

# Math 211

Lecture #27

Degenerate Planar Systems  
Trace-Determinant Plane

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

- Find the eigenvalues of  $A$ , which are the roots of  $\det(A - \lambda I) = 0$ .
- For each eigenvalue  $\lambda$  find the eigenspace, which is equal to  $\text{null}(A - \lambda I)$ .
- If  $\lambda$  is an eigenvalue and  $\mathbf{v}$  is an associated nonzero eigenvector,  $\mathbf{x}(t) = e^{\lambda t}\mathbf{v}$  is a solution.
- Show that  $n$  of these are linearly independent, *if we can*.
  - ♦ This must be explored further.

Return

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

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

- The characteristic polynomial is

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

where

- ♦  $T = \text{tr } A = a_{11} + a_{22}$  and
- ♦  $D = \det A = a_{11}a_{22} - a_{12}a_{21}$ .

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Procedure

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

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

- 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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### Real Distinct Eigenvalues

Suppose  $A$  is a real  $2 \times 2$  matrix with real eigenvalues  $\lambda_1 \neq \lambda_2$ , and associated nonzero eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$ .

- $\mathbf{x}_1(t) = e^{\lambda_1 t} \mathbf{v}_1$  and  $\mathbf{x}_2(t) = e^{\lambda_2 t} \mathbf{v}_2$  form a fundamental set of solutions.

Return

Cases

### Complex Eigenvalues

Suppose  $A$  is a real  $2 \times 2$  matrix with complex conjugate eigenvalues  $\lambda$  and  $\bar{\lambda}$ , and associated nonzero eigenvectors  $\mathbf{w}$  and  $\bar{\mathbf{w}}$ .

- $\mathbf{z}(t) = e^{\lambda t} \mathbf{w}$  and  $\bar{\mathbf{z}}(t) = e^{\bar{\lambda} t} \bar{\mathbf{w}}$  form a complex valued fundamental set of solutions, and
- $\mathbf{x}(t) = \operatorname{Re}(\mathbf{z}(t))$  and  $\mathbf{y}(t) = \operatorname{Im}(\mathbf{z}(t))$  form a real valued fundamental set of solutions.

Return

Cases

### Examples

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

1.  $A = \begin{pmatrix} -5 & 20 \\ -2 & 7 \end{pmatrix}$ .

2.  $A = \begin{pmatrix} -4 & 10 \\ -2 & 4 \end{pmatrix}$

3.  $A = \begin{pmatrix} 7 & 30 \\ -3 & -11 \end{pmatrix}$

Complex eigenvalues

### Double Real Root

In this case  $T^2 - 4D = 0$ .

- There is only one eigenvalue

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

- The eigenspace of  $\lambda$  has dimension 1 or 2.
  - ♦ If the dimension is 2, then  $A = \lambda I$ .
  - ♦ Every vector is an eigenvector. Every solution has the form

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

Return

Cases

### Example 4

$$\mathbf{x}' = A\mathbf{x} \quad \text{where} \quad A = \begin{pmatrix} 1 & 9 \\ -1 & -5 \end{pmatrix}$$

- $p(\lambda) = \lambda^2 + 4\lambda + 4 = (\lambda + 2)^2$ ;  $\lambda = -2$
- $A - \lambda I = \begin{pmatrix} 3 & 9 \\ -1 & -3 \end{pmatrix}$ ;  $\mathbf{v}_1 = \begin{pmatrix} -3 \\ 1 \end{pmatrix}$
- The eigenspace has dimension 1, with basis  $\mathbf{v}_1$ .
- The standard procedure yields only one solution,  $\mathbf{x}_1(t) = e^{tA} \mathbf{v}_1 = e^{\lambda t} \mathbf{v}_1 = e^{-2t} (-3, 1)^T$ .

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### $e^{tA}$ for Example 4

- Motivated by the Proposition, we compute

$$[A - \lambda I]^2 = \begin{pmatrix} 3 & 9 \\ -1 & -3 \end{pmatrix} \begin{pmatrix} 3 & 9 \\ -1 & -3 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}.$$

- Hence

$$\begin{aligned} e^{tA} &= e^{\lambda t} e^{t[A - \lambda I]} \\ &= e^{\lambda t} \left( I + t[A - \lambda I] + \frac{1}{2}[A - \lambda I]^2 + \dots \right) \\ &= e^{\lambda t} (I + t[A - \lambda I]) \\ &= e^{-2t} \begin{pmatrix} 1 + 3t & 9t \\ -t & 1 - 3t \end{pmatrix} \end{aligned}$$

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### The Degenerate Planar Case

- Find the (only) eigenvalue  $\lambda$ .
- $[A - \lambda I]^2 = 0I$ , so  $e^{tA} = e^{\lambda t} (I + t[A - \lambda I])$ .
  - The solution to the initial value problem  $\mathbf{x}' = A\mathbf{x}$  with  $\mathbf{x} = \mathbf{x}_0$  is  $\mathbf{x}(t) = e^{tA}\mathbf{x}_0$ .
- To get a fundamental set of solutions solve the IVP with any two linearly independent initial vectors.
  - The column vectors in  $e^{tA}$ .
  - $\mathbf{x}_1(t) = e^{\lambda t}\mathbf{v}_1$  and  $\mathbf{x}_2(t) = e^{\lambda t}[\mathbf{v}_2 + t\mathbf{v}_1]$ , where  $\mathbf{v}_1$  is a nonzero eigenvector, and  $\mathbf{v}_2$  is chosen so that  $[A - \lambda I]\mathbf{v}_2 = \mathbf{v}_1$ .

Return

Cases

### Degenerate Examples

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

5.  $A = \begin{pmatrix} -2 & 1 \\ 0 & -2 \end{pmatrix}$

6.  $A = \begin{pmatrix} 0 & 9 \\ -1 & -6 \end{pmatrix}$

Procedure

### Qualitative Analysis of Planar Systems

- Equilibrium points for the system  $\mathbf{x}' = A\mathbf{x}$ .
  - ♦ Set of equilibrium points equals  $\text{null}(A)$ .
  - ♦ If  $A$  is nonsingular the only equilibrium point is  $\mathbf{0}$ .
- Can we list the types of all possible equilibrium points for planar linear systems?
  - ♦ Six most important cases.
  - ♦ Look at solution curves in the phase plane.

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### Distinct Real Eigenvalues

- $p(\lambda) = \lambda^2 - T\lambda + D$  with  $T^2 - 4D > 0$ .

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

- Eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$ . The general solution is

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

- There are three subcases depending on the signs of the eigenvalues.

Return

Cases

### Exponential Solutions

$$\mathbf{x}(t) = C e^{\lambda t} \mathbf{v}$$

- The solution curve is a straight half-line through  $C\mathbf{v}$ .  
Sometimes called *half-line* solutions.
- If  $\lambda > 0$  the solution starts at  $\mathbf{0}$  for  $t = -\infty$ , and tends to  $\infty$  as  $t \rightarrow \infty$ . *Unstable solution*
- If  $\lambda < 0$  the solution starts at  $\infty$  for  $t = -\infty$ , and tends to  $\mathbf{0}$  as  $t \rightarrow \infty$ . *Stable solution*

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Real case

### Saddle Point

- $\lambda_1 < 0 < \lambda_2$
- General solution  $\mathbf{x}(t) = C_1 e^{\lambda_1 t} \mathbf{v}_1 + C_2 e^{\lambda_2 t} \mathbf{v}_2$
- Two stable exponential solutions ( $C_2 = 0$ )
- Two unstable exponential solutions ( $C_1 = 0$ ).
- $C_1 \neq 0$  and  $C_2 \neq 0$ .
  - ♦ As  $t \rightarrow \infty$ ,  $\mathbf{x}(t) \rightarrow \infty$ , approaching the half-line through  $C_2 \mathbf{v}_2$ .
  - ♦ As  $t \rightarrow -\infty$ ,  $\mathbf{x}(t) \rightarrow \infty$ , approaching the half-line through  $C_2 \mathbf{v}_1$ .

Return

Real eigenvalues

### Nodal Sink

- $\lambda_1 < \lambda_2 < 0$
- General solution  $\mathbf{x}(t) = C_1 e^{\lambda_1 t} \mathbf{v}_1 + C_2 e^{\lambda_2 t} \mathbf{v}_2$
- Four stable exponential solutions.
- All solutions  $\rightarrow \mathbf{0}$  as  $t \rightarrow \infty$ . (Stable)
  - ♦ Tangent to  $C_2 \mathbf{v}_2$  if  $C_2 \neq 0$ .
- All solutions  $\rightarrow \infty$  as  $t \rightarrow -\infty$ .
  - ♦  $\parallel$  to the half line through  $C_1 \mathbf{v}_1$  if  $C_1 \neq 0$ .

Return

Real eigenvalues

### Nodal Source

- $0 < \lambda_1 < \lambda_2$
- General solution  $\mathbf{x}(t) = C_1 e^{\lambda_1 t} \mathbf{v}_1 + C_2 e^{\lambda_2 t} \mathbf{v}_2$
- Four unstable exponential solutions.
- All solutions  $\rightarrow \mathbf{0}$  as  $t \rightarrow -\infty$ .
  - ♦ Tangent to  $C_1 \mathbf{v}_1$  if  $C_1 \neq 0$ .
- All solutions  $\rightarrow \infty$  as  $t \rightarrow \infty$ . (Unstable)
  - ♦  $\parallel$  to the half line through  $C_2 \mathbf{v}_2$  if  $C_2 \neq 0$ .

Return

Real eigenvalues

### 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  $\mathbf{v}_1 \neq 0$  and  $\mathbf{v}_2 \neq 0$  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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### Proposition

**Proposition:** Suppose that  $A$  is an  $n \times n$  matrix,  $\lambda$  is a number, and  $\mathbf{v}$  is a vector.

1. If  $[A - \lambda I]\mathbf{v} = \mathbf{0}$ , then  $e^{tA}\mathbf{v} = e^{\lambda t}\mathbf{v}$ .
2. If  $[A - \lambda I]^2\mathbf{v} = \mathbf{0}$ , then  $e^{tA}\mathbf{v} = e^{\lambda t}(\mathbf{v} + t[A - \lambda I]\mathbf{v})$ .
3. If  $[A - \lambda I]^k\mathbf{v} = \mathbf{0}$ , then

$$e^{tA}\mathbf{v} = e^{\lambda t} \left( \mathbf{v} + t[A - \lambda I]\mathbf{v} + \frac{t^2}{2!}[A - \lambda I]^2\mathbf{v} + \cdots + \frac{t^{k-1}}{(k-1)!}[A - \lambda I]^{k-1}\mathbf{v} \right).$$

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