

# **Lecture 20: Diagonalization; Linear Differential Equations (Sections 5.2--5.4)**

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# Diagonalization

Diagonal matrices are very easy to work with.

► **Example.** For instance, it is easy to compute their powers.

Let's consider  $A = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 4 \end{bmatrix}$ , then

$$A^2 = \begin{bmatrix} 2^2 & & \\ & 3^2 & \\ & & 4^2 \end{bmatrix} \quad \text{and} \quad A^{100} = \begin{bmatrix} 2^{100} & & \\ & 3^{100} & \\ & & 4^{100} \end{bmatrix}.$$

## Diagonalization

► **Example.** If  $A = \begin{bmatrix} 6 & -1 \\ 2 & 3 \end{bmatrix}$ , then  $A^{100} = ?$  Let's find eigenvalues and eigenvectors of  $A$ :

- $\lambda_1 = 4 \Rightarrow$  eigenvector  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ .
- $\lambda_2 = 5 \Rightarrow$  eigenvector  $\mathbf{v}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ .

Key observation: if  $\mathbf{v}$  is an eigenvector of  $A$  corresponding to an eigenvalue  $\lambda$ ,

$$A^m \mathbf{v} = \lambda^m \mathbf{v}.$$

## Diagonalization

Let  $B = A^{100} = [\mathbf{b}_1 \ \mathbf{b}_2]$ . Then  $\mathbf{b}_1 = A^{100} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\mathbf{b}_2 = A^{100} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ .

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Therefore, finding  $A^{100}$  is equivalent to finding  $A^{100} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and

$$A^{100} \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$

Observe that

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} = -\begin{bmatrix} 1 \\ 2 \end{bmatrix} + 2\begin{bmatrix} 1 \\ 1 \end{bmatrix} = -\mathbf{v}_1 + 2\mathbf{v}_2.$$

$$A^{100} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = -4^{100} \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 2 \cdot 5^{100} \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

Exercise: find  $A^{100} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ .

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$$A \begin{bmatrix} | & & | \\ \mathbf{x}_1 & \cdots & \mathbf{x}_n \\ | & & | \end{bmatrix} = \begin{bmatrix} | & & | \\ \lambda_1 \mathbf{x}_1 & \cdots & \lambda_n \mathbf{x}_n \\ | & & | \end{bmatrix}$$
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$$AS = S\Lambda$$



# Diagonalization

Suppose that  $A$  is  $n \times n$  and has independent eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$ . Then  $A$  can be diagonalized as  $A = S\Lambda S^{-1}$ .

- the columns of  $S$  are the eigenvectors
- the diagonal matrix  $\Lambda$  has the eigenvalues on the diagonal.

(Such a diagonalization is possible if and only if  $A$  has enough eigenvectors.)

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Eigenvalues and eigenvectors:

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Then

$$S = \begin{bmatrix} 1 & 1 \\ -i & i \end{bmatrix}, \Lambda = \begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix} \Rightarrow \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -i & i \end{bmatrix} \begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -i & i \end{bmatrix}^{-1}.$$

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Additionally, if  $A$  is invertible,

$$A^{-1} = S\Lambda^{-1} S^{-1}.$$

## Linear differential equations

► **Example.** The **differential equation**  $y' = ay$  with **initial condition**  $y(0) = C$  is solved by  $y(t) = Ce^{at}$ .

► **Example.** Our goal is to solve systems of differential equations:

$$y_1' = 2y_1$$

$$y_1(0) = 1$$

$$y_2' = -y_1 + 3y_2 + y_3$$

$$y_2(0) = 2$$

$$y_3' = -y_1 + y_2 + 3y_3$$

$$y_3(0) = 1.$$

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In matrix form:

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

## Linear differential equations

► **Definition.** Let  $A$  be  $n \times n$ . The **matrix exponential** is

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Then

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The solution to  $\mathbf{y}' = A\mathbf{y}, \mathbf{y}(0) = \mathbf{y}_0$  is

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

## Linear differential equations

► **Example.** If  $A = \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$ , then

$$e^A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix} + \frac{1}{2!} \begin{bmatrix} 2^2 & 0 \\ 0 & 5^2 \end{bmatrix} + \dots = \begin{bmatrix} e^2 & 0 \\ 0 & e^5 \end{bmatrix}$$

$$e^{At} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 2t & 0 \\ 0 & 5t \end{bmatrix} + \frac{1}{2!} \begin{bmatrix} (2t)^2 & 0 \\ 0 & (5t)^2 \end{bmatrix} + \dots = \begin{bmatrix} e^{2t} & 0 \\ 0 & e^{5t} \end{bmatrix}$$

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**Theorem.** Suppose  $A = S\Lambda S^{-1}$ . Then  $e^A = Se^{\Lambda}S^{-1}$ .



## Linear differential equations

► **Example.** Solve the differential equation

$$\mathbf{y}' = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \mathbf{y}, \quad \mathbf{y}(0) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

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Step 1: Diagonalize  $A$

- $\lambda_1 = 1 \Rightarrow \mathbf{x}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$
- $\lambda_2 = -1 \Rightarrow \mathbf{x}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

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Hence,

$$A = S \Lambda S^{-1}, \quad \text{where } S = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \text{ and } \Lambda = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}.$$

## Linear differential equations

Step 2: Compute the solution  $\mathbf{y}(t) = e^{At}\mathbf{y}_0$ .

$$\mathbf{y} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} e^t & 0 \\ 0 & e^{-t} \end{bmatrix} \frac{1}{2} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} e^t + e^{-t} \\ e^t - e^{-t} \end{bmatrix}.$$