

# Math 211

Lecture #25

Exponential Solutions

October 26, 2001

## Homogeneous Systems

- These are systems of the form

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

where  $A$  is an  $n \times n$  matrix.

- We are looking primarily at homogeneous systems with constant coefficients.

## Structure of the Solution Space

**Theorem:** Suppose that  $\mathbf{x}_1(t)$ ,  $\mathbf{x}_2(t)$ ,  $\dots$ , and  $\mathbf{x}_n(t)$  are linearly independent solutions to the  $n \times n$  homogeneous system  $\mathbf{x}' = A\mathbf{x}$  on the interval  $I$ . Then every solution is a linear combination of  $\mathbf{x}_1(t)$ ,  $\mathbf{x}_2(t)$ ,  $\dots$ , and  $\mathbf{x}_n(t)$ .

- That is, if  $\mathbf{x}(t)$  is a solution, then there are constants  $C_1$ ,  $C_2$ ,  $\dots$ , and  $C_n$  such that

$$\mathbf{x}(t) = C_1\mathbf{x}_1(t) + C_2\mathbf{x}_2(t) + \dots + C_n\mathbf{x}_n(t).$$

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## Solution Strategy

- The obvious strategy for completely solving the system is to look for  $n$  linearly independent solutions.

**Definition:** A set of  $n$  linear independent solutions to the  $n \times n$  homogeneous system  $\mathbf{x}' = A\mathbf{x}$  is called a *fundamental set of solutions*.

- We will look for fundamental sets of solutions.

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## Procedure to Solve $\mathbf{x}' = A\mathbf{x}$

- Find the eigenvalues of  $A$ 
  - ♦ the roots of  $p(\lambda) = \det(A - \lambda I) = 0$
- For each eigenvalue  $\lambda$  find the eigenspace
  - ♦  $= \text{null}(A - \lambda I)$
- If  $\lambda$  is an eigenvalue and  $\mathbf{v}$  is an associated eigenvector,  $\mathbf{x}(t) = e^{\lambda t}\mathbf{v}$  is a solution.
- Hope that  $n$  of these are linearly independent.

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## Example

$$\mathbf{x}' = A\mathbf{x} \quad \text{where} \quad A = \begin{pmatrix} -4 & 2 \\ -3 & 1 \end{pmatrix}$$

- $A$  has eigenvalues  $\lambda_1 = -1$  and  $\lambda_2 = -2$ .
- Look for associated eigenvectors.
  - ♦  $\lambda_1 = -1$ . Eigenvector:  $\mathbf{v}_1 = (2, 3)^T$ .
    - ▶ Solution:  $\mathbf{x}_1(t) = e^{\lambda_1 t}\mathbf{v}_1 = e^{-t}(2, 3)^T$ .
  - ♦  $\lambda_2 = -2$ . Eigenvector:  $\mathbf{v}_2 = (1, 1)^T$ .
    - ▶ Solution:  $\mathbf{x}_2(t) = e^{\lambda_2 t}\mathbf{v}_2 = e^{-2t}(1, 1)^T$ .

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- $x_1(0) = \mathbf{v}_1 = (2, 3)^T$  and  $x_2(0) = \mathbf{v}_2 = (1, 1)^T$  are linearly independent.
- $\mathbf{x}_1$  and  $\mathbf{x}_2$  form a fundamental set of solutions.
- The general solution is the set of all linear combinations:

$$\begin{aligned}\mathbf{x}(t) &= C_1\mathbf{x}_1(t) + C_2\mathbf{x}_2(t) \\ &= C_1e^{-t} \begin{pmatrix} 2 \\ 3 \end{pmatrix} + C_2e^{-2t} \begin{pmatrix} 1 \\ 1 \end{pmatrix}\end{aligned}$$

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Example

Procedure

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

Cases to be Considered

- Distinct real eigenvalues.
  - ♦ In this case the method works as described.
- Complex eigenvalues.
  - ♦ The method yields complex solutions, but we will want real solutions.
- Repeated eigenvalues.
  - ♦ The method does not always give enough solutions.
    - ▶ This is the hard case.

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## Planar System $\mathbf{x}' = A\mathbf{x}$

$$A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \quad \text{and} \quad \mathbf{x}(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$$

In nonvector form

$$x_1' = a_{11}x_1 + a_{12}x_2$$

$$x_2' = a_{21}x_1 + a_{22}x_2$$

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Procedure

## The Characteristic Polynomial

$$\begin{aligned}
 p(\lambda) &= \det(A - \lambda I) \\
 &= \det \begin{pmatrix} a_{11} - \lambda & a_{12} \\ a_{21} & a_{22} - \lambda \end{pmatrix} \\
 &= \lambda^2 - (a_{11} + a_{22})\lambda + (a_{11}a_{22} - a_{12}a_{21}) \\
 &= \lambda^2 - T\lambda + D,
 \end{aligned}$$

where

- $D = a_{11}a_{22} - a_{12}a_{21} = \det(A)$
- $T = a_{11} + a_{22} = \text{tr}(A)$  is the *trace* of  $A$ .
  - ♦ The *trace* of a matrix is the sum of its diagonal elements.

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## The Eigenvalues of $A$

- The eigenvalues of  $A$  are the roots of the characteristic equation  $p(\lambda) = \lambda^2 - T\lambda + D = 0$ .

$$\lambda = \frac{T \pm \sqrt{T^2 - 4D}}{2}.$$

- Three cases:
  - ♦ 2 distinct real roots if  $T^2 - 4D > 0$
  - ♦ 2 complex conjugate roots if  $T^2 - 4D < 0$
  - ♦ Double real root if  $T^2 - 4D = 0$

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## Eigenvectors are Linearly Independent

The problem of determining that solutions are linearly independent is eased by the following result.

**Proposition:** Suppose that  $\lambda_1 \neq \lambda_2$  are eigenvalues of the  $n \times n$  matrix  $A$ , and that  $\mathbf{v}_1 \neq \mathbf{0}$  and  $\mathbf{v}_2 \neq \mathbf{0}$  are eigenvectors associated with  $\lambda_1$  and  $\lambda_2$ , respectively. Then  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are linearly independent.

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## Two Distinct Real Eigenvalues

$$\lambda_1 = \frac{T - \sqrt{T^2 - 4D}}{2}, \quad \lambda_2 = \frac{T + \sqrt{T^2 - 4D}}{2}$$

- $T^2 - 4D > 0$  so  $\lambda_1 < \lambda_2$ .
- There are associated nonzero eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$ .
- Solutions  $\mathbf{x}_1(t) = e^{\lambda_1 t} \mathbf{v}_1$  and  $\mathbf{x}_2(t) = e^{\lambda_2 t} \mathbf{v}_2$ .
- $\mathbf{x}_1(0) = \mathbf{v}_1$  and  $\mathbf{x}_2(0) = \mathbf{v}_2$  are linearly independent;  $\mathbf{x}_1(t)$  and  $\mathbf{x}_2(t)$  form a fundamental set of solutions.
- The general solution is  $\mathbf{x}(t) = C_1 e^{\lambda_1 t} \mathbf{v}_1 + C_2 e^{\lambda_2 t} \mathbf{v}_2$ .

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Cases

Procedure

## Example

$$\mathbf{x}' = A\mathbf{x} \quad \text{where} \quad A = \begin{pmatrix} -6 & -8 \\ 4 & 6 \end{pmatrix}$$

- Characteristic polynomial:  $p(\lambda) = \lambda^2 - 4$ .
- Eigenvalues:  $\lambda_1 = -2$  and  $\lambda_2 = 2$ .
  - ♦  $\lambda_1 = -2$ . Eigenvector:  $\mathbf{v}_1 = (-2, 1)^T$ .
    - ▶ Solution:  $\mathbf{x}_1(t) = e^{\lambda_1 t} \mathbf{v}_1 = e^{-2t} (-2, 1)^T$ .
  - ♦  $\lambda_2 = 2$ . Eigenvector:  $\mathbf{v}_2 = (-1, 1)^T$ .
    - ▶ Solution:  $\mathbf{x}_2(t) = e^{\lambda_2 t} \mathbf{v}_2 = e^{2t} (-1, 1)^T$ .

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- $\mathbf{x}_1$  and  $\mathbf{x}_2$  are a fundamental set of solutions.
- The general solution is

$$\begin{aligned} \mathbf{x}(t) &= C_1 \mathbf{x}_1(t) + C_2 \mathbf{x}_2(t) \\ &= C_1 e^{-2t} \begin{pmatrix} -2 \\ 1 \end{pmatrix} + C_2 e^{2t} \begin{pmatrix} -1 \\ 1 \end{pmatrix}. \end{aligned}$$

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Example

## Initial Value Problem

Solve  $\mathbf{x}' = A\mathbf{x}$  with the initial condition  $\mathbf{x}(0) = (1, 4)^T$ .

- We need

$$\mathbf{x}(0) = C_1\mathbf{x}_1(0) + C_2\mathbf{x}_2(0)$$

- $C_1 = -5$  and  $C_2 = 9$ .

- The solution is

$$\begin{aligned}\mathbf{x}(t) &= -5\mathbf{x}_1(t) + 9\mathbf{x}_2(t) \\ &= \begin{pmatrix} 10e^{-2t} - 9e^{2t} \\ -5e^{-2t} + 9e^{2t} \end{pmatrix}.\end{aligned}$$

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IVP

## Homogeneous Systems

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

**Proposition:** Suppose that  $\mathbf{x}_1(t)$ ,  $\mathbf{x}_2(t)$ ,  $\dots$ , and  $\mathbf{x}_k(t)$  are solutions to the homogeneous system, and  $c_1, c_2, \dots$ , and  $c_k$  are scalars. Then

$$\mathbf{x}(t) = c_1\mathbf{x}_1(t) + c_2\mathbf{x}_2(t) + \dots + c_k\mathbf{x}_k(t)$$

is also a solution.

- Any linear combination of solutions to the homogeneous system is also a solution.

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## Linear Independence

**Definition:** A set of  $k$  solutions to the linear system  $\mathbf{x}' = A\mathbf{x}$  is *linearly independent* if they are linearly independent at one value of  $t$ .

- Proposition  $\Rightarrow$  the solutions are linearly independent for all values of  $t$ .

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## Eigenvalues & Eigenvectors

**Definition:**  $\lambda$  is an *eigenvalue* of  $A$  if there is a nonzero vector  $\mathbf{v}$  such that  $A\mathbf{v} = \lambda\mathbf{v}$ . If  $\lambda$  is an eigenvalue of  $A$ , then any vector  $\mathbf{v}$  such that  $A\mathbf{v} = \lambda\mathbf{v}$  is called an *eigenvector associated with  $\lambda$* .

- $\lambda$  an eigenvalue of  $A$ ,  $\mathbf{v}$  an associated eigenvector  
 $\Rightarrow \mathbf{x}(t) = e^{\lambda t}\mathbf{v}$  is a solution to  $\mathbf{x}' = A\mathbf{x}$ .
- The set of eigenvectors associated with the eigenvalue  $\lambda$  is the subspace  $\text{null}(A - \lambda I)$ , and is called the eigenspace of  $\lambda$ .

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