

# Math 211

Lecture #28

Higher Dimensional Systems

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# Planar Systems

$$A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

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

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

- $\lambda_1$  &  $\lambda_2$  are the roots of  $p(\lambda)$ , so

$$\begin{aligned} p(\lambda) &= \lambda^2 - T\lambda + D \\ &= (\lambda - \lambda_1)(\lambda - \lambda_2) \\ &= \lambda^2 - (\lambda_1 + \lambda_2)\lambda + \lambda_1\lambda_2 \end{aligned}$$

- $T = \lambda_1 + \lambda_2$  and  $D = \lambda_1\lambda_2$ .
- Duality between  $(\lambda_1, \lambda_2)$  and  $(T, D)$ .
- Represent systems by location of  $(T, D)$  in the  $TD$ -plane.

## Trace-Determinant Plane

- $T^2 - 4D > 0$ 
  - ◇  $\Rightarrow$  distinct real eigenvalues  $\lambda_1$  &  $\lambda_2$
  - ◇  $D = \lambda_1 \lambda_2 < 0 \Rightarrow$  Saddle point.
  - ◇  $D = \lambda_1 \lambda_2 > 0 \Rightarrow$  Eigenvalues have the same sign.
    - ★  $T = \lambda_1 + \lambda_2 > 0 \Rightarrow$  Nodal source.
    - ★  $T = \lambda_1 + \lambda_2 < 0 \Rightarrow$  Nodal sink.

- $T^2 - 4D < 0 \Rightarrow$  complex eigenvalues

$$\lambda = \alpha + i\beta \quad \text{and} \quad \bar{\lambda} = \alpha - i\beta.$$

- ◇  $T = \lambda + \bar{\lambda} = 2\alpha > 0 \Rightarrow$  Spiral source.
- ◇  $T = \lambda + \bar{\lambda} = 2\alpha < 0 \Rightarrow$  Spiral sink.
- ◇  $T = \lambda + \bar{\lambda} = 2\alpha = 0 \Rightarrow$  Center.

# Types of Equilibrium Points

- *Generic* types
  - ◇ Saddle, nodal source, nodal sink, spiral source, and spiral sink.
  - ◇ All occupy large open subsets of the trace-determinant plane.
- *Nongeneric* types
  - ◇ Center and eight others. Occupy pieces of the boundaries.

# Higher Dimensional Systems

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

- $A$  is a real  $n \times n$  matrix.
- If  $\lambda$  is an eigenvalue and  $\mathbf{v} \neq 0$  is an associated eigenvector, then  $\mathbf{x}(t) = e^{\lambda t} \mathbf{v}$  is a solution.
- Much like the planar case, but now we need  $n$  linearly independent solutions.

**Proposition:** Suppose that  $\lambda_1, \dots, \lambda_k$  are distinct eigenvalues of  $A$ , and that  $\mathbf{v}_1, \dots, \mathbf{v}_k$  are associated nonzero eigenvectors. Then  $\mathbf{v}_1, \dots, \mathbf{v}_k$  are linearly independent.

**Theorem:** Suppose the  $n \times n$  real matrix  $A$  has  $n$  distinct eigenvalues  $\lambda_1, \dots, \lambda_n$ , and that  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are associated nonzero eigenvectors. Then the exponential solutions  $\mathbf{x}_i(t) = e^{\lambda_i t} \mathbf{v}_i$ ,  $1 \leq i \leq n$  form a fundamental set of solutions for the system  $\mathbf{x}' = A\mathbf{x}$ .

## Example

$$A = \begin{pmatrix} 17 & -30 & -8 \\ 16 & -29 & -8 \\ -12 & 24 & 7 \end{pmatrix}$$

- Use MATLAB.

# Complex Eigenvalues

A real  $n \times n$  matrix with a complex eigenvalue  $\lambda$  and associate eigenvector  $\mathbf{w}$ .

- $\Rightarrow \bar{\lambda}$  is an eigenvalue and  $\bar{\mathbf{w}}$  is an associated nonzero eigenvector.
- Complex valued solutions:  $\mathbf{z}(t) = e^{\lambda t} \mathbf{w}$   
 $\bar{\mathbf{z}}(t) = e^{\bar{\lambda} t} \bar{\mathbf{w}}.$
- Real solutions:  $\mathbf{x}(t) = \text{Re}(\mathbf{z}(t))$   
 $\mathbf{y}(t) = \text{Im}(\mathbf{z}(t)).$

## Example

$$A = \begin{pmatrix} 21 & 10 & 4 \\ -70 & -31 & -10 \\ 30 & 10 & -1 \end{pmatrix}$$

- The **theorem** applies if some of the eigenvalues are complex and we replace complex conjugate pairs of solutions by their real and imaginary parts.

# Repeated Eigenvalues – Example 1

$$A = \begin{pmatrix} -5 & -10 & 6 \\ 8 & 19 & -12 \\ 12 & 30 & -19 \end{pmatrix}$$

- $p(\lambda) = (\lambda + 3)(\lambda + 1)^2$
- $\lambda_1 = -3$ 
  - ◇ Eigenspace has dimension 1  $\Rightarrow$  one exponential solution

$$\mathbf{x}_1(t) = e^{-3t} (-1/3, 2/3, 1)^T$$

- $\lambda_2 = -1$ 
  - ◇ Eigenspace has dimension 2  $\Rightarrow$  two linearly independent exponential solutions

$$\mathbf{x}_2(t) = e^{-t} \begin{pmatrix} -5/2 \\ 1 \\ 0 \end{pmatrix} \quad \& \quad \mathbf{x}_3(t) = e^{-t} \begin{pmatrix} 3/2 \\ 0 \\ 1 \end{pmatrix}$$

- $\mathbf{x}_1$ ,  $\mathbf{x}_2$ , and  $\mathbf{x}_3$  are a fundamental set of solutions.

## Repeated Eigenvalues – Example 2

$$A = \begin{pmatrix} 1 & 2 & -1 \\ -4 & -7 & 4 \\ -4 & -4 & 1 \end{pmatrix}$$

- $p(\lambda) = (\lambda + 3)(\lambda + 1)^2$
- $\lambda_1 = -3$ 
  - ◇ Eigenspace has dimension 1  $\Rightarrow$  one exponential solution

$$\mathbf{x}_1(t) = e^{-3t} (-1/2, 3/2, 1)^T$$

- $\lambda_2 = -1$ 
  - ◇ Eigenspace has dimension 1  $\Rightarrow$  only one exponential solution

$$\mathbf{x}_2(t) = e^{-t} \begin{pmatrix} -1/2 \\ 1 \\ 1 \end{pmatrix}$$

- Need a third solution.
- Need a new idea.

# Multiplicities

$A$  an  $n \times n$  matrix

- Distinct eigenvalues  $\lambda_1, \dots, \lambda_k$ .
- The characteristic polynomial is

$$p(\lambda) = (\lambda - \lambda_1)^{q_1} (\lambda - \lambda_2)^{q_2} \cdot \dots \cdot (\lambda - \lambda_k)^{q_k}.$$

- The *algebraic multiplicity* of  $\lambda_j$  is  $q_j$ .
- The *geometric multiplicity* of  $\lambda_j$  is  $d_j$ , the dimension of the eigenspace of  $\lambda_j$ .

- We always have:
  - ◇  $q_1 + q_2 + \cdots + q_k = n$ .
  - ◇  $1 \leq d_j \leq q_j$ .
  - ◇ There are  $d_j$  linearly independent exponential solutions corresponding to  $\lambda_j$ .
  - ◇ If  $d_j = q_j$  for all  $j$  we have  $n$  linearly independent solutions.

## Examples

- In both  $p(\lambda) = (\lambda + 1)^2(\lambda + 3)$ .
- In both  $\lambda_2 = -1$  has algebraic multiplicity 2.
- In **Ex. 1**  $\lambda_2 = -1$  has **geom. multiplicity 2**.
- In **Ex. 2**  $\lambda_2 = -1$  has geom. multiplicity 1.
- Problems arise when  $d_j < q_j$ .

## New Approach

- $D = 1 : x' = ax$ 
  - ◇ Solution  $x(t) = Ce^{at}$ .
- $D > 1 : \mathbf{x}' = A\mathbf{x}$ 
  - ◇ Tried  $\mathbf{x}(t) = e^{\lambda t}\mathbf{v}$ .
  - ◇ Why not  $\mathbf{x}(t) = e^{tA}\mathbf{v}$ ?
- But what is  $e^{tA}$ ?

## Exponential of a Matrix

**Definition:** The *exponential* of the  $n \times n$  matrix  $A$  is the  $n \times n$  matrix

$$\begin{aligned} e^A &= I + A + \frac{1}{2!}A^2 + \frac{1}{3!}A^3 + \dots \\ &= \sum_0^{\infty} \frac{1}{n!}A^n. \end{aligned}$$

## Examples

- $A = \begin{pmatrix} r_1 & 0 \\ 0 & r_2 \end{pmatrix}$

$$e^A = \begin{pmatrix} e^{r_1} & 0 \\ 0 & e^{r_2} \end{pmatrix}.$$

- $e^{\lambda I} = e^\lambda I.$

- $e^{0I} = I.$

# Properties

- $A$  commutes with  $e^A$ ,

$$Ae^A = e^A A.$$

- If  $A$  and  $B$  commute (i.e.,  $AB = BA$ ), then

$$e^{A+B} = e^A \cdot e^B.$$

- The inverse of  $e^A$  is  $e^{-A}$ .

- $\frac{d}{dt} e^{tA} = A e^{tA}$ .

## Important Fact

The solution to the initial value problem

$$\mathbf{x}' = A\mathbf{x} \quad \text{with} \quad \mathbf{x}(0) = \mathbf{v}$$

is

$$\mathbf{x}(t) = e^{tA}\mathbf{v}.$$

- However computing  $e^{tA}$  is not easy.