

# Math 211

Lecture #17

October 26, 2000

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

- Homogeneous linear system.
- Assume the system has constant coefficients, so  $A$  is a constant matrix.

- Example  $A = \begin{pmatrix} -4 & 2 \\ -3 & 1 \end{pmatrix}$

- or 
$$\begin{aligned} x_1' &= -4x_1 + 2x_2 \\ x_2' &= -3x_1 + x_2 \end{aligned}$$

## Solution Strategy

**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.

- The obvious strategy for completely solving the system is to look for  $n$  linearly independent solutions — a fundamental set of solutions.

## 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 to the system 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

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

## 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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$$D = 1$$

- One equation:  $x' = ax$ 
  - ◇  $a$  is a constant.
- Solution:  $x(t) = Ce^{at}$
- Solutions are exponentials. Can we find exponential solutions to a system of equations?

## Exponential Solutions to $\mathbf{x}' = A\mathbf{x}$

- Can we find a solution of the form  $\mathbf{x}(t) = e^{\lambda t}\mathbf{v}$ , where  $\mathbf{v}$  is a vector with constant entries?
- $\mathbf{x}' = \lambda e^{\lambda t}\mathbf{v}$
- $A\mathbf{x} = e^{\lambda t}A\mathbf{v}$
- $\mathbf{x}' = A\mathbf{x} \Leftrightarrow A\mathbf{v} = \lambda\mathbf{v}$ .
- If  $A\mathbf{v} = \lambda\mathbf{v}$  then  $\mathbf{x}(t) = e^{\lambda t}\mathbf{v}$  is a solution.
- Can we find  $\lambda$  and  $\mathbf{v}$  such that  $A\mathbf{v} = \lambda\mathbf{v}$ ?

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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  $\lambda$ .

- If  $\lambda$  is an eigenvalue of  $A$ , then  $\mathbf{x}(t) = e^{\lambda t}\mathbf{v}$  is a solution to  $\mathbf{x}' = A\mathbf{x}$  for any associated eigenvector  $\mathbf{v}$ .
- How do we find eigenvalues and eigenvectors?

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## Finding Eigenvalues

$\lambda$  is an eigenvalue of  $A$  if there is a vector  $\mathbf{v} \neq \mathbf{0}$  such that  $A\mathbf{v} = \lambda\mathbf{v}$ .

$\Leftrightarrow \mathbf{v} \neq \mathbf{0}$  and

$$\begin{aligned} \mathbf{0} &= A\mathbf{v} - \lambda\mathbf{v} \\ &= A\mathbf{v} - \lambda I\mathbf{v} \\ &= (A - \lambda I)\mathbf{v} \end{aligned}$$

$\Leftrightarrow A - \lambda I$  has a nontrivial nullspace.

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

Definition

return

### Example

$$A = \begin{pmatrix} -4 & 2 \\ -3 & 1 \end{pmatrix}$$

$$A - \lambda I = \begin{pmatrix} -4 - \lambda & 2 \\ -3 & 1 - \lambda \end{pmatrix}$$

$$\det(A - \lambda I) = (-4 - \lambda)(1 - \lambda) + 6$$

$$= \lambda^2 + 3\lambda + 2$$

$$= (\lambda + 1)(\lambda + 2)$$

- $A$  has eigenvalues  $\lambda_1 = -1$  and  $\lambda_2 = -2$ .

return

### Characteristic Polynomial of $A$

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{pmatrix}$$

$$A - \lambda I = \begin{pmatrix} a_{11} - \lambda & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} - \lambda & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} - \lambda \end{pmatrix}$$

return

### Characteristic Polynomial of $A$

- The characteristic polynomial of  $A$  is  $p(\lambda) = \det(A - \lambda I)$ .
- If  $A$  is an  $n \times n$  matrix  $p(\lambda)$  is a polynomial of degree  $n$ .
- Each root of  $p(\lambda) = 0$  is an eigenvalue of  $A$ .
- We find the eigenvalues of  $A$  by finding the roots of the characteristic equation  $p(\lambda) = 0$ .
- Usually  $p(\lambda) = 0$  has  $n$  roots. Usually  $A$  has  $n$  eigenvalues.

Example

Matrix

## Finding Eigenvectors

$\mathbf{v}$  is an eigenvector associated with  $\lambda$  if

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

$$\Leftrightarrow (A - \lambda I)\mathbf{v} = \mathbf{0}.$$

$$\Leftrightarrow \mathbf{v} \in \text{null}(A - \lambda I).$$

- The set of all eigenvectors associated to the eigenvalue  $\lambda$  is equal to the nullspace of  $A - \lambda I$ . It is therefore a subspace of  $\mathbf{R}^n$ , called the eigenspace of  $\lambda$ .

Eigenvalues

Definition

return

## Example

$A = \begin{pmatrix} -4 & 2 \\ -3 & 1 \end{pmatrix}$  has eigenvalues  $\lambda_1 = -1$  and  $\lambda_2 = -2$ .

- $\lambda_1 = -1$

$$A - \lambda_1 I = \begin{pmatrix} -4+1 & 2 \\ -3 & 1+1 \end{pmatrix} = \begin{pmatrix} -3 & 2 \\ -3 & 2 \end{pmatrix}$$

$\mathbf{v}_1 = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$  is an eigenvector;

$\mathbf{x}_1(t) = e^{\lambda_1 t} \mathbf{v}_1 = e^{-t} \begin{pmatrix} 2 \\ 3 \end{pmatrix}$  is a solution.

Example

return

## Example (cont.)

- $\lambda_2 = -2$

$$A - \lambda_2 I = \begin{pmatrix} -4+2 & 2 \\ -3 & 1+2 \end{pmatrix} = \begin{pmatrix} -2 & 2 \\ -3 & 3 \end{pmatrix}$$

$\mathbf{v}_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  is an eigenvector;

$\mathbf{x}_2(t) = e^{\lambda_2 t} \mathbf{v}_2 = e^{-2t} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  is a solution.

Example

return

### Example (cont.)

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

has solutions

$$\mathbf{x}_1(t) = e^{-t} \begin{pmatrix} 2 \\ 3 \end{pmatrix} \quad \text{and} \quad \mathbf{x}_2(t) = e^{-2t} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$\mathbf{x}_1(0) = \mathbf{v}_1$  and  $\mathbf{x}_2(0) = \mathbf{v}_2$  are linearly independent. They form a fundamental set of solutions.

[Example1](#)

[Example2](#)

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[return](#)

### Example (cont.)

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

- The general solution is the set of all linear combinations:

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

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

- Find the eigenvalues of  $A$ 
  - ◊ the roots of  $\det(A - \lambda I) = 0$
- 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.
- Show that  $n$  of these are linearly independent.

[Eigenvalues](#)

[Eigenvector1](#)

[Eigenvector2](#)

[return](#)

## Cases

- Distinct real eigenvalues.
  - ◊ In this case the method works as described.
- Complex eigenvalues.
  - ◊ The method yields complex solutions.
- Repeated eigenvalues.
  - ◊ The method does not always give enough solutions.

## 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

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

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

Procedure

return

## Characteristic Polynomial

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

System

return

### Cont.

- Set  $D = \det(A) = a_{11}a_{22} - a_{12}a_{21}$
- The trace of  $A$  is  $\text{tr}(A) = a_{11} + a_{22}$ .  
Set  $T = \text{tr}(A)$ .

- Then

$$p(\lambda) = \lambda^2 - T\lambda + D.$$

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### Eigenvalues of $A$

- Roots of  $p(\lambda) = \lambda^2 - T\lambda + D = 0$ .

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

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

return

### Eigenvectors

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

**Proposition:**  $\lambda_1 \neq \lambda_2$  eigenvalues of  $A$ .

$\mathbf{v}_1 \neq 0$  and  $\mathbf{v}_2 \neq 0$  eigenvectors associated with  $\lambda_1$  and  $\lambda_2$ , resp. 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  $\lambda_1 < \lambda_2$ .
- Associated nonzero eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$ .
- Solutions  $\mathbf{x}_1(t) = e^{\lambda_1 t} \mathbf{v}_1$  and  $\mathbf{x}_2(t) = e^{\lambda_2 t} \mathbf{v}_2$ .
- $\lambda_1 \neq \lambda_2 \Rightarrow \mathbf{x}_1(0) = \mathbf{v}_1$  and  $\mathbf{x}_2(0) = \mathbf{v}_2$  are linearly independent.

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Cases

return

## Two Distinct Real Eigenvalues

If  $A$  is a  $2 \times 2$  matrix with

- two real eigenvalues  $\lambda_1 \neq \lambda_2$ , and
- associated nonzero eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$

the general solution is

$$\mathbf{x}(t) = C_1 e^{\lambda_1 t} \mathbf{v}_1 + C_2 e^{\lambda_2 t} \mathbf{v}_2.$$

back

return

## Example

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

- $p(\lambda) = \lambda^2 - 4$ . Eigenvalues:  $-2$  and  $2$ .
- Eigenvectors:  $\begin{pmatrix} -2 \\ 1 \end{pmatrix}$  and  $\begin{pmatrix} -1 \\ 1 \end{pmatrix}$
- Fundamental set of solutions:

$$\mathbf{x}_1(t) = e^{-2t} \begin{pmatrix} -2 \\ 1 \end{pmatrix} \quad \text{and} \quad \mathbf{x}_2(t) = e^{2t} \begin{pmatrix} -1 \\ 1 \end{pmatrix}.$$

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## Complex Eigenvalues

$$\lambda = \frac{T + i\sqrt{4D - T^2}}{2}, \quad \bar{\lambda} = \frac{T - i\sqrt{4D - T^2}}{2}$$

Example:  $\begin{pmatrix} -5 & 20 \\ -2 & 7 \end{pmatrix}$

- $p(\lambda) = \lambda^2 - 2\lambda + 5$ .
- Eigenvalue:  $\lambda = 1 + 2i$
- Eigenvector:  $\mathbf{w} = \begin{pmatrix} 3 - i \\ 1 \end{pmatrix}$

Eigenvalues

## Example

- Solution

$$\begin{aligned} \mathbf{z}(t) &= e^{\lambda t} \mathbf{w} \\ &= e^{(1+2i)t} \begin{pmatrix} 3 - i \\ 1 \end{pmatrix} \end{aligned}$$

- What is  $e^{(1+2i)t}$ ?

## Euler's Formula

- Define

$$e^{iy} = \cos y + i \sin y$$

- Define

$$\begin{aligned} e^{x+iy} &= e^x e^{iy} \\ &= e^x [\cos y + i \sin y]. \end{aligned}$$

- Then

$$e^{(1+2i)t} = e^t [\cos 2t + i \sin 2t]$$

## Example

- Solution

$$\begin{aligned}\mathbf{z}(t) &= e^{\lambda t} \mathbf{w} \\ &= e^{(1+2i)t} \begin{pmatrix} 3-i \\ 1 \end{pmatrix}\end{aligned}$$

- Solution is complex valued.
- Solution corresponding to  $\bar{\lambda}$  is

$$\bar{\mathbf{z}}(t) = e^{(1-2i)t} \begin{pmatrix} 3+i \\ 1 \end{pmatrix}$$

## Complex Conjugate Eigenvalues

If  $A$  is a  $2 \times 2$  matrix with

- complex eigenvalues  $\lambda$  and  $\bar{\lambda}$ , and
- associated nonzero eigenvectors  $\mathbf{w}$  and  $\bar{\mathbf{w}}$

the general complex solution is

$$C_1 e^{\lambda t} \mathbf{w} + C_2 e^{\bar{\lambda} t} \bar{\mathbf{w}}.$$

- We want real solutions.

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## Real Solutions

We have solutions

$$\mathbf{z}(t) = \mathbf{x}(t) + i\mathbf{y}(t) \quad \text{and} \quad \bar{\mathbf{z}}(t) = \mathbf{x}(t) - i\mathbf{y}(t)$$

Thus

$$\mathbf{x}(t) = \frac{1}{2}(\mathbf{z}(t) + \bar{\mathbf{z}}(t))$$

$$\mathbf{y}(t) = \frac{1}{2i}(\mathbf{z}(t) - \bar{\mathbf{z}}(t))$$

are also solutions, and they are real valued.

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