

# Math 211

Lecture #31

Stability of Solutions  
Higher Order Equations

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# Exponential of a Matrix

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

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

**Theorem:** The solution to the initial value problem

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

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

## Computing $e^{tA}\mathbf{v}$

- If  $\lambda$  is an eigenvalue and  $\mathbf{v}$  is an associated eigenvector, then  $e^{tA}\mathbf{v} = e^{\lambda t}\mathbf{v}$ .
- If  $(A - \lambda I)^p\mathbf{v} = \mathbf{0}$  for some integer  $p \geq 1$ , then

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

# Generalized Eigenvectors

**Definition:** If  $\lambda$  is an eigenvalue of  $A$  and  $(A - \lambda I)^p \mathbf{v} = \mathbf{0}$  for some integer  $p \geq 1$ , then  $\mathbf{v}$  is called a *generalized eigenvector* associated with  $\lambda$ .

- We can compute  $e^{tA} \mathbf{v}$  for all such  $\mathbf{v}$ .

**Theorem:** If  $\lambda$  is an eigenvalue of  $A$  with algebraic *multiplicity*  $q$ , then there is an integer  $p \leq q$  such that  $\text{null}((A - \lambda I)^p)$  has dimension  $q$ .

- We can find  $q$  linearly independent solutions associated with the eigenvalue  $\lambda$ .

## Procedure for $\lambda$ of algebraic multiplicity $q$

To find  $q$  linearly independent solutions associated with  $\lambda$ :

- **Find** the smallest integer  $p$  such that  $\text{null}((A - \lambda I)^p)$  has dimension  $q$ .
- Find a basis  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_q$  of  $\text{null}((A - \lambda I)^p)$ .
- For  $j = 1, 2, \dots, q$

$$\begin{aligned} \mathbf{x}_j(t) &= e^{tA} \mathbf{v}_j \\ &= e^{\lambda t} \left[ \mathbf{v}_j + t(A - \lambda I) \mathbf{v}_j + \frac{t^2}{2!} (A - \lambda I)^2 \mathbf{v}_j \right. \\ &\quad \left. + \dots + \frac{t^{p-1}}{(p-1)!} (A - \lambda I)^{p-1} \mathbf{v}_j \right] \end{aligned}$$

# Example

- Use MATLAB.

## Procedure for a Complex Eigenvalue

If  $\lambda$  is a complex eigenvalue of algebraic multiplicity  $q$ .  
Then  $\bar{\lambda}$  also has algebraic multiplicity  $q$ .

- Find the smallest integer  $p$  such that  $\text{null}((A - \lambda I)^p)$  has dimension  $q$ .
- Find a basis  $\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_q$  of  $\text{null}((A - \lambda I)^p)$ .

- For  $j = 1, 2, \dots, q$  we have solutions

$$\begin{aligned} \mathbf{z}_j(t) &= e^{tA} \mathbf{w}_j \\ &= e^{\lambda t} \left[ \mathbf{w}_j + t(A - \lambda I) \mathbf{w}_j + \frac{t^2}{2!} (A - \lambda I)^2 \mathbf{w}_j \right. \\ &\quad \left. + \dots + \frac{t^{p-1}}{(p-1)!} (A - \lambda I)^{p-1} \mathbf{w}_j \right] \end{aligned}$$

- $\mathbf{z}_1, \dots, \mathbf{z}_q$  together with  $\overline{\mathbf{z}}_1, \dots, \overline{\mathbf{z}}_q$  are  $2q$  linearly independent complex valued solutions.
- For  $j = 1, 2, \dots, q$  set  $\mathbf{x}_j(t) = \operatorname{Re}(\mathbf{z}_j(t))$  and  $\mathbf{y}_j(t) = \operatorname{Im}(\mathbf{z}_j(t))$ . These are  $2q$  linearly independent real valued solutions.

# Stability

Autonomous system  $\mathbf{x}' = \mathbf{f}(\mathbf{x})$  with an equilibrium point at  $\mathbf{x}_0$ .

- Basic question: What happens to *all solutions* as  $t \rightarrow \infty$ ?
- $\mathbf{x}_0$  is *stable* if for every  $\epsilon > 0$  there is a  $\delta > 0$  such that a solution  $\mathbf{x}(t)$  with  $|\mathbf{x}(0) - \mathbf{x}_0| < \delta \Rightarrow |\mathbf{x}(t) - \mathbf{x}_0| < \epsilon$  for all  $t \geq 0$ .
  - ◆ Every solution that starts close to  $\mathbf{x}_0$  stays close to  $\mathbf{x}_0$ .

- $\mathbf{x}_0$  is *asymptotically stable* if it is *stable* and there is an  $\eta > 0$  such that if  $\mathbf{x}(t)$  is a solution with  $|\mathbf{x}(0) - \mathbf{x}_0| < \eta$ , then  $\mathbf{x}(t) \rightarrow \mathbf{x}_0$  as  $t \rightarrow \infty$ .
  - ◆  $\mathbf{x}_0$  is called a *sink*.
  - ◆ Every solution that starts close to  $\mathbf{x}_0$  approaches  $\mathbf{x}_0$ .
- $\mathbf{x}_0$  is *unstable* if there is an  $\epsilon > 0$  such that for any  $\delta > 0$  there is a solution  $\mathbf{x}(t)$  with  $|\mathbf{x}(0) - \mathbf{x}_0| < \delta$  with the property that there are values of  $t > 0$  such that  $|\mathbf{x}(t) - \mathbf{x}_0| > \epsilon$ .
  - ◆ There are solutions starting arbitrarily close to  $\mathbf{x}_0$  that move away from  $\mathbf{x}_0$ .

## Examples $D = 2$

- Sinks are *asymptotically stable*.
  - ◆ The eigenvalues have negative real part.
- Sources are unstable.
  - ◆ The eigenvalues have positive real part.
- Saddles are unstable.
  - ◆ One eigenvalue has positive real part.
- Centers are *stable* but not asymptotically stable.
  - ◆ The eigenvalues have real part = 0.

**Theorem:** Let  $A$  be an  $n \times n$  real matrix.

- Suppose the real part of every eigenvalue of  $A$  is negative. Then  $\mathbf{0}$  is an **asymptotically stable** equilibrium point for the system  $\mathbf{x}' = A\mathbf{x}$ .
- Suppose  $A$  has at least one eigenvalue with positive real part. Then  $\mathbf{0}$  is an **unstable** equilibrium point for the system  $\mathbf{x}' = A\mathbf{x}$ .

# Examples

- $D = 2$ 
  - ♦  $T^2 - 4D = 0$ .
    - ▶  $T < 0 \Rightarrow$  sink.  $T > 0 \Rightarrow$  source.

- $\mathbf{y}' = A\mathbf{y}$ ,

$$A = \begin{pmatrix} -2 & -18 & -7 & -14 \\ 1 & 6 & 2 & 5 \\ 2 & 2 & -3 & 0 \\ -2 & -8 & -1 & -6 \end{pmatrix}.$$

- ♦  $A$  has eigenvalues  $-1$ ,  $-2$ , &  $-1 \pm i$ .
- ♦  $\mathbf{0}$  is asymptotically stable.

# Higher Order Equations

$$y^{(n)} + a_1 y^{(n-1)} + \cdots + a_{n-1} y' + a_n y = 0$$

- Second order:  $y'' + py' + qy = 0$ .
- Equivalent system:  $\mathbf{x}' = A\mathbf{x}$ , where

$$\mathbf{x} = \begin{pmatrix} y \\ y' \end{pmatrix} \quad \text{and} \quad A = \begin{pmatrix} 0 & 1 \\ -q & -p \end{pmatrix}.$$

- ♦ A fundamental set of solutions for the system consists of two linearly independent solutions.

## Linear Independence

**Definition:** Two functions  $u(t)$  and  $v(t)$  are *linearly independent* if neither is a constant multiple of the other.

- $u(t)$  and  $v(t)$  are linearly independent solutions to  $y'' + py' + qy = 0 \Leftrightarrow \begin{pmatrix} u \\ u' \end{pmatrix} \& \begin{pmatrix} v \\ v' \end{pmatrix}$  are linearly independent solutions to the equivalent **system**.

## General Solution

**Theorem:** Suppose that  $y_1(t)$  &  $y_2(t)$  are linearly independent solutions to the equation

$$y'' + py' + qy = 0.$$

Then the general solution is

$$y(t) = C_1y_1(t) + C_2y_2(t).$$

**Definition:** A set of two linearly independent solutions is called a *fundamental set of solutions*.

## Solutions to $y'' + py' + qy = 0$ .

- Equivalent system:  $\mathbf{x}' = A\mathbf{x}$ , where

$$\mathbf{x} = \begin{pmatrix} y \\ y' \end{pmatrix} \quad \text{and} \quad A = \begin{pmatrix} 0 & 1 \\ -q & -p \end{pmatrix}.$$

- Look for exponential solutions  $y(t) = e^{\lambda t}$ .
- *Characteristic equation:*  $\lambda^2 + p\lambda + q = 0$ .
- *Characteristic polynomial:*  $\lambda^2 + p\lambda + q$ .
- Same for the 2<sup>nd</sup> order equation and the system.

## Real Roots

- If  $\lambda$  is a root to the **characteristic polynomial** then  $y(t) = e^{\lambda t}$  is a solution.
- If  $\lambda$  is a root to the characteristic polynomial of multiplicity 2, then  $y_1(t) = e^{\lambda t}$  and  $y_2(t) = te^{\lambda t}$  are linearly independent solutions.

## Complex Roots

- If  $\lambda = \alpha + i\beta$  is a complex root of the **characteristic equation**, then so is  $\bar{\lambda} = \alpha - i\beta$ .
- A complex valued fundamental set of solutions is

$$z(t) = e^{\lambda t} \quad \text{and} \quad \bar{z}(t) = e^{\bar{\lambda} t}.$$

- A real valued fundamental set of solutions is

$$x(t) = e^{\alpha t} \cos \beta t \quad \text{and} \quad y(t) = e^{\alpha t} \sin \beta t.$$

## Examples

- $y'' - 5y' + 6y = 0.$
- $y'' + 25y = 0.$
- $y'' + 4y' + 13y = 0.$

## Key to Computing $e^{tA}$ or $e^{tA}\mathbf{v}$

Suppose that  $A$  an  $n \times n$  matrix, and  $\lambda$  a number (an eigenvalue).

- $A = \lambda I + (A - \lambda I)$ ; ( $\lambda I$  &  $A - \lambda I$  commute.)

$$\begin{aligned}
 e^{tA} &= e^{t[\lambda I + (A - \lambda I)]} \\
 &= e^{t\lambda I} \cdot e^{t(A - \lambda I)} \\
 &= e^{\lambda t} \cdot e^{t(A - \lambda I)} \\
 &= e^{\lambda t} \cdot [I + t(A - \lambda I) + \frac{t^2}{2!}(A - \lambda I)^2 + \dots]
 \end{aligned}$$

# 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  $\lambda_j$ .