


CDS 101: Lecture 4.1

Linear Systems

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18 October 2004



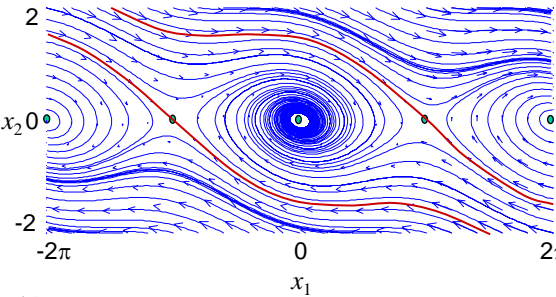
Goals:

- Describe linear system models: properties, examples, and tools
- Characterize stability and performance of linear systems in terms of eigenvalues
- Compute linearization of a nonlinear systems around an equilibrium point

Reading:

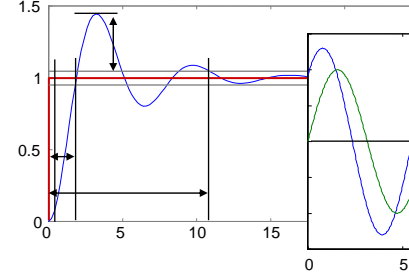
- Åström and Murray, *Analysis and Design of Feedback Systems*, Ch 4
- Packard, Poola and Horowitz, *Dynamic Systems and Feedback*, Sections 19, 20, 22 (available via course web page)

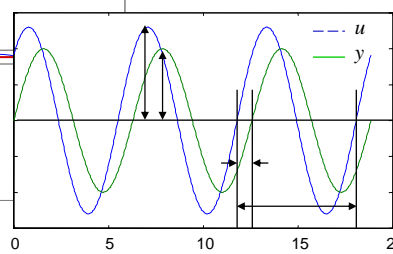
Review from Last Week

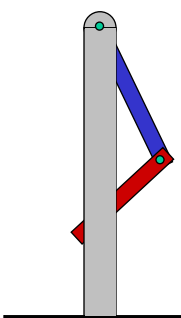


Key topics for this lecture

- Stability of equilibrium points
- Local versus global behavior
- Performance specification via step and frequency response







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What is a *Linear System*?

Linearity of functions: $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$

- Zero at the origin: $f(0) = 0$
- Addition: $f(x + y) = f(x) + f(y)$
- Scaling: $f(\alpha x) = \alpha f(x)$

$$\left. \begin{array}{l} f(\alpha x + \beta y) = \\ \alpha f(x) + \beta f(y) \end{array} \right\} \text{Canonical example: } f(x) = Ax$$

Linearity of systems: sums of solutions

<p style="text-align: center; color: red;">Dynamical system</p> $\dot{x} = Ax$ $x(0) = x_{10} \quad x(0) = x_{20}$ $\rightarrow x(t) = x_1(t) \quad \rightarrow x(t) = x_2(t)$ <p style="text-align: center;">⇓</p> $x(0) = \alpha x_{10} + \beta x_{20}$ $\rightarrow x(t) = \alpha x_1(t) + \beta x_2(t)$	<p style="text-align: center; color: red;">Control system</p> $\dot{x} = Ax + Bu$ $y = Cx + Du$ $x(0) = 0, u(t) = u_1(t) \quad x(0) = 0, u(t) = u_2(t)$ $\rightarrow y(t) = y_1(t) \quad \rightarrow y(t) = y_2(t)$ <p style="text-align: center;">⇓</p> $x(0) = 0, u(t) = \alpha u_1(t) + \beta u_2(t)$ $\rightarrow y(t) = \alpha y_1(t) + \beta y_2(t)$
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Linear Systems

u

$$\begin{aligned} \dot{x} &= Ax + Bu \\ y &= Cx + Du \\ x(0) &= 0 \end{aligned}$$

y

u_1

y_1

+

u_2

y_2

+

$u_1 + u_2$

$y_1 + y_2$

Input/output linearity at $x(0) = 0$

- Linear systems are linear in initial condition *and* input \Rightarrow need to use $x(0) = 0$ to add outputs together
- For different initial conditions, you need to be more careful (sounds like a good midterm question)

Linear system \Rightarrow step response and frequency response scale with input amplitude

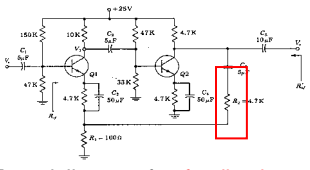
- 2X input \Rightarrow 2X output
- Allows us to use *ratios* and *percentages* in step/freq response. These are *independent* of input amplitude
- Limitation: input saturation \Rightarrow only holds up to certain input amplitude

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Why are Linear Systems Important?

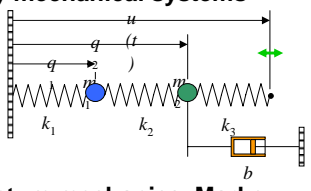
Many important examples

Electronic circuits



- Especially true after **feedback**
- Frequency response is key performance specification (think telephones)

Many mechanical systems



Quantum mechanics, Markov chains, ...

Many important tools

Frequency response, step response, etc

- Traditional tools of control theory
- Developed in 1930's at Bell Labs; intercontinental telecom

Classical control design toolbox } CDS 101/110a

- Nyquist plots, gain/phase margin
- Loop shaping

Optimal control and estimators } CDS 110b

- Linear quadratic regulators
- Kalman estimators

Robust control design } CDS 110b/212

- H_∞ control design
- μ analysis for structured uncertainty

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Solutions of Linear Systems: The Matrix Exponential

$$\begin{aligned} \dot{x} &= Ax + Bu \\ y &= Cx + Du \end{aligned} \quad \longrightarrow \quad y(t) = ???$$

Scalar linear system, with no input

$$\begin{aligned} \dot{x} &= ax \\ y &= cx \end{aligned} \quad x(0) = x_0 \quad \longrightarrow \quad x(t) = e^{at} x_0 \quad \longrightarrow \quad y(t) = ce^{at} x_0$$

Matrix version, with no input

$$\begin{aligned} \dot{x} &= Ax \\ y &= Cx \end{aligned} \quad x(0) = x_0 \quad \longrightarrow \quad x(t) = e^{At} x_0 \quad \longrightarrow \quad y(t) = Ce^{At} x_0$$

initial(A,B,C,D,x0);

Matrix exponential

- Analog to the scalar case; defined by series expansion:

$$e^M = I + M + \frac{1}{2!} M^2 + \frac{1}{3!} M^3 + \dots \quad P = \text{expm}(M)$$

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Stability of Linear Systems

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

$$x(t) = e^{At}x_0$$

Q: when is the system asymptotically stable?

$$\lim_{t \rightarrow \infty} x(t) = 0$$

Stability is determined by the eigenvalues of the matrix A

- Simple case: diagonal system

$$\dot{x} = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} x \Rightarrow x(t) = \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} x_0$$

Stable if $\lambda_i \leq 0$
 Asy stable if $\lambda_i < 0$
 Unstable if $\lambda_i > 0$

- More generally: transform to "Jordan" form

$$\dot{x} = T^{-1}JTx \quad J = \begin{bmatrix} J_1 & & 0 \\ & \ddots & \\ 0 & & J_k \end{bmatrix} \quad J_i = \begin{bmatrix} \lambda_i & 1 & 0 \\ & \ddots & 1 \\ 0 & & \lambda_i \end{bmatrix}$$

Asy stable if $\text{Re}(\lambda_i) < 0$
 Unstable if $\text{Re}(\lambda_i) > 0$
 Indeterminate if $\text{Re}(\lambda_i) = 0$

Form of eigenvalues determines system behavior
Linear systems are automatically globally stable or unstable

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Eigenstructure of Linear Systems

Real e-values
 $\text{Re}(\lambda_i) < 0$

Real e-values
 $\text{Re}(\lambda_i) < 0$
 $\text{Re}(\lambda_j) > 0$

Complex e-values
 $\text{Re}(\lambda_i) = 0$

Complex e-values
 $\text{Re}(\lambda_i) < 0$

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Step and Frequency Response

$\dot{x} = Ax + Bu$
 $y = Cx + Du$

$u(t) = 1(t)$

$u(t) = A\sin(\omega t)$

Effect of eigenstructure on step response

- Complex eigenvalues with small real part lead to oscillatory response
- Frequency of oscillations $\approx \omega_i$

Effects of eigenstructure on frequency response

- Eigenvalues determine "break points" for frequency response
- Complex eigenvalues lead to peaks in response function near ω_i

The figure contains three subplots. The top-left is a pole-zero map with poles marked by red 'x' and zeros by red dots on a complex plane. The top-right is a step response plot showing an oscillatory signal with a period $T \approx 2\pi/\omega_p$. The bottom plot is a Bode diagram showing magnitude and phase plots with a resonance peak at $\approx \omega_p$.

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Computing Frequency Responses

Technique #1: plot input and output, measure relative amplitude and phase

- Use MATLAB or SIMULINK to generate response of system to sinusoidal output
- Gain = A_y/A_u
- Phase = $2\pi \cdot \Delta T/T$
- Note: In general, gain and phase will depend on the input amplitude

Technique #2 (linear systems): use MATLAB bode command

- Assumes linear dynamics in state space form:

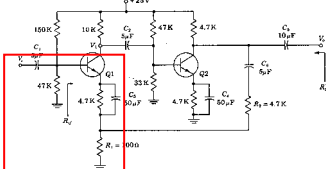
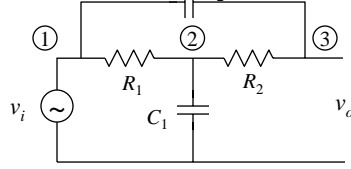
$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$
- Gain plotted on log-log scale
 - dB = $20 \log_{10}(\text{gain})$
- Phase plotted on linear-log scale

The figure shows a sinusoidal input u (blue) and output y (green) with measured amplitudes A_u and A_y , and a phase shift ΔT relative to period T . Below is a Bode plot for $\text{bode}(A,B,C,D)$ showing magnitude and phase.

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Example: Electrical Circuit

"Bridged Tee Circuit"

Derivation based on Kirchoff's laws for electrical circuits (Ph 2)

- Sum of currents at nodes = 0:

$$C_1 \frac{dv_2}{dt} = \frac{v_1 - v_2}{R_1} - \frac{v_2 - v_3}{R_2} \qquad C_2 \frac{d(v_3 - v_1)}{dt} = -\frac{v_3 - v_2}{R_2}$$

- Rewrite in terms of new states: $v_{c1} = v_2, v_{c2} = v_3 - v_1$

$$\frac{d}{dt} \begin{bmatrix} v_{c1} \\ v_{c2} \end{bmatrix} = \begin{bmatrix} -\frac{1}{C_1} \left(\frac{1}{R_1} + \frac{1}{R_2} \right) & -\frac{1}{C_1 R_2} \\ -\frac{1}{C_2 R_2} & -\frac{1}{C_2 R_2} \end{bmatrix} \begin{bmatrix} v_{c1} \\ v_{c2} \end{bmatrix} + \begin{bmatrix} \frac{1}{C_1} \left(\frac{1}{R_1} + \frac{1}{R_2} \right) \\ V_{c2} \end{bmatrix} v_i \qquad v_o = \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} v_{c1} \\ v_{c2} \end{bmatrix} + v_i$$

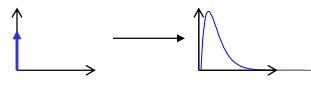
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Linear Control Systems and Convolution

$$\begin{aligned} \dot{x} &= Ax + Bu \\ y &= Cx + Du \end{aligned} \qquad \longrightarrow \qquad y(t) = \underbrace{Ce^{At}x(0)}_{\text{homogeneous}} + ???$$

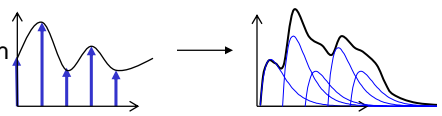
Impulse response, $h(t) = Ce^{At}B$

- Response to input "impulse"
- Equivalent to "Green's function"



Linearity \Rightarrow compose response to arbitrary $u(t)$ using convolution

- Decompose input into "sum" of shifted impulse functions
- Compute impulse response for each
- "Sum" impulse response to find $y(t)$



Complete solution: use integral instead of "sum"

$$y(t) = Ce^{At}x(0) + \int_{\tau=0}^t Ce^{A(t-\tau)}Bu(\tau)d\tau + Du(t)$$

- linear with respect to initial condition *and* input
- 2X input \Rightarrow 2X output when $x(0) = 0$

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Matlab Tools for Linear Systems

$$y(t) = \underbrace{Ce^{At}x(0)}_{\tau=0} + \int_0^t \underbrace{Ce^{A(t-\tau)}Bu(\tau)}_{\tau=0} d\tau + Du(t)$$

```

A = [-1 1; 0 -1]; B = [0; 1];
C = [1 0]; D = [0];
x0 = [1; 0.5];

sys = ss(A,B,C,D);
initial(sys, x0);
impulse(sys);

t = 0:0.1:10;
u = 0.2*sin(5*t) + cos(2*t);
lsim(sys, u, t, x0);
    
```

Initial Condition Results

Linear Simulation Results

Other MATLAB commands

- gensig, square, sawtooth – produce signals of diff. types
- step, impulse, initial, lsim – time domain analysis
- bode, freqresp, evalfr – frequency domain analysis

Itview – linear time invariant system plots

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Linearization Around an Equilibrium Point

$$\begin{aligned} \dot{x} &= f(x,u) & \dot{z} &= Az + Bv \\ y &= h(x,u) & w &= Cz + Dv \end{aligned}$$

"Linearize" around $x=x_e$

$f(x_e, u_e) = 0 \quad y_e = h(x_e, u_e)$

$z = x - x_e \quad v = u - u_e \quad w = y - y_e$

$$A = \left. \frac{\partial f}{\partial x} \right|_{(x_e, u_e)} \quad B = \left. \frac{\partial f}{\partial u} \right|_{(x_e, u_e)}$$

$$C = \left. \frac{\partial h}{\partial x} \right|_{(x_e, u_e)} \quad D = \left. \frac{\partial h}{\partial u} \right|_{(x_e, u_e)}$$

Remarks

- In examples, this is often equivalent to small angle approximations, etc
- Only works *near* to equilibrium point

Full nonlinear model

Linear model (honest!)

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Local Stability of Nonlinear Systems

Asymptotic stability of the linearization implies local asymptotic stability of equilibrium point

- Linearization around equilibrium point captures “tangent” dynamics

$$\dot{x} = f(x) = A \cdot (x - x_e) + o(x - x_e) \leftarrow \text{higher order terms}$$

- If linearization is *unstable*, can conclude that nonlinear system is locally unstable
- If linearization is *stable* but not *asymptotically stable*, can't conclude anything about nonlinear system:

$$\dot{x} = \pm x^3 \xrightarrow{\text{linearize}} \dot{x} = 0 \quad \begin{array}{l} \bullet \text{ linearization is stable (but not asy stable)} \\ \bullet \text{ nonlinear system can be asy stable or unstable} \end{array}$$

Local approximation particularly appropriate for control systems design

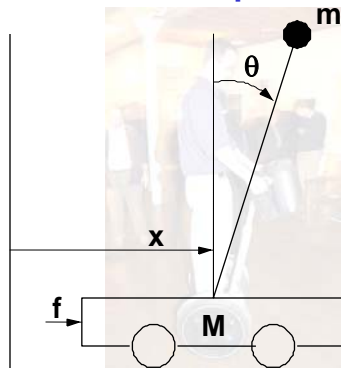
- Control often used to *ensure* system stays near desired equilibrium point
- If dynamics are well-approximated by linearization near equilibrium point, can use this to design the controller that keeps you there (!)

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Example: Inverted Pendulum on a Cart



$$\begin{aligned} (M + m)\ddot{x} + ml \cos \theta \ddot{\theta} &= -b\dot{x} + ml \sin \theta \dot{\theta}^2 + f \\ (J + ml^2)\ddot{\theta} + ml \cos \theta \ddot{x} &= -mgl \sin \theta \end{aligned}$$

- State: $x, \theta, \dot{x}, \dot{\theta}$
- Input: $u = F$
- Output: $y = x$
- Linearize according to previous formula around $\theta = \pi$

$$\frac{d}{dt} \begin{bmatrix} x \\ \theta \\ \dot{x} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & \frac{m^2 g l^2}{J(M+m) + Mml^2} & \frac{-(J + ml^2)b}{J(M+m) + Mml^2} & 0 \\ 0 & \frac{mgl(M+m)}{J(M+m) + Mml^2} & \frac{-mlb}{J(M+m) + Mml^2} & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ \frac{J + ml^2}{J(M+m) + Mml^2} \\ \frac{ml}{J(M+m) + Mml^2} \end{bmatrix} u$$

$$y = \begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix} x$$

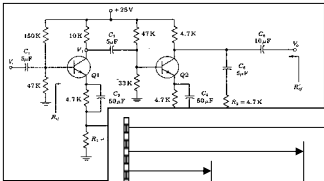
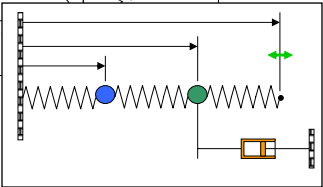
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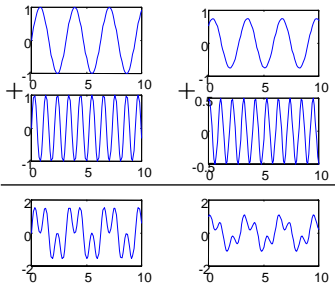
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Summary: Linear Systems

$$\begin{aligned} \dot{x} &= Ax + Bu \\ y &= Cx + Du \\ x(0) &= 0 \end{aligned}$$



Properties of linear systems

- Linearity with respect to initial condition and inputs
- Stability characterized by eigenvalues
- Many applications and tools available
- Provide local description for nonlinear systems

$$y(t) = Ce^{At}x(0) + \int_{\tau=0}^t Ce^{A(t-\tau)}Bu(\tau)d\tau + Du(t)$$

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