sin t)$$

=

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5.1 - 5.4 Recap



Next we look at $\mathbf{x}^{(1)}(t)$:

$$\begin{aligned}\mathbf{x}^{(1)}(t) &= \boldsymbol{\xi}^{(1)} e^{r_1 t} = \begin{bmatrix} 1 \\ i \end{bmatrix} e^{-\frac{t}{2}} (\cos t + i \sin t) \\ &= e^{-\frac{t}{2}} \begin{bmatrix} \cos t + i \sin t \\ i \cos t - \sin t \end{bmatrix}\end{aligned}$$

=

5.1 - 5.4 Recap

Next we look at $\mathbf{x}^{(1)}(t)$:

$$\begin{aligned}
 \mathbf{x}^{(1)}(t) &= \boldsymbol{\xi}^{(1)} e^{r_1 t} = \begin{bmatrix} 1 \\ i \end{bmatrix} e^{-\frac{t}{2}} (\cos t + i \sin t) \\
 &= e^{-\frac{t}{2}} \begin{bmatrix} \cos t + i \sin t \\ i \cos t - \sin t \end{bmatrix} \\
 &= e^{-\frac{t}{2}} \begin{bmatrix} \cos t \\ -\sin t \end{bmatrix} + i e^{-\frac{t}{2}} \begin{bmatrix} \sin t \\ \cos t \end{bmatrix}
 \end{aligned}$$

5.1 - 5.4 Recap



Next we look at $\mathbf{x}^{(1)}(t)$:

$$\begin{aligned}\mathbf{x}^{(1)}(t) &= \boldsymbol{\xi}^{(1)} e^{r_1 t} = \begin{bmatrix} 1 \\ i \end{bmatrix} e^{-\frac{t}{2}} (\cos t + i \sin t) \\ &= e^{-\frac{t}{2}} \begin{bmatrix} \cos t + i \sin t \\ i \cos t - \sin t \end{bmatrix} \\ &= e^{-\frac{t}{2}} \begin{bmatrix} \cos t \\ -\sin t \end{bmatrix} + i e^{-\frac{t}{2}} \begin{bmatrix} \sin t \\ \cos t \end{bmatrix} \\ &= \mathbf{u}(t) + i\mathbf{v}(t).\end{aligned}$$

5.1 - 5.4 Recap

Next we look at $\mathbf{x}^{(1)}(t)$:

$$\begin{aligned}
 \mathbf{x}^{(1)}(t) &= \boldsymbol{\xi}^{(1)} e^{r_1 t} = \begin{bmatrix} 1 \\ i \end{bmatrix} e^{-\frac{t}{2}} (\cos t + i \sin t) \\
 &= e^{-\frac{t}{2}} \begin{bmatrix} \cos t + i \sin t \\ i \cos t - \sin t \end{bmatrix} \\
 &= e^{-\frac{t}{2}} \begin{bmatrix} \cos t \\ -\sin t \end{bmatrix} + i e^{-\frac{t}{2}} \begin{bmatrix} \sin t \\ \cos t \end{bmatrix} \\
 &= \mathbf{u}(t) + i \mathbf{v}(t).
 \end{aligned}$$

Hence we choose

$$\mathbf{u}(t) = e^{-\frac{t}{2}} \begin{bmatrix} \cos t \\ -\sin t \end{bmatrix} \quad \text{and} \quad \mathbf{v}(t) = e^{-\frac{t}{2}} \begin{bmatrix} \sin t \\ \cos t \end{bmatrix}.$$

5.1 - 5.4 Recap



$$\mathbf{u}(t) = e^{-\frac{t}{2}} \begin{bmatrix} \cos t \\ -\sin t \end{bmatrix} \quad \text{and} \quad \mathbf{v}(t) = e^{-\frac{t}{2}} \begin{bmatrix} \sin t \\ \cos t \end{bmatrix}.$$

Therefore the general solution to $\mathbf{x}' = \begin{bmatrix} -\frac{1}{2} & 1 \\ -1 & -\frac{1}{2} \end{bmatrix} \mathbf{x}$ is

$$\boxed{\mathbf{x}(t) = c_1 e^{-\frac{t}{2}} \begin{bmatrix} \cos t \\ -\sin t \end{bmatrix} + c_2 e^{-\frac{t}{2}} \begin{bmatrix} \sin t \\ \cos t \end{bmatrix}}.$$

5.1 - 5.4 Recap

Remark

Our method is

1. Find the eigenvalues;
2. Find the eigenvectors;
3.
 - If r_j is real, just use the solution $\mathbf{x}^{(j)}(t) = \boldsymbol{\xi}^{(j)} e^{r_j t}$;
 - But if r_j is complex, write

$$\mathbf{x}^{(j)}(t) = \boldsymbol{\xi}^{(j)} e^{r_j t} = \begin{pmatrix} \text{real valued} \\ \text{function} \end{pmatrix} + i \begin{pmatrix} \text{real valued} \\ \text{function} \end{pmatrix}$$

and use these two functions.

5.1 - 5.4 Recap



There are lots more details and examples in Sections 5.1-5.4.

Please either read them in the lecture notes or watch the recording of the lecture from last spring.



Fundamental Matrices

5.5 Fundamental Matrices



Now consider

$$\mathbf{x}' = P(t)\mathbf{x}$$

where $P(t)$ is an $n \times n$ matrix function.

5.5 Fundamental Matrices



Now consider

$$\mathbf{x}' = P(t)\mathbf{x}$$

where $P(t)$ is an $n \times n$ matrix function.

Suppose that $\mathbf{x}^{(1)}, \mathbf{x}^{(2)}, \dots, \mathbf{x}^{(n)}$ are linearly independent solutions to this ODE. In other words, suppose that $\mathbf{x}^{(1)}, \mathbf{x}^{(2)}, \dots, \mathbf{x}^{(n)}$ form a *fundamental set of solutions* to this ODE.

5.5 Fundamental Matrices



Definition

The matrix

$$\Psi(t) = \begin{bmatrix} \mathbf{x}^{(1)} & \mathbf{x}^{(2)} & \dots & \mathbf{x}^{(n)} \end{bmatrix} = \begin{bmatrix} x_1^{(1)}(t) & x_1^{(2)}(t) & \dots & x_1^{(n)}(t) \\ x_2^{(1)}(t) & x_2^{(2)}(t) & \dots & x_2^{(n)}(t) \\ \vdots & \vdots & & \vdots \\ x_n^{(1)}(t) & x_n^{(2)}(t) & \dots & x_n^{(n)}(t) \end{bmatrix}$$

is called a *fundamental matrix* of $\mathbf{x}' = P(t)\mathbf{x}$.

5.5 Fundamental Matrices

Example

Find a fundamental matrix for

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

5.5 Fundamental Matrices

Example

Find a fundamental matrix for

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

Recall that

$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ 2 \end{bmatrix} e^{3t} \quad \text{and} \quad \mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ -2 \end{bmatrix} e^{-t}$$

form a fundamental set of solutions to this ODE.

5.5 Fundamental Matrices

Example

Find a fundamental matrix for

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

Recall that

$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ 2 \end{bmatrix} e^{3t} \quad \text{and} \quad \mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ -2 \end{bmatrix} e^{-t}$$

form a fundamental set of solutions to this ODE. Therefore

$$\Psi(t) = \begin{bmatrix} e^{3t} & e^{-t} \\ 2e^{3t} & -2e^{-t} \end{bmatrix}$$

is a fundamental matrix of this ODE.

5.5 Fundamental Matrices

Now, the general solution to

$$\mathbf{x}' = A\mathbf{x}$$

is

$$\mathbf{x}(t) = c_1 \mathbf{x}^{(1)}(t) + c_2 \mathbf{x}^{(2)}(t) + \dots + c_n \mathbf{x}^{(n)}(t)$$

5.5 Fundamental Matrices

Now, the general solution to

$$\mathbf{x}' = A\mathbf{x}$$

is

$$\mathbf{x}(t) = c_1 \mathbf{x}^{(1)}(t) + c_2 \mathbf{x}^{(2)}(t) + \dots + c_n \mathbf{x}^{(n)}(t) = \Psi(t)\mathbf{c}$$

where

$$\mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} \in \mathbb{R}^n.$$

5.5 Fundamental Matrices

Now, the general solution to

$$\mathbf{x}' = A\mathbf{x}$$

is

$$\mathbf{x}(t) = c_1 \mathbf{x}^{(1)}(t) + c_2 \mathbf{x}^{(2)}(t) + \dots + c_n \mathbf{x}^{(n)}(t) = \Psi(t)\mathbf{c}$$

where

$$\mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} \in \mathbb{R}^n.$$

If we have an initial condition $\mathbf{x}(t_0) = \mathbf{x}^0$, then

$$\Psi(t_0)\mathbf{c} = \mathbf{x}^0.$$

5.5 Fundamental Matrices



$$\mathbf{x}(t) = \Psi(t)\mathbf{c} \quad \Psi(t_0)\mathbf{c} = \mathbf{x}^0$$

But

$$\mathbf{x}^{(1)}, \dots, \mathbf{x}^{(n)}$$

are linearly
independent

5.5 Fundamental Matrices



$$\mathbf{x}(t) = \Psi(t)\mathbf{c} \quad \Psi(t_0)\mathbf{c} = \mathbf{x}^0$$

But

$\mathbf{x}^{(1)}, \dots, \mathbf{x}^{(n)}$
are linearly independent $\implies \Psi(t)$ is invertible

5.5 Fundamental Matrices



$$\mathbf{x}(t) = \Psi(t)\mathbf{c} \quad \Psi(t_0)\mathbf{c} = \mathbf{x}^0$$

But

$$\begin{aligned} \mathbf{x}^{(1)}, \dots, \mathbf{x}^{(n)} \\ \text{are linearly independent} &\implies \Psi(t) \text{ is invertible} \\ &\implies \mathbf{c} = \Psi^{-1}(t_0)\mathbf{x}^0. \end{aligned}$$

5.5 Fundamental Matrices



$$\mathbf{x}(t) = \Psi(t)\mathbf{c} \quad \Psi(t_0)\mathbf{c} = \mathbf{x}^0$$

But

$$\begin{aligned} \mathbf{x}^{(1)}, \dots, \mathbf{x}^{(n)} \\ \text{are linearly independent} \end{aligned} \implies \Psi(t) \text{ is invertible} \implies \mathbf{c} = \Psi^{-1}(t_0)\mathbf{x}^0.$$

Therefore the solution to the IVP

$$\begin{cases} \mathbf{x}' = A\mathbf{x} \\ \mathbf{x}(t_0) = \mathbf{x}^0 \end{cases}$$

is

$$\boxed{\mathbf{x}(t) = \Psi(t)\Psi^{-1}(t_0)\mathbf{x}^0.}$$

5.5 Fundamental Matrices



Theorem

Suppose that $\Psi(t)$ is a fundamental matrix for $\mathbf{x}' = P(t)\mathbf{x}$. Then $\Psi(t)$ solves the differential equation $\Psi' = P(t)\Psi$.

(You prove)

5.5 Fundamental Matrices



Remark

It is possible to find a *special fundamental matrix*, $\Phi(t)$, which satisfies

$$\Phi(t_0) = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} = I.$$

5.5 Fundamental Matrices



Remark

It is possible to find a *special fundamental matrix*, $\Phi(t)$, which satisfies

$$\Phi(t_0) = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} = I.$$

We will use Φ for this special fundamental matrix, and Ψ for any fundamental matrix.

5.5 Fundamental Matrices



Example

Consider

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

Find the special fundamental matrix which satisfies $\Phi(0) = I$.

5.5 Fundamental Matrices



To find the matrix Φ which satisfies

$$\Phi(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix},$$

we must solve two IVPs:

$$\begin{cases} \mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x} \\ \mathbf{x}(0) = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \end{cases} \quad \text{and}$$

$$\begin{cases} \mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x} \\ \mathbf{x}(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}. \end{cases}$$

5.5 Fundamental Matrices



The general solution to the ODE is

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} e^{3t} + c_2 \begin{bmatrix} 1 \\ -2 \end{bmatrix} e^{-t}.$$

5.5 Fundamental Matrices

We calculate that

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ -2 \end{bmatrix} \implies \begin{aligned} c_1 &= \frac{1}{2} \\ c_2 &= \frac{1}{2} \end{aligned}$$

$$\implies \mathbf{x}(t) = \begin{bmatrix} \frac{1}{2}e^{3t} + \frac{1}{2}e^{-t} \\ e^{3t} - e^{-t} \end{bmatrix}$$

5.5 Fundamental Matrices

We calculate that

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ -2 \end{bmatrix} \implies \begin{aligned} c_1 &= \frac{1}{2} \\ c_2 &= \frac{1}{2} \end{aligned}$$

$$\implies \mathbf{x}(t) = \begin{bmatrix} \frac{1}{2}e^{3t} + \frac{1}{2}e^{-t} \\ e^{3t} - e^{-t} \end{bmatrix}$$

and

$$\begin{bmatrix} 0 \\ 1 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ -2 \end{bmatrix} \implies \begin{aligned} c_1 &= \frac{1}{4} \\ c_2 &= -\frac{1}{4} \end{aligned}$$

$$\implies \mathbf{x}(t) = \begin{bmatrix} \frac{1}{4}e^{3t} - \frac{1}{4}e^{-t} \\ \frac{1}{2}e^{3t} + \frac{1}{2}e^{-t} \end{bmatrix}.$$

5.5 Fundamental Matrices



Therefore the special fundamental matrix is

$$\Phi(t) = \begin{bmatrix} \frac{1}{2}e^{3t} + \frac{1}{2}e^{-t} & \frac{1}{4}e^{3t} - \frac{1}{4}e^{-t} \\ e^{3t} - e^{-t} & \frac{1}{2}e^{3t} + \frac{1}{2}e^{-t} \end{bmatrix}.$$

5.5 Fundamental Matrices



What is e^{At} ?

Recall that the solution to

$$\begin{cases} x' = ax \quad (a \in \mathbb{R}) \\ x(0) = x_0 \end{cases}$$

is

$$x(t) = x_0 e^{at} = x_0 \exp(at)$$

and recall that

$$\exp(at) = 1 + \sum_{n=1}^{\infty} \frac{a^n t^n}{n!}.$$

5.5 Fundamental Matrices



Now consider

$$\begin{cases} \mathbf{x}' = A\mathbf{x} \\ \mathbf{x}(0) = \mathbf{x}^0 \end{cases}$$

for $A \in \mathbb{R}^{n \times n}$.

5.5 Fundamental Matrices



Now consider

$$\begin{cases} \mathbf{x}' = A\mathbf{x} \\ \mathbf{x}(0) = \mathbf{x}^0 \end{cases}$$

for $A \in \mathbb{R}^{n \times n}$.

Definition

$$\exp(At) = I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} = I + At + \frac{A^2 t^2}{2!} + \frac{A^3 t^3}{3!} + \dots$$

5.5 Fundamental Matrices



Note that

$$\frac{d}{dt} \exp(At) = \quad =$$

$$= \quad =$$

$$=$$

$$= \quad =$$

.

5.5 Fundamental Matrices

Note that

$$\frac{d}{dt} \exp(At) = \frac{d}{dt} \left(I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} \right) =$$

$$= =$$

$$=$$

$$= =$$

.

5.5 Fundamental Matrices

Note that

$$\frac{d}{dt} \exp(At) = \frac{d}{dt} \left(I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} \right) = \mathbf{0} + \sum_{n=1}^{\infty} \frac{d}{dt} \left(\frac{A^n t^n}{n!} \right)$$

$$= =$$

$$=$$

$$= = .$$

5.5 Fundamental Matrices

Note that

$$\begin{aligned}
 \frac{d}{dt} \exp(At) &= \frac{d}{dt} \left(I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} \right) = \mathbf{0} + \sum_{n=1}^{\infty} \frac{d}{dt} \left(\frac{A^n t^n}{n!} \right) \\
 &= \sum_{n=1}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} = \\
 &= \dots = \dots = .
 \end{aligned}$$

5.5 Fundamental Matrices



Note that

$$\frac{d}{dt} \exp(At) = \frac{d}{dt} \left(I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} \right) = \mathbf{0} + \sum_{n=1}^{\infty} \frac{d}{dt} \left(\frac{A^n t^n}{n!} \right)$$

$$= \sum_{n=1}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} = A + \sum_{n=2}^{\infty} \frac{A^n t^{n-1}}{(n-1)!}$$

=

=

.

5.5 Fundamental Matrices

Note that

$$\begin{aligned}
 \frac{d}{dt} \exp(At) &= \frac{d}{dt} \left(I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} \right) = \mathbf{0} + \sum_{n=1}^{\infty} \frac{d}{dt} \left(\frac{A^n t^n}{n!} \right) \\
 &= \sum_{n=1}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} = A + \sum_{n=2}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} \\
 &= A + \sum_{k=1}^{\infty} \frac{A^{k+1} t^k}{k!} \quad (k = n-1) \\
 \\
 &= \quad \quad \quad = \quad \quad \quad .
 \end{aligned}$$

5.5 Fundamental Matrices



Note that

$$\begin{aligned}\frac{d}{dt} \exp(At) &= \frac{d}{dt} \left(I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} \right) = \mathbf{0} + \sum_{n=1}^{\infty} \frac{d}{dt} \left(\frac{A^n t^n}{n!} \right) \\ &= \sum_{n=1}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} = A + \sum_{n=2}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} \\ &= A + \sum_{k=1}^{\infty} \frac{A^{k+1} t^k}{k!} \quad (k = n-1) \\ &= A \left(I + \sum_{k=1}^{\infty} \frac{A^k t^k}{k!} \right) = .\end{aligned}$$

5.5 Fundamental Matrices



Note that

$$\begin{aligned}\frac{d}{dt} \exp(At) &= \frac{d}{dt} \left(I + \sum_{n=1}^{\infty} \frac{A^n t^n}{n!} \right) = \mathbf{0} + \sum_{n=1}^{\infty} \frac{d}{dt} \left(\frac{A^n t^n}{n!} \right) \\ &= \sum_{n=1}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} = A + \sum_{n=2}^{\infty} \frac{A^n t^{n-1}}{(n-1)!} \\ &= A + \sum_{k=1}^{\infty} \frac{A^{k+1} t^k}{k!} \quad (k = n-1) \\ &= A \left(I + \sum_{k=1}^{\infty} \frac{A^k t^k}{k!} \right) = A \exp(At).\end{aligned}$$

5.5 Fundamental Matrices



This means that $\exp(At)$ solves

$$\begin{cases} (\exp(At))' = A \exp(At) \\ \exp(At)|_{t=0} = I. \end{cases}$$

5.5 Fundamental Matrices

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$$\begin{cases} (\exp(At))' = A \exp(At) \\ \exp(At)|_{t=0} = I. \end{cases}$$

But remember that Φ solves

$$\begin{cases} \Phi' = A\Phi \\ \Phi(0) = I. \end{cases}$$

5.5 Fundamental Matrices

This means that $\exp(At)$ solves

$$\begin{cases} (\exp(At))' = A \exp(At) \\ \exp(At)|_{t=0} = I. \end{cases}$$

But remember that Φ solves

$$\begin{cases} \Phi' = A\Phi \\ \Phi(0) = I. \end{cases}$$

Therefore

$$\boxed{\Phi(t) = \exp(At).}$$

5.5 Fundamental Matrices



Example

Let $A = \begin{bmatrix} 8 & -1 \\ 6 & 1 \end{bmatrix}$. Find $\exp(At)$.

5.5 Fundamental Matrices



Example

Let $A = \begin{bmatrix} 8 & -1 \\ 6 & 1 \end{bmatrix}$. Find $\exp(At)$.

We have previously found that the general solution to $\mathbf{x}' = A\mathbf{x}$ is

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + c_2 \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t}.$$

5.5 Fundamental Matrices



To satisfy $\mathbf{x}(0) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ we require $c_1 = \frac{6}{5}$ and $c_2 = -\frac{1}{5}$. Hence

$$\mathbf{x}(t) = \frac{6}{5} \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} - \frac{1}{5} \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t} = \begin{bmatrix} \frac{6}{5}e^{7t} - \frac{1}{5}e^{2t} \\ \frac{6}{5}e^{7t} - \frac{6}{5}e^{2t} \end{bmatrix}.$$

5.5 Fundamental Matrices

To satisfy $\mathbf{x}(0) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ we require $c_1 = \frac{6}{5}$ and $c_2 = -\frac{1}{5}$. Hence

$$\mathbf{x}(t) = \frac{6}{5} \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} - \frac{1}{5} \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t} = \begin{bmatrix} \frac{6}{5}e^{7t} - \frac{1}{5}e^{2t} \\ \frac{6}{5}e^{7t} - \frac{6}{5}e^{2t} \end{bmatrix}.$$

To satisfy $\mathbf{x}(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ we require $c_1 = -\frac{1}{5}$ and $c_2 = \frac{1}{5}$. Hence

$$\mathbf{x}(t) = -\frac{1}{5} \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{7t} + \frac{1}{5} \begin{bmatrix} 1 \\ 6 \end{bmatrix} e^{2t} = \begin{bmatrix} -\frac{1}{5}e^{7t} + \frac{1}{5}e^{2t} \\ -\frac{1}{5}e^{7t} + \frac{6}{5}e^{2t} \end{bmatrix}.$$

5.5 Fundamental Matrices



Therefore the answer is

$$\exp(At) = \Phi(t) = \begin{bmatrix} \frac{6}{5}e^{7t} - \frac{1}{5}e^{2t} & -\frac{1}{5}e^{7t} + \frac{1}{5}e^{2t} \\ \frac{6}{5}e^{7t} - \frac{6}{5}e^{2t} & -\frac{1}{5}e^{7t} + \frac{6}{5}e^{2t} \end{bmatrix}.$$

Diagonalisable Matrices

If

$$D = \begin{bmatrix} r_1 & 0 & 0 & \dots & 0 \\ 0 & r_2 & 0 & \dots & 0 \\ 0 & 0 & r_3 & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & 0 & \dots & r_n \end{bmatrix}$$

is a diagonal matrix, then it is easy to calculate $\exp(Dt)$. We simply have

$$\exp(Dt) = \begin{bmatrix} e^{r_1 t} & 0 & 0 & \dots & 0 \\ 0 & e^{r_2 t} & 0 & \dots & 0 \\ 0 & 0 & e^{r_3 t} & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & 0 & \dots & e^{r_n t} \end{bmatrix}.$$

5.5 Fundamental Matrices



Now consider

$$\mathbf{x}' = A\mathbf{x}$$

for $A \in \mathbb{R}^{n \times n}$.

5.5 Fundamental Matrices



Now consider

$$\mathbf{x}' = A\mathbf{x}$$

for $A \in \mathbb{R}^{n \times n}$. Recall how we diagonalise a matrix: If $\boldsymbol{\xi}^{(1)}, \boldsymbol{\xi}^{(2)}, \dots, \boldsymbol{\xi}^{(n)}$ are the eigenvectors of A , we let

$$T = \begin{bmatrix} \boldsymbol{\xi}^{(1)} & \boldsymbol{\xi}^{(2)} & \dots & \boldsymbol{\xi}^{(n)} \end{bmatrix}.$$

5.5 Fundamental Matrices

Now consider

$$\mathbf{x}' = A\mathbf{x}$$

for $A \in \mathbb{R}^{n \times n}$. Recall how we diagonalise a matrix: If $\boldsymbol{\xi}^{(1)}, \boldsymbol{\xi}^{(2)}, \dots, \boldsymbol{\xi}^{(n)}$ are the eigenvectors of A , we let

$$T = \begin{bmatrix} \boldsymbol{\xi}^{(1)} & \boldsymbol{\xi}^{(2)} & \dots & \boldsymbol{\xi}^{(n)} \end{bmatrix}.$$

Then

$$\det(T) \neq 0 \implies T^{-1} \text{ exists}$$

5.5 Fundamental Matrices

Now consider

$$\mathbf{x}' = A\mathbf{x}$$

for $A \in \mathbb{R}^{n \times n}$. Recall how we diagonalise a matrix: If $\xi^{(1)}, \xi^{(2)}, \dots, \xi^{(n)}$ are the eigenvectors of A , we let

$$T = \begin{bmatrix} \xi^{(1)} & \xi^{(2)} & \dots & \xi^{(n)} \end{bmatrix}.$$

Then

$$\det(T) \neq 0 \implies \begin{array}{l} T^{-1} \\ \text{exists} \end{array} \implies \begin{array}{l} T^{-1}AT \\ \text{is diagonal} \end{array}$$

5.5 Fundamental Matrices

Now consider

$$\mathbf{x}' = A\mathbf{x}$$

for $A \in \mathbb{R}^{n \times n}$. Recall how we diagonalise a matrix: If $\boldsymbol{\xi}^{(1)}, \boldsymbol{\xi}^{(2)}, \dots, \boldsymbol{\xi}^{(n)}$ are the eigenvectors of A , we let

$$T = \begin{bmatrix} \boldsymbol{\xi}^{(1)} & \boldsymbol{\xi}^{(2)} & \dots & \boldsymbol{\xi}^{(n)} \end{bmatrix}.$$

Then

$$\det(T) \neq 0 \implies T^{-1} \text{ exists} \implies T^{-1}AT \text{ is diagonal} \implies A \text{ is diagonalisable.}$$

5.5 Fundamental Matrices

Example

Diagonalise

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}.$$

5.5 Fundamental Matrices

Example

Diagonalise

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}.$$

The eigenvalues are $r_1 = 3$ and $r_2 = -1$. The corresponding eigenvectors are $\xi^{(1)} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $\xi^{(2)} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$.

5.5 Fundamental Matrices

Example

Diagonalise

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}.$$

The eigenvalues are $r_1 = 3$ and $r_2 = -1$. The corresponding eigenvectors are $\xi^{(1)} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $\xi^{(2)} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$. Thus

$$T = \begin{bmatrix} 1 & 1 \\ 2 & -2 \end{bmatrix}$$

5.5 Fundamental Matrices



Example

Diagonalise

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}.$$

The eigenvalues are $r_1 = 3$ and $r_2 = -1$. The corresponding eigenvectors are $\xi^{(1)} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $\xi^{(2)} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$. Thus

$$T = \begin{bmatrix} 1 & 1 \\ 2 & -2 \end{bmatrix} \quad \text{and} \quad T^{-1} = \begin{bmatrix} \frac{1}{2} & \frac{1}{4} \\ \frac{1}{2} & -\frac{1}{4} \end{bmatrix}.$$

5.5 Fundamental Matrices

Example

Diagonalise

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}.$$

The eigenvalues are $r_1 = 3$ and $r_2 = -1$. The corresponding eigenvectors are $\xi^{(1)} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $\xi^{(2)} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$. Thus

$$T = \begin{bmatrix} 1 & 1 \\ 2 & -2 \end{bmatrix} \quad \text{and} \quad T^{-1} = \begin{bmatrix} \frac{1}{2} & \frac{1}{4} \\ \frac{1}{2} & -\frac{1}{4} \end{bmatrix}.$$

It follows that

$$D = T^{-1}AT = \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix}.$$

5.5 Fundamental Matrices



Now consider

$$\mathbf{x}' = A\mathbf{x}.$$

5.5 Fundamental Matrices



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Define a new variable \mathbf{y} by

$$\mathbf{x} = T\mathbf{y} \quad \text{or} \quad \mathbf{y} = T^{-1}\mathbf{x}.$$

5.5 Fundamental Matrices

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5.5 Fundamental Matrices

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5.5 Fundamental Matrices



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Define a new variable \mathbf{y} by

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Then we calculate that

$$\mathbf{x}' = A\mathbf{x}$$

$$T\mathbf{y}' = AT\mathbf{y}$$

$$\mathbf{y}' = T^{-1}AT\mathbf{y} = D\mathbf{y}.$$

5.5 Fundamental Matrices



We know that a fundamental matrix for $\mathbf{y}' = D\mathbf{y}$ is

$$\exp(Dt) = \begin{bmatrix} e^{r_1 t} & 0 & \dots & 0 \\ 0 & e^{r_2 t} & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & e^{r_n t} \end{bmatrix}.$$

5.5 Fundamental Matrices



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Therefore a fundamental matrix for $\mathbf{x}' = A\mathbf{x}$ is

$$\Psi = T \exp(Dt) = \begin{bmatrix} \boldsymbol{\xi}^{(1)} e^{r_1 t} & \boldsymbol{\xi}^{(2)} e^{r_2 t} & \dots & \boldsymbol{\xi}^{(n)} e^{r_n t} \end{bmatrix}.$$

5.5 Fundamental Matrices



Example

Find a fundamental matrix for

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

5.5 Fundamental Matrices

Example

Find a fundamental matrix for

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

Recall that $T = \begin{bmatrix} 1 & 1 \\ 2 & -2 \end{bmatrix}$.

5.5 Fundamental Matrices

Example

Find a fundamental matrix for

$$\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}.$$

Recall that $T = \begin{bmatrix} 1 & 1 \\ 2 & -2 \end{bmatrix}$. Letting $\mathbf{y} = T^{-1}\mathbf{x}$, we have

$$\mathbf{y}' = \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix} \mathbf{y}.$$

5.5 Fundamental Matrices

A fundamental matrix for $\mathbf{y}' = \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix} \mathbf{y}$ is

$$\exp(Dt) = e^{Dt} = \begin{bmatrix} e^{3t} & 0 \\ 0 & e^{-t} \end{bmatrix}.$$

5.5 Fundamental Matrices

A fundamental matrix for $\mathbf{y}' = \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix} \mathbf{y}$ is

$$\exp(Dt) = e^{Dt} = \begin{bmatrix} e^{3t} & 0 \\ 0 & e^{-t} \end{bmatrix}.$$

Hence a fundamental matrix for $\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}$ is

$$\Psi(t) = T \exp(Dt)$$

5.5 Fundamental Matrices

A fundamental matrix for $\mathbf{y}' = \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix} \mathbf{y}$ is

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5.5 Fundamental Matrices

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Hence a fundamental matrix for $\mathbf{x}' = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \mathbf{x}$ is

$$\Psi(t) = T \exp(Dt) = \begin{bmatrix} 1 & 1 \\ 2 & -2 \end{bmatrix} \begin{bmatrix} e^{3t} & 0 \\ 0 & e^{-t} \end{bmatrix} = \begin{bmatrix} e^{3t} & e^{-t} \\ 2e^{3t} & -2e^{-t} \end{bmatrix}.$$



Repeated Eigenvalues

5.6 Repeated Eigenvalues

Example

Find the eigenvalues and eigenvectors of $A = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix}$.

5.6 Repeated Eigenvalues

Example

Find the eigenvalues and eigenvectors of $A = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix}$.

We calculate that

$$0 = \det(A - rI) = \begin{vmatrix} 1-r & -1 \\ 1 & 3-r \end{vmatrix} = r^2 - 4r + 4 = (r - 2)^2.$$

Therefore $r_1 = 2 = r_2$.

5.6 Repeated Eigenvalues

Example

Find the eigenvalues and eigenvectors of $A = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix}$.

We calculate that

$$0 = \det(A - rI) = \begin{vmatrix} 1-r & -1 \\ 1 & 3-r \end{vmatrix} = r^2 - 4r + 4 = (r-2)^2.$$

Therefore $r_1 = 2 = r_2$. Moreover

$$\mathbf{0} = (A - rI)\boldsymbol{\xi} = \begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \xi_1 \\ \xi_2 \end{bmatrix} \implies \xi_1 + \xi_2 = 0 \implies \boldsymbol{\xi} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

5.6 Repeated Eigenvalues

Example

Find the eigenvalues and eigenvectors of $A = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix}$.

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Note that A has only one linearly independent eigenvector.

5.6 Repeated Eigenvalues



Example

Solve

$$\mathbf{x}' = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix} \mathbf{x}.$$

5.6 Repeated Eigenvalues



Example

Solve

$$\mathbf{x}' = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix} \mathbf{x}.$$

We know that

$$\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{2t}$$

is a solution. But we need two solutions.

5.6 Repeated Eigenvalues



Guess 1: I guess that

$$\mathbf{x}^{(2)}(t) = \xi t e^{2t}$$

for some $\xi \in \mathbb{R}^2$.

5.6 Repeated Eigenvalues



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for some $\xi \in \mathbb{R}^2$. Then we have

$$\mathbf{x}^{(2)\prime} = A\mathbf{x}^{(2)}$$

5.6 Repeated Eigenvalues



Guess 1: I guess that

$$\mathbf{x}^{(2)}(t) = \xi t e^{2t}$$

for some $\xi \in \mathbb{R}^2$. Then we have

$$\xi e^{2t} + 2\xi t e^{2t} = \mathbf{x}^{(2)\prime} = A\mathbf{x}^{(2)}$$

5.6 Repeated Eigenvalues

Guess 1: I guess that

$$\mathbf{x}^{(2)}(t) = \boldsymbol{\xi} t e^{2t}$$

for some $\boldsymbol{\xi} \in \mathbb{R}^2$. Then we have

$$\boldsymbol{\xi} e^{2t} + 2\boldsymbol{\xi} t e^{2t} = \mathbf{x}^{(2)\prime} = A\mathbf{x}^{(2)} = A\boldsymbol{\xi} t e^{2t}$$

5.6 Repeated Eigenvalues



Guess 1: I guess that

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for some $\boldsymbol{\xi} \in \mathbb{R}^2$. Then we have

$$\begin{aligned}\boldsymbol{\xi} e^{2t} + 2\boldsymbol{\xi} t e^{2t} &= \mathbf{x}^{(2)\prime} = A\mathbf{x}^{(2)} = A\boldsymbol{\xi} t e^{2t} \\ \boldsymbol{\xi} + (2\boldsymbol{\xi} - A\boldsymbol{\xi})t &= \mathbf{0} \quad \forall t\end{aligned}$$

5.6 Repeated Eigenvalues

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This guess did not work.

5.6 Repeated Eigenvalues



Guess 2: Now I guess that

$$\mathbf{x}^{(2)}(t) = \xi t e^{2t} + \eta e^{2t}$$

for some $\xi, \eta \in \mathbb{R}^2$.

5.6 Repeated Eigenvalues



Guess 2: Now I guess that

$$\mathbf{x}^{(2)}(t) = \xi t e^{2t} + \eta e^{2t}$$

for some $\xi, \eta \in \mathbb{R}^2$. Then we have

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5.6 Repeated Eigenvalues

Guess 2: Now I guess that

$$\mathbf{x}^{(2)}(t) = \xi te^{2t} + \eta e^{2t}$$

for some $\xi, \eta \in \mathbb{R}^2$. Then we have

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5.6 Repeated Eigenvalues

Guess 2: Now I guess that

$$\mathbf{x}^{(2)}(t) = \xi te^{2t} + \eta e^{2t}$$

for some $\xi, \eta \in \mathbb{R}^2$. Then we have

$$\xi e^{2t} + 2\xi t e^{2t} + 2\eta e^{2t} = \mathbf{x}^{(2)\prime} = A\mathbf{x}^{(2)} = A(\xi te^{2t} + \eta e^{2t})$$

5.6 Repeated Eigenvalues

Guess 2: Now I guess that

$$\mathbf{x}^{(2)}(t) = \boldsymbol{\xi} t e^{2t} + \boldsymbol{\eta} e^{2t}$$

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and

$$(2\boldsymbol{\xi} - A\boldsymbol{\xi})t + (\boldsymbol{\xi} + 2\boldsymbol{\eta} - A\boldsymbol{\eta}) = \mathbf{0}.$$

5.6 Repeated Eigenvalues

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$$\mathbf{x}^{(2)}(t) = \boldsymbol{\xi} t e^{2t} + \boldsymbol{\eta} e^{2t}$$

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and

$$(2\boldsymbol{\xi} - A\boldsymbol{\xi})t + (\boldsymbol{\xi} + 2\boldsymbol{\eta} - A\boldsymbol{\eta}) = \mathbf{0}.$$

Since this must be true $\forall t$, we must have

$$(A - 2I)\boldsymbol{\xi} = \mathbf{0} \quad \text{and} \quad (A - 2I)\boldsymbol{\eta} = \boldsymbol{\xi}.$$

5.6

$$(A - 2I) \boldsymbol{\xi} = \mathbf{0} \quad (A - 2I) \boldsymbol{\eta} = \boldsymbol{\xi}$$



Clearly $\boldsymbol{\xi} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$.

5.6

$$(A - 2I) \boldsymbol{\xi} = \mathbf{0} \quad (A - 2I) \boldsymbol{\eta} = \boldsymbol{\xi}$$



Clearly $\boldsymbol{\xi} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$. Then we calculate that

$$\begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \implies \eta_1 + \eta_2 = -1$$

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$$\implies \boldsymbol{\eta} = \begin{bmatrix} k \\ -1 - k \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix} + k \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

for some k .

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for some k . So

$$\mathbf{x}^{(2)}(t) = \boldsymbol{\xi} t e^{2t} + \boldsymbol{\eta} e^{2t} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} t e^{2t} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^{2t} + k \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{2t}$$

5.6

$$(A - 2I) \boldsymbol{\xi} = \mathbf{0} \quad (A - 2I) \boldsymbol{\eta} = \boldsymbol{\xi}$$



Clearly $\boldsymbol{\xi} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$. Then we calculate that

$$\begin{aligned} \begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} &= \begin{bmatrix} 1 \\ -1 \end{bmatrix} \implies \eta_1 + \eta_2 = -1 \\ \implies \boldsymbol{\eta} &= \begin{bmatrix} k \\ -1 - k \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix} + k \begin{bmatrix} 1 \\ -1 \end{bmatrix} \end{aligned}$$

for some k . So

$$\begin{aligned} \mathbf{x}^{(2)}(t) &= \boldsymbol{\xi} t e^{2t} + \boldsymbol{\eta} e^{2t} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} t e^{2t} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^{2t} + k \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{2t} \\ &= \begin{bmatrix} 1 \\ -1 \end{bmatrix} t e^{2t} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^{2t} + k \mathbf{x}^{(1)}(t). \end{aligned}$$

5.6 Repeated Eigenvalues



$$\mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ -1 \end{bmatrix} te^{2t} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^{2t} + k\mathbf{x}^{(1)}(t)$$

Because we already have $\mathbf{x}^{(1)}(t)$, we can choose $k = 0$. So

$$\mathbf{x}^{(2)}(t) = \xi te^{2t} + \eta e^{2t} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} te^{2t} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^{2t}.$$

5.6 Repeated Eigenvalues



The general solution of $\mathbf{x}' = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix} \mathbf{x}$ is therefore

$$\mathbf{x}(t) = c_1 \mathbf{x}^{(1)} + c_2 \mathbf{x}^{(2)} = c_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{2t} + c_2 \left(\begin{bmatrix} 1 \\ -1 \end{bmatrix} t e^{2t} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^{2t} \right).$$

5.6 Repeated Eigenvalues

Example

Find a fundamental matrix for

$$\mathbf{x}' = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix} \mathbf{x}.$$

Then find the special fundamental matrix $\Phi(t)$ which satisfies $\Phi(0) = I$.

5.6 Repeated Eigenvalues



Since $\mathbf{x}^{(1)}(t) = \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{2t}$ and $\mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ -1 \end{bmatrix} te^{2t} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^{2t}$ we have that

$$\Psi(t) = \begin{bmatrix} \mathbf{x}^{(1)} & \mathbf{x}^{(2)} \end{bmatrix} = \begin{bmatrix} e^{2t} & te^{2t} \\ -e^{2t} & -te^{2t} - e^{2t} \end{bmatrix}$$

is a fundamental matrix for this system.

5.6

$$\Psi(t) = \begin{bmatrix} e^{2t} & te^{2t} \\ -e^{2t} & -te^{2t} - e^{2t} \end{bmatrix}$$



Now

$$\Psi(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix} \quad \text{and} \quad \Psi^{-1}(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix}.$$

5.6

$$\Psi(t) = \begin{bmatrix} e^{2t} & te^{2t} \\ -e^{2t} & -te^{2t} - e^{2t} \end{bmatrix}$$



Now

$$\Psi(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix} \quad \text{and} \quad \Psi^{-1}(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix}.$$

Therefore

$$\exp(At) = \Phi(t) = \Psi(t)\Psi^{-1}(0)$$

5.6

$$\Psi(t) = \begin{bmatrix} e^{2t} & te^{2t} \\ -e^{2t} & -te^{2t} - e^{2t} \end{bmatrix}$$



Now

$$\Psi(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix} \quad \text{and} \quad \Psi^{-1}(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix}.$$

Therefore

$$\exp(At) = \Phi(t) = \Psi(t)\Psi^{-1}(0) = e^{2t} \begin{bmatrix} 1 & t \\ -1 & -1-t \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix}$$

5.6

$$\Psi(t) = \begin{bmatrix} e^{2t} & te^{2t} \\ -e^{2t} & -te^{2t} - e^{2t} \end{bmatrix}$$



Now

$$\Psi(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix} \quad \text{and} \quad \Psi^{-1}(0) = \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix}.$$

Therefore

$$\begin{aligned} \exp(At) &= \Phi(t) = \Psi(t)\Psi^{-1}(0) = e^{2t} \begin{bmatrix} 1 & t \\ -1 & -1-t \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -1 & -1 \end{bmatrix} \\ &= e^{2t} \begin{bmatrix} 1-t & -t \\ t & 1+t \end{bmatrix}. \end{aligned}$$

5.6 Repeated Eigenvalues



Remark

$$\mathbf{x}' = A\mathbf{x}$$

For two repeated eigenvalues (but with only one linearly independent eigenvector), the key equations to remember are

$$\boxed{\mathbf{x}^{(2)}(t) = \xi t e^{rt} + \eta e^{rt}} \quad \text{and} \quad \boxed{(A - rI)\eta = \xi}.$$

5.6 Repeated Eigenvalues

Remark

$$\mathbf{x}' = A\mathbf{x}$$

For two repeated eigenvalues (but with only one linearly independent eigenvector), the key equations to remember are

$$\mathbf{x}^{(2)}(t) = \xi t e^{rt} + \eta e^{rt}$$

and

$$(A - rI)\eta = \xi.$$

Definition

η is called a *generalised eigenvector* of A .

5.6 Repeated Eigenvalues



Remark

If you have 2 repeated eigenvalues (but with only one linearly independent eigenvector), the method is:

- 1 Find the eigenvalues and eigenvectors;
- 2 The first solution is $\mathbf{x}^{(1)}(t) = \boldsymbol{\xi} e^{rt}$;
- 3 Use $(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$ to find a generalised eigenvector $\boldsymbol{\eta}$;
- 4 The second solution is $\mathbf{x}^{(2)}(t) = \boldsymbol{\xi} t e^{rt} + \boldsymbol{\eta} e^{rt}$.

5.6 Repeated Eigenvalues



Example

Solve

$$\begin{cases} \mathbf{x}' = \begin{bmatrix} -\frac{5}{2} & \frac{3}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{bmatrix} \mathbf{x}, \\ \mathbf{x}(0) = \begin{bmatrix} 3 \\ -1 \end{bmatrix}. \end{cases}$$

5.6 Repeated Eigenvalues



The only eigenvalue of the matrix is $r = -1$. The corresponding eigenvector is $\xi = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$. Therefore one solution of the linear system is

$$\mathbf{x}^{(1)}(t) = \xi e^{rt} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-t}.$$

5.6 Repeated Eigenvalues



We need to find a second, linearly independent solution. We will consider the ansatz

$$\mathbf{x}^{(2)}(t) = \boldsymbol{\xi} t e^{-t} + \boldsymbol{\eta} e^{-t}$$

where $\boldsymbol{\xi} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ as above and $\boldsymbol{\eta}$ is a generalised eigenvector solving $(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$.

5.6 Repeated Eigenvalues



Solving the latter equation,

$$(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$$

5.6 Repeated Eigenvalues



Solving the latter equation,

$$(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$$
$$\begin{bmatrix} -\frac{3}{2} & \frac{3}{2} \\ -\frac{3}{2} & \frac{3}{2} \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

5.6 Repeated Eigenvalues



Solving the latter equation,

$$(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$$
$$\begin{bmatrix} -\frac{3}{2} & \frac{3}{2} \\ -\frac{3}{2} & \frac{3}{2} \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
$$-\frac{3}{2}\eta_1 + \frac{3}{2}\eta_2 = 1$$

5.6 Repeated Eigenvalues

Solving the latter equation,

$$(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$$
$$\begin{bmatrix} -\frac{3}{2} & \frac{3}{2} \\ -\frac{3}{2} & \frac{3}{2} \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
$$-\frac{3}{2}\eta_1 + \frac{3}{2}\eta_2 = 1$$
$$-\eta_1 + \eta_2 = \frac{2}{3}$$

5.6 Repeated Eigenvalues

Solving the latter equation,

$$(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$$

$$\begin{bmatrix} -\frac{3}{2} & \frac{3}{2} \\ -\frac{3}{2} & \frac{3}{2} \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$-\frac{3}{2}\eta_1 + \frac{3}{2}\eta_2 = 1$$

$$-\eta_1 + \eta_2 = \frac{2}{3}$$

we can choose $\boldsymbol{\eta} = \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$.

5.6 Repeated Eigenvalues



Note that we don't need to find *every* generalised eigenvector

$$\boldsymbol{\eta} = \begin{bmatrix} k \\ k + \frac{2}{3} \end{bmatrix} = k \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix} = k\boldsymbol{\xi} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$$

because we already have $\mathbf{x}^{(1)}(t) = \boldsymbol{\xi} e^{rt}$.

5.6 Repeated Eigenvalues



Note that we don't need to find *every* generalised eigenvector

$$\boldsymbol{\eta} = \begin{bmatrix} k \\ k + \frac{2}{3} \end{bmatrix} = k \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix} = k\boldsymbol{\xi} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$$

because we already have $\mathbf{x}^{(1)}(t) = \boldsymbol{\xi} e^{rt}$.

Instead we only need to find *one* generalised eigenvector – that means that we can choose any k that we want.

5.6 Repeated Eigenvalues



Note that we don't need to find *every* generalised eigenvector

$$\boldsymbol{\eta} = \begin{bmatrix} k \\ k + \frac{2}{3} \end{bmatrix} = k \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix} = k\boldsymbol{\xi} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$$

because we already have $\mathbf{x}^{(1)}(t) = \boldsymbol{\xi} e^{rt}$.

Instead we only need to find *one* generalised eigenvector – that means that we can choose any k that we want.

Hence I have chosen $k = 0$ which gives $\boldsymbol{\eta} = \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$.

5.6 Repeated Eigenvalues

eigenvector

$$\xi = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

generalised eigenvector

$$\eta = \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$$

Thus

$$\mathbf{x}^{(1)}(t) = \xi e^{-t} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-t}$$

and

$$\mathbf{x}^{(2)}(t) = \xi t e^{-t} + \eta e^{-t} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-t} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix} e^{-t}.$$

5.6 Repeated Eigenvalues

eigenvector

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Thus

$$\mathbf{x}^{(1)}(t) = \xi e^{-t} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-t}$$

and

$$\mathbf{x}^{(2)}(t) = \xi t e^{-t} + \eta e^{-t} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-t} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix} e^{-t}.$$

Hence the general solution to the linear system is

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-t} + c_2 \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-t} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix} e^{-t} \right).$$

5.6 Repeated Eigenvalues



The initial condition gives

$$\begin{bmatrix} 3 \\ -1 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$$

which implies that $c_1 = 3$ and $c_2 = -6$.

5.6 Repeated Eigenvalues

The initial condition gives

$$\begin{bmatrix} 3 \\ -1 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix}$$

which implies that $c_1 = 3$ and $c_2 = -6$.

Therefore the solution to the IVP is

$$\boxed{\mathbf{x}(t) = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-t} - 6 \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-t} + \begin{bmatrix} 0 \\ \frac{2}{3} \end{bmatrix} e^{-t} \right) = \begin{bmatrix} 3 \\ -1 \end{bmatrix} e^{-t} - 6 \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-t}.}$$

5.6 Repeated Eigenvalues



Example

Solve

$$\mathbf{x}' = \begin{bmatrix} 1 & -4 \\ 4 & -7 \end{bmatrix} \mathbf{x}, \quad \mathbf{x}(0) = \begin{bmatrix} 3 \\ 2 \end{bmatrix}.$$

5.6 Repeated Eigenvalues



Example

Solve

$$\mathbf{x}' = \begin{bmatrix} 1 & -4 \\ 4 & -7 \end{bmatrix} \mathbf{x}, \quad \mathbf{x}(0) = \begin{bmatrix} 3 \\ 2 \end{bmatrix}.$$

The only eigenvalue of the matrix is $r = -3$. The corresponding eigenvector is $\boldsymbol{\xi} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$.

5.6 Repeated Eigenvalues

Example

Solve

$$\mathbf{x}' = \begin{bmatrix} 1 & -4 \\ 4 & -7 \end{bmatrix} \mathbf{x}, \quad \mathbf{x}(0) = \begin{bmatrix} 3 \\ 2 \end{bmatrix}.$$

The only eigenvalue of the matrix is $r = -3$. The corresponding eigenvector is $\boldsymbol{\xi} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$. Therefore one solution of the linear system is

$$\mathbf{x}^{(1)}(t) = \boldsymbol{\xi} e^{rt} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-3t}.$$

5.6 Repeated Eigenvalues



Next we need to find a generalised eigenvector η .

5.6 Repeated Eigenvalues



We calculate that

$$(A - rI)\boldsymbol{\eta} = \boldsymbol{\xi}$$

5.6 Repeated Eigenvalues



We calculate that

$$\begin{bmatrix} 4 & -4 \\ 4 & -4 \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

5.6 Repeated Eigenvalues



We calculate that

$$4\eta_1 - 4\eta_2 = 1$$

5.6 Repeated Eigenvalues



We calculate that

$$-\eta_1 + \eta_2 = -\frac{1}{4}$$

5.6 Repeated Eigenvalues



We calculate that

$$\eta_2 = \eta_1 - \frac{1}{4}.$$

5.6 Repeated Eigenvalues



We calculate that

$$\eta_2 = \eta_1 - \frac{1}{4}.$$

We can choose any vector $\boldsymbol{\eta}$ that satisfies $\eta_2 = \eta_1 - \frac{1}{4}$.

5.6 Repeated Eigenvalues



We calculate that

$$\eta_2 = \eta_1 - \frac{1}{4}.$$

We can choose any vector $\boldsymbol{\eta}$ that satisfies $\eta_2 = \eta_1 - \frac{1}{4}$. Thus we may choose $\boldsymbol{\eta} = \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix}$.

5.6 Repeated Eigenvalues



eigenvector

$$\xi = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

generalised eigenvector

$$\eta = \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix}$$

Therefore

$$\mathbf{x}^{(2)}(t) = \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-3t} + \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix} e^{-3t}.$$

5.6 Repeated Eigenvalues



Hence the general solution to the ODE is

$$\mathbf{x}(t) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-3t} + c_2 \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-3t} + \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix} e^{-3t} \right).$$

5.6 Repeated Eigenvalues



The initial condition gives

$$\begin{bmatrix} 3 \\ 2 \end{bmatrix} = \mathbf{x}(0) = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix}$$

which implies that $c_1 = 3$ and $c_2 = 4$.

5.6 Repeated Eigenvalues



Therefore the solution to the IVP is

$$\mathbf{x}(t) = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-3t} + 4 \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-3t} + \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix} e^{-3t} \right)$$

=

=

.

5.6 Repeated Eigenvalues



Therefore the solution to the IVP is

$$\begin{aligned}\mathbf{x}(t) &= 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-3t} + 4 \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-3t} + \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix} e^{-3t} \right) \\ &= \begin{bmatrix} 3 \\ 2 \end{bmatrix} e^{-3t} + 4 \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-3t} \\ &= .\end{aligned}$$

5.6 Repeated Eigenvalues



Therefore the solution to the IVP is

$$\begin{aligned}\mathbf{x}(t) &= 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-3t} + 4 \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-3t} + \begin{bmatrix} 0 \\ -\frac{1}{4} \end{bmatrix} e^{-3t} \right) \\ &= \begin{bmatrix} 3 \\ 2 \end{bmatrix} e^{-3t} + 4 \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{-3t} \\ &= \begin{bmatrix} 3 + 4t \\ 2 + 4t \end{bmatrix} e^{-3t}.\end{aligned}$$



Next Time

- 5.7 Nonhomogeneous Linear Systems