

# MATH 20D Spring 2023 Lecture 26.

## Solving Systems of Equations Using Eigenvectors.

- 1 Review of Eigenvalues and Eigenvectors
- 2 Solving Systems of Linear Equations

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- Let  $A$  be a 2-by-2 matrix. **Eigenvalues** of  $A$  are the scalars  $\lambda \in \mathbb{C}$  such that

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

where  $I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$  is the 2-by-2 identity matrix.

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$$(A - \sqrt{5}I)\mathbf{v} = \begin{pmatrix} 2 - \sqrt{5} & 1 \\ 1 & -2 - \sqrt{5} \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \mathbf{0}.$$



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So  $\mathbf{v} = \text{col}(1, -(2 - \sqrt{5}))$  is an eigenvector with eigenvalue  $\lambda_1$ .

## Linearly Independent Eigenvectors

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For each of the matrices below. Determine whether  $A$  admits a pair of linearly independent eigenvectors.

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- If  $\mathbf{v}$  is an eigenvector of  $A$  with eigenvalue  $\lambda$  then any non-zero scalar multiple of  $\mathbf{v}$  is an eigenvector of  $A$  with eigenvalue  $\lambda$ . So eigenvectors of  $A$  corresponding to **distinct** eigenvalues are always linearly independent.

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## Review of Constant Coefficient Equations

- Let  $a \neq 0$ ,  $b$ , and  $c$  be constants. If  $y_1(t)$  and  $y_2(t)$  are linearly independent solutions to

$$ay''(t) + by'(t) + cy(t) = 0 \quad (1)$$

then a **general solution** to (1) is

$$y(t) = C_1y_1(t) + C_2y_2(t)$$

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- Indeed  $a\lambda^2 + b\lambda + c\lambda = a \cdot \det(A - \lambda I)$  and the matrix  $A$  can be used to express (1) as the matrix equation

$$\mathbf{x}'(t) = A\mathbf{x}(t) \quad \text{where } \mathbf{x}(t) = \text{col}(y(t), y'(t)).$$

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### Definition

We say that two vector valued function  $\mathbf{x}_1, \mathbf{x}_2: \mathbb{R} \rightarrow \mathbb{C}^2$  are **linearly dependent on**  $(-\infty, \infty)$  if there exists a scalar  $\alpha \in \mathbb{C}$  such that either

$$\mathbf{x}_1(t) = \alpha\mathbf{x}_2(t) \quad \text{or} \quad \mathbf{x}_2(t) = \alpha \cdot \mathbf{x}_1(t)$$

for all  $t \in \mathbb{R}$ .

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Suppose  $\mathbf{x}_1(t)$  and  $\mathbf{x}_2(t)$  are linearly independent solutions to (2).



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where  $C_1$  and  $C_2$  are constant.

- To prove the theorem on the previous slide one shows that for any initial value problem

$$\mathbf{x}'(t) = A\mathbf{x}(t), \quad \mathbf{x}(0) = \mathbf{x}_0 \quad (3)$$

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A **fundamental matrix for the system**  $\mathbf{x}'(t) = A\mathbf{x}(t)$  is any matrix of the form

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- Hence the condition  $\mathbf{x}(0) = \mathbf{x}_0$  implies that  $\text{col}(C_1, C_2) = X(0)^{-1}\mathbf{x}_0$ .
- We have  $\det(X(0)) = \text{Wr}[\mathbf{x}_1, \mathbf{x}_2](0) \neq 0$  by a version of Abel's theorem.

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$$\mathbf{x}_1(t) = e^{r_1 t} \mathbf{v}_1 \quad \text{and} \quad \mathbf{x}_2(t) = e^{r_2 t} \mathbf{v}_2$$

*give a pair of linearly independent solutions to (4).*

### Example

Write down a general solution to the equation

$$\mathbf{x}'(t) = \begin{pmatrix} 2 & -3 \\ 1 & -2 \end{pmatrix} \mathbf{x}(t)$$