## Day 21: August 4th

# Chapter: 6 Laplace Transforms Chapter 6 not on exam Review

# Laplace Transforms

another way to solve initial value problems

$$y^{''} + ay^{'} + by = F(t), y(0) = \alpha, y^{'}(0) = \beta$$

- linear autonomous systems
  - electrical circuits
  - harmonic oscillators
  - optical devices
  - mechanical systems.

t-domain  $y(t) \xleftarrow{\mathcal{L}} Y(s)$  s-domain Laplace Transform

• Definition: 
$$\mathcal{L}[y(t)] = \int_{0}^{\infty} e^{-st}y(t)dt = Y(s)$$

• <u>Application</u>: initial value problem

$$\begin{aligned} y' + ay &= F(t) \\ y(0) &= y_0 \end{aligned} \qquad \begin{bmatrix} y'' + ay' + by &= F(t) \\ y(0) &= \alpha, y'(0) &= \beta \end{aligned} \\ \bullet \text{ take Laplace Transform of both sides:} \\ \mathcal{L}[y'] + a\mathcal{L}[y] &= \mathcal{L}[F(t)) \\ \mathcal{L}[y''] + a\mathcal{L}[y'] + b\mathcal{L}[y] &= \mathcal{L}[F(t)) \end{aligned}$$

• Laplace Transform:  

$$\mathcal{L}[y^{''}] = s^2 \mathcal{L}[y] - sy(0) - y^{'}(0)$$

$$\mathcal{L}[y^{'}] = s\mathcal{L}[y] - y(0)$$
• Table  

$$\mathcal{L}[c] = \frac{c}{s}, s > 0 \qquad \mathcal{L}[e^{at}] = \frac{1}{s-a}, s > a$$

$$\mathcal{L}[u_a(t)] = \frac{e^{-as}}{s}$$

$$\mathcal{L}[\sin(bt)] = \frac{b}{s^2 + b^2} \mathcal{L}[\cos(bt)] = \frac{s}{s^2 + b^2}$$





- Differential Equations
  - Modeling
    - Come up with models.
    - Analyze models.
  - Solve
    - Analytical: Separate and Integrate
      - Verify Solutions
    - Qualitatively:
    - Numerically: Euler's Method.



## Non-homogeneous:

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$$\frac{dy}{dt} + g(t) \cdot y = r(t)$$
 | Non separable

1. Find general solution for (H) equation:  $y_{\rm H}(t)$ 

2. Find particular solution for (NH) equation:  $y_{\rm p}(t)$ 

General solution for (NH) equation:

$$y_{\rm NH}(t) = y_{\rm H}(t) + y_{\rm p}(t)$$

# Existence and Uniqueness $\frac{dy}{dt} = F(y,t)$

If: Continuously differentiable in y and t, i.e. F is differentiable in y and t and the derivative is continuous

**Then**: There exists a unique solution to initial value problem  $y(t_0) = y_0$ defined for  $t_0 - A < t < t_0 + A$ 

**Result:** Solutions can not cross.

# $\frac{Bifurcations}{dy} = F_a(y)$

A bifurcation is a BIG change in the over all behavior of solutions as a changes

#### Most common bifurcations

number of equilibrium point changes

type of equilibrium point changes

## Numerical Method

#### Euler's Method:

Solve 
$$\frac{dy}{dt} = F(y,t) \ y(t_0) = y_0$$
 numerically  
Idea: Travel along slope field with step size  $\Delta t$   
Formula

$$y_{n+1} = y_n + F(t_n, y_n) \cdot \Delta t$$
$$t_{n+1} = t_n + \Delta t$$

$$\vec{\mathbf{Y}}' = \mathbf{A} \cdot \vec{\mathbf{Y}} \text{ where } \vec{\mathbf{Y}} = \begin{pmatrix} x \\ y \end{pmatrix}$$

#### • Equilibrium points

- If det  $\mathbf{A} \neq 0 \Rightarrow (x, y) = (0, 0)$  only eq point.
- If  $\det \mathbf{A} = 0$  straight line of equilibrium points.
- Characteristic Equation

$$\det \begin{pmatrix} a-\lambda & b \\ c & d-\lambda \end{pmatrix} = 0$$

Eigenvalues roots of characteristic equation

• Eigenvectors  

$$\begin{aligned}
\mathbf{A} \cdot \vec{\mathbf{v}} &= \lambda \vec{\mathbf{v}} \\
\hline
\mathbf{A} \cdot \vec{\mathbf{v}} &= \lambda \vec{\mathbf{v}}
\end{aligned}$$
• Finding eigenvectors  
• Non-zero solutions of  

$$\begin{pmatrix}
\mathbf{A} - \lambda \mathbf{I} \\
y \\
z
\end{pmatrix} = \begin{pmatrix}
0 \\
0 \\
0
\end{pmatrix}$$

$$\begin{pmatrix}
a - \lambda & b \\
c & d - \lambda
\end{pmatrix} \begin{pmatrix}
x \\
y \\
z
\end{pmatrix} = \begin{pmatrix}
0 \\
0 \\
0
\end{pmatrix}$$

always get a redundant system of equations

### Real Distinct Eigenvalues • if we have two distinct real eigenvalues: $\lambda_1, \lambda_2$ and can find two distinct eigenvectors: $\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$ $\vec{\mathbf{Y}}_1(t) = e^{\lambda_1 t} \cdot \vec{\mathbf{v}}_1$ two straight $\vec{\mathbf{Y}}_2(t) = e^{\lambda_2 t} \cdot \vec{\mathbf{v}}_2$ line solutions • If $\vec{\mathbf{Y}}_1(0)$ and $\vec{\mathbf{Y}}_2(0)$ are linearly independent $|\vec{\mathbf{Y}}(t) = k_1 \vec{\mathbf{Y}}_1(t) + k_2 \vec{\mathbf{Y}}_2(t)|$ $= k_1 e^{\lambda_1 t} \cdot \vec{\mathbf{v}}_1 + k_2 e^{\lambda_2 t} \cdot \vec{\mathbf{v}}_2$



#### <u>Complex Eigenvalues</u> • One solution: $\vec{\mathbf{Y}}_1(t) = e^{\lambda_1 t} \cdot \vec{\mathbf{v}}_1$ break up solution $\vec{\mathbf{Y}}_{\mathrm{Im}}(t)$ $\vec{\mathbf{Y}}_{\mathrm{Re}}(t)$ Real part of solution Imaginary part of solution • Fact: Both $\vec{\mathbf{Y}}_{Re}(t)$ and $\vec{\mathbf{Y}}_{Im}(t)$ are solutions. ${}_{\bullet}\,{\rm lf}\,\vec{\mathbf{Y}}_{\rm Re}(t)\,{\rm and}\,\vec{\mathbf{Y}}_{\rm Im}(t)\,{\rm are}$ linearly independent <u>Gen sol:</u> $\vec{\mathbf{Y}}(t) = k_1 \vec{\mathbf{Y}}_{\text{Re}}(t) + k_2 \vec{\mathbf{Y}}_{\text{Im}}(t)$









• Strategy for solving: 
$$y'' + by' + ky = 0$$
  
guess:  $y = e^{st}$   
 $s^2 + bs + k = 0$   
find roots:  $s_1, s_2$   
• Case 1  
 $s_1, s_2 \in \mathbb{R}$   $s_1 \neq s_2$   
get solutions:  $y_1(t) = e^{s_1 t}$   $y_2(t) = e^{s_2 t}$   
general solution:  $y(t) = k_1 e^{s_1 t} + k_2 e^{s_2 t}$ 

• Case 2: complex roots:  $s = \alpha \pm i\beta$ Formula for general solution:  $y(t) = e^{\alpha t} (k_1 \cos(\beta t) + k_2 \sin(\beta t))$ • <u>Case 3</u>: repeated roots:  $s_1$ get solutions:  $y_1(t) = e^{s_1 t}$   $y_2(t) = t \cdot e^{s_1 t}$ general solution:  $y(t) = k_1 e^{s_1 t} + k_2 t e^{s_1 t}$ 

## Non-linear Systems

$$\frac{dx}{dt} = F(x, y)$$
$$\frac{dy}{dt} = G(x, y)$$

- <u>Two techniques</u>:
  - Equilibrium Point Analysis
    - Linearization
  - Nullclines

Equilibrium Point Analysis  

$$\begin{array}{l} & \left\{ \begin{matrix} x' \\ y' \\ = F(x,y) \end{matrix} \right\} & \text{Non-linear System} \\ & \text{System} \\ & \text{Equilibrium point: } (x_0,y_0) & \text{evaluated at } (x_0,y_0) \\ & \text{Jacobian:} & \mathbf{J}_{|(x_0,y_0)} = \begin{pmatrix} \begin{array}{c} \partial F \\ \partial x \\ \partial G \\ \partial x \\ \partial g \\ \partial y \end{pmatrix} \\ & \text{near} (x_0,y_0) \text{ solutions of } \star \text{ "resemble" solutions of } \\ & \mathbf{J}_{|(x_0,y_0)|} = \begin{pmatrix} \mathbf{J}_{|(x_0,y_0)|} & \mathbf{J}_{|(x_0,y_0)|} \\ & \mathbf{J}_{|(x_0,y_0)|} & \mathbf{J}_{|(x_0,y_0)|} \\ & \mathbf{J}_{|(x_0,y_0)|} = \begin{pmatrix} \begin{array}{c} \partial F \\ \partial x \\ \partial g \\ \partial x \\ \partial y \\ \partial y \end{pmatrix} \\ & \mathbf{J}_{|(x_0,y_0)|} \\ & \mathbf{J}_{|(x_0,y_0)|} = \begin{pmatrix} \begin{array}{c} \partial F \\ \partial x \\ \partial g \\ \partial y \\ \partial y \\ \partial y \\ \partial y \end{pmatrix} \\ & \mathbf{J}_{|(x_0,y_0)|} \\ & \mathbf{J}_{|(x_0,y_0)|$$

 $\vec{\mathbf{Y}} = \mathbf{J} \cdot \vec{\mathbf{Y}}$ near (0,0)

## Null-clines

• where 
$$x = 0$$

vector field is vertical

1

$$x' = F(x, y) = 0$$

gives x-nullcline

• <u>y-nullcline</u>, • where y = 0

y

vector field is horizontal

$$y' = G(x, y) = 0$$
  
gives y-nullcline

