16.513 Control Systems: Lecture note #2

≻ Last Time:

- Introduction
 - Motivation
 - Course Overview
 - Course project
- Matrix Operations -- Fundamental to Linear Algebra
 - Determinant
 - Matrix Multiplication
 - Eigenvalue
 - Rank

Today:

- Math. Descriptions of Systems
 - Classification of systems
 - Linear systems
 - Linear-time-invariant systems
 - State variable description
 - Linearization
- Modeling of electric circuits

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2. Mathematical Descriptions of Systems



- Classification of systems

- Linear systems

- Linear time invariant (LTI) systems

2.1 Classification of Systems

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- Basic assumption: When an input signal is applied to the system, a unique output is obtained
- **Q.** How do we classify systems?
 - Number of inputs/outputs; with/without memory; causality; dimensionality; linearity; time invariance
- The number of inputs and outputs
 - When p = q = 1, it is called a single-input singleoutput (SISO) system
 - When p > 1 and q > 1, it is called a multi-input multi-output (MIMO) system
 - MISO, SIMO defined similarly

• Memoryless vs. with Memory

- If y(t) depends on u(t) only, the system is said to be memoryless, otherwise, it has memory
- An example of a memoryless system?

$$u(t) \xrightarrow{+} R_1 \xrightarrow{R_2} y(t) \xrightarrow{+} y(t) = \frac{R_2}{R_1 + R_2} u(t) \sim \text{Memoryless}$$

- An example of a system with memory?

$$\begin{array}{cccc}
 & Ri + L\frac{di}{dt} = u \quad or \quad \frac{di}{dt} + \frac{R}{L}i = \frac{1}{L}u \\
 & i(t) = e^{-\frac{R}{L}(t-t_0)}i(t_0) + \frac{1}{L}\int_{t_0}^t e^{-\frac{R}{L}(t-\tau)}u(\tau)d\tau \\
\end{array}$$

- -i(t) depends on $i(t_0)$ and $u(\tau)$ for $t_0 \le \tau \le t$, not just u(t)
- A system with memory
- Causality: No output before an input is applied

- A system is causal or non-anticipatory if $y(t_0)$ depends only on u(t) for $t \le t_0$ and is independent of u(t) for $t > t_0$
- Is the circuit discussed last time causal?







- Can you truly build a physical system like this?

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- All physical systems are causal!

The Concept of State

- The state of a system at t_0 is the information at t_0 that, together with $u_{[t_0,\infty)}$, uniquely determines the behavior of the system for $t \ge t_0$
- The number of state variables = the number of ICs needed to solve the problem
- For an RLC circuit, the number of state variables = the number of C + the number of L (except for degenerated cases)
- A natural way to choose state variables as what we have done earlier: $\{v_c\}$ and $\{i_L\}$
- Is this the unique way to choose state variables?

 Any invertible transformation of the above can serve as a state, e.g.,

$\begin{bmatrix} x_1(t) \end{bmatrix}$	2	1	$\left\lceil v(t) \right\rceil_{-}$	$\left\lceil 2v(t) + i(t) \right\rceil$
$\begin{bmatrix} x_2(t) \end{bmatrix}^{=}$	0	1	$\lfloor i(t) \rfloor^{=}$	i(t)

- Although the number of state variables = 2, there are infinite numbers of representations
- Order of dimension of a system: The number of state variables
 - If the dimension is a finite number ⇒ Finite dimensional (or lumped) system
 - Otherwise, an infinite dimensional (or distributed) system

An example of an infinite dimensional system

$$u(t)$$
 System $y(t) = u(t-1)$ A delay line

- Given u(t) for $t \ge 0$, what information is needed to know y(t) for $t \ge 0$?



We need an infinite amount of information ⇒ An infinite dimensional system

2.2 Linear Systems

Linearity

- Double the efforts double the outcome?
 - Suppose we have the following (state,input)-output pairs:

$$\begin{cases} x_1(t_0) \\ u_1(t), t \ge t_0 \end{cases} \rightarrow y_1(t), t \ge t_0$$

$$\begin{cases} x_2(t_0) \\ u_2(t), t \ge t_0 \end{cases} \rightarrow y_2(t), t \ge t_0$$

– What would be the output of

$$\begin{cases} x_1(t_0) + x_2(t_0) \\ u_1(t) + u_2(t), t \ge t_0 \end{cases} \to y_1(t) + y_2(t), t \ge t_0$$

- If this is true ~ Additivity
- How about

$$\left. \begin{array}{c} \alpha \ x_1(t_0) \\ \alpha \ u_1(t), t \ge t_0 \end{array} \right\} \rightarrow \quad \alpha \ y_1(t), t \ge t_0$$

- If this is true ~ Homogeneity
- Combined together to have:

$$\frac{\alpha_{1}x_{1}(t_{0}) + \alpha_{2}x_{2}(t_{0})}{\alpha_{1}u_{1}(t) + \alpha_{2}u_{2}(t), t \ge t_{0}} \rightarrow \alpha_{1}y_{1}(t) + \alpha_{2}y_{2}(t)$$

- If this is true ~ Superposition or linearity property
- A system with such a property: a Linear System

• Are R, L, and C linear elements?

$$v_R = Ri_R, \quad v_L = L\frac{di_L}{dt}, \quad i_C = C\frac{dv_C}{dt}$$

- Yes (differentiation is a linear operation)



• Also, KVL and KCL are linear constraints. When put together, we have a linear system



• The additivity property implies that

$$y(t)$$
 due to $\begin{cases} x_1(t_0) \\ u_1(t), t \ge t_0 \end{cases} = y(t)$ due to $\begin{cases} x_1(t_0) \\ u_1(t) \equiv 0 \end{cases} + y(t)$ due to $\begin{cases} x_1(t_0) = 0 \\ u_1(t), t \ge t_0 \end{cases}$

- Response = zero-input response + zero-state response

Response of a Linear System



- How can we determine the output y(t)?
- Can be derived from u(t) + the unit impulse response based on linearity

• Let $\delta_{\Delta}(t-\tau)$ be a square **pulse** at time τ with width Δ and height $1/\Delta$



• As $\Delta \rightarrow 0$, we obtain a shifted unit impulse



Let the unit impulse response be g (t, τ). Based on linearity,

$$y(t) = \int_{-\infty}^{\infty} g(t,\tau) u(\tau) d\tau$$

• If the system is causal,

$$g(t,\tau) = 0$$
 for $t < \tau$ $y(t) = \int_{-\infty}^{t} g(t,\tau)u(\tau)d\tau$

- A system is said to be relaxed at t₀ if the initial state at t₀ is 0
 - In this case, y(t) for $t \ge t_0$ is caused exclusively by u(t) for $t \ge t_0$

$$y(t) = \int_{t_0}^t g(t,\tau)u(\tau)d\tau$$

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- How about a system with p inputs and q outputs?
 - Have to analyze the relationship for input/output pairs

$$y(t) = \int_{0}^{t_{0}} G(t,\tau)u(\tau)d\tau$$

$$G(t,\tau) = \begin{bmatrix} g_{11}(t,\tau) & g_{12}(t,\tau) & g_{1p}(t,\tau) \\ g_{21}(t,\tau) & g_{22}(t,\tau) & g_{2p}(t,\tau) \\ g_{q1}(t,\tau) & g_{q2}(t,\tau) & g_{qp}(t,\tau) \end{bmatrix} \qquad g_{ij}(t,\tau): \text{ The impulse response between the jth input and ith output}$$

State-Space Description

• A linear system can be described by

$$\dot{x}(t) = A(t)x(t) + B(t)u(t)$$
$$y(t) = C(t)x(t) + D(t)u(t)$$

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2.3 Linear Time-Invariant (LTI) Systems

- Time Invariance: The characteristics of a system do not change over time
 - What are some of the LTI examples? Time-varying examples?
 - What happens for an LTI system if u(t) is delayed by T?



If the same IC is also shifted by T

- This property can be stated as:

$$\begin{array}{c} x(0) = x_0 \\ u(t), \ t \ge 0 \end{array} \end{array} \rightarrow y(t), \ t \ge t_0$$

$$\begin{array}{c} & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ x(T) = x_0 \\ u(t-T), \ t \ge T \end{array} \end{array} \rightarrow y(t-T), \ t \ge T$$

Practice: Suppose $u(t) \rightarrow y(t)=1-exp(-t)$, y(t)=0 for t<0. What is the response to u(t+1)+u(t-1)?

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• What happens to the unit impulse response when the system is LTI?

$$g(t,\tau) = g(t+T,\tau+T) \text{ for any } T$$

$$g(t,\tau) = g(t-\tau,\tau-\tau) = g(t-\tau,0) = g(t-\tau)$$

- Only the difference between t and τ matters
- What happens to y(t)?

$$y(t) = \int_{t_0}^{t} g(t,\tau)u(\tau)d\tau$$

= $\int_{t_0}^{t} g(t-\tau)u(\tau)d\tau$
= $\int_{t_0}^{t} g(\tau)u(t-\tau)d\tau$
= $g(t)*u(t) \sim \text{Convolution integral}$
 $\hat{y}(s) = \hat{g}(s)\hat{u}(s)$

Proof of
$$\hat{y}(s) = \hat{g}(s) \hat{u}(s)$$

 $\hat{y}(s) = \int_{0}^{\infty} y(t)e^{-st}dt$
 $= \int_{t=0}^{\infty} \left(\int_{\tau=0}^{\infty} g(t-\tau)u(\tau)d\tau\right)e^{-st}dt$
 $= \int_{t=0}^{\infty} \left(\int_{\tau=0}^{\infty} g(t-\tau)u(\tau)d\tau\right)e^{-s(t-\tau)}e^{-s\tau}dt$
 $= \int_{\tau=0}^{\infty} \left(\int_{t=0}^{\infty} g(t-\tau)e^{-s(t-\tau)}dt\right)u(\tau)e^{-s\tau}d\tau,$
(Let $v = t - \tau$)

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$$\hat{y}(s) = \int_{\tau=0}^{\infty} \left(\int_{\upsilon=-\tau}^{\infty} g(\upsilon) e^{-s\upsilon} d\upsilon \right) u(\tau) e^{-s\tau} d\tau,$$
(Note $g(\upsilon) = 0$ for $\upsilon < 0$)
$$= \int_{\tau=0}^{\infty} \left(\int_{\upsilon=0}^{\infty} g(\upsilon) e^{-s\upsilon} d\upsilon \right) u(\tau) e^{-s\tau} d\tau$$

$$= \left(\int_{\upsilon=0}^{\infty} g(\upsilon) e^{-s\upsilon} d\upsilon \right) \left(\int_{\tau=0}^{\infty} u(\tau) e^{-s\tau} d\tau \right)$$

$$\hat{y}(s) = \hat{g}(s) \cdot \hat{u}(s)$$

Transfer-Function Matrix

- For SISO system, $\hat{y}(s) = \hat{g}(s) \cdot \hat{u}(s)$
- $\hat{g}(s) \sim$ Transfer function, the Laplace transform of the unit impulse response
 - For MIMO system,

$$\hat{y}(s) = \hat{G}(s) \cdot \hat{u}(s)$$

$$\hat{G}(s) = \begin{bmatrix} \hat{g}_{11}(s) & \hat{g}_{12}(s) & \hat{g}_{1p}(s) \\ \hat{g}_{21}(s) & \hat{g}_{22}(s) & \hat{g}_{2p}(s) \\ \hat{g}_{q1}(s) & \hat{g}_{q2}(s) & \hat{g}_{qp}(s) \end{bmatrix} \sim \text{Transfer-function matrix, or transfer matrix}$$

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State Variable Description

• Start with a general lumped (finite-dimensional) system:

 $\dot{\mathbf{x}}(t) = \mathbf{h}(\mathbf{x}(t), \mathbf{u}(t), t)$ $\mathbf{y}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t)$

• If the system is linear, the above reduces to:

 $\dot{x}(t) = A(t)x(t) + B(t)u(t)$ y(t) = C(t)x(t) + D(t)u(t)

• If the system is linear and time-invariant, then:

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

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 To find an LTI system's response to a particular input u(t), we can use Laplace transform:

$$\begin{split} s\hat{x}(s) - x_0 &= A\hat{x}(s) + B\hat{u}(s) \\ \hat{y}(s) &= C\hat{x}(s) + D\hat{u}(s) \end{split}$$

• Solve the above linear algebraic equations:

$$\hat{\mathbf{x}}(\mathbf{s}) = (\mathbf{s}\mathbf{I} - \mathbf{A})^{-1} \mathbf{B}\hat{\mathbf{u}}(\mathbf{s}) + (\mathbf{s}\mathbf{I} - \mathbf{A})^{-1} \mathbf{x}_{0}$$
$$\hat{\mathbf{y}}(\mathbf{s}) = \boxed{\mathbf{C}(\mathbf{s}\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D}}\hat{\mathbf{u}}(\mathbf{s}) + \mathbf{C}(\mathbf{s}\mathbf{I} - \mathbf{A})^{-1} \mathbf{x}_{0}$$
Transfer function matrix $\hat{\mathbf{G}}(\mathbf{s})$

x₀ is the information needed to determine x(t) and y(t) for t>0, in addition to the input u(t).

 \square x is the state

2.4 Linearization

- There are many results on linear systems while nonlinear systems are generally difficult to analyze
 - What to do with a nonlinear system described by

 $\dot{x}(t) = h(x(t), u(t), t)$ y(t) = f(x(t), u(t), t)

- Linearization. How? Under what conditions?
 - Using Taylor series expansion based on a nominal trajectory, ignoring second order terms and higher
 - Effects are not bad if first order Taylor series expansion is a reasonable approximation over the duration under consideration

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• Suppose that with x_o(t) and u_o(t), we have

 $\dot{\mathbf{x}}_{0}(t) = \mathbf{h}(\mathbf{x}_{0}(t), \mathbf{u}_{0}(t), t)$

- Suppose that the input is perturbed to $u_o(t) + \overline{u}(t)$
- Assume the solution is $x_0(t) + \overline{x}(t)$, with $\overline{x}(t)$ satisfying

 $\dot{\mathbf{x}}_{0}(t)$ + $\dot{\mathbf{x}}(t) = \mathbf{h}(\mathbf{x}_{0}(t) + \mathbf{\overline{x}}(t), \mathbf{u}_{0}(t) + \mathbf{\overline{u}}(t), t)$

$$\frac{\partial h}{\partial x} = \begin{pmatrix} \frac{\partial h_1}{\partial x_1} & \frac{\partial h_1}{\partial x_2} & \cdots & \frac{\partial h_1}{\partial x_n} \\ \frac{\partial h_2}{\partial x_1} & \frac{\partial h_2}{\partial x_2} & \cdots & \frac{\partial h_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \cdots & \frac{\partial h_n}{\partial x_n} \end{pmatrix}, \\ \frac{\partial h}{\partial u} = \begin{pmatrix} \frac{\partial h_1}{\partial u_1} & \frac{\partial h_1}{\partial u_2} & \cdots & \frac{\partial h_1}{\partial u_p} \\ \frac{\partial h_2}{\partial u_1} & \frac{\partial h_2}{\partial u_2} & \cdots & \frac{\partial h_2}{\partial u_p} \\ \vdots & \vdots & \cdots & \vdots \\ \frac{\partial h_n}{\partial u_1} & \frac{\partial h_n}{\partial u_2} & \cdots & \frac{\partial h_n}{\partial u_p} \end{pmatrix} \sim \text{Jacobians}$$

• Then the perturbed system can be described by

$$\dot{\mathbf{x}}(\mathbf{t}) = \frac{\partial \mathbf{h}}{\partial \mathbf{x}}\Big|_{\mathbf{0}} \mathbf{x} + \frac{\partial \mathbf{h}}{\partial \mathbf{u}}\Big|_{\mathbf{0}} \mathbf{u} \sim \mathbf{A} \text{ linear system}$$

- The above is valid if the first order Taylor series expansion works out well within the time duration under consideration. It may lead to wrong prediction.
- What to do with the output y(t) = f(x(t), u(t), t)?
 - The output equation can be similarly linearized, but most often there is no need for linearization unless with output feedback

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There is another approach to deal with nonlinear time-varying systems: Conservative but reliable



$$\dot{x}_{1} = h_{1}(x) = x_{2}$$

$$\dot{x}_{2} = h_{2}(x) = -\frac{g}{l} \sin x_{1} + \frac{1}{ml} \cos x_{1} \quad u$$
torque = force×arm
Linearize the system at x₁=0, x₂=0, u=0,
$$\frac{\partial h_{1}}{\partial x_{1}} = 0, \quad \frac{\partial h_{1}}{\partial x_{2}} = 1, \quad \frac{\partial h_{1}}{\partial u} = 0$$

$$\frac{\partial h_2}{\partial x_1}\Big|_0 = \left(-\frac{g}{l}\cos x_1 - \frac{1}{ml}\sin x_1\right)\Big|_0 = -\frac{g}{l}, \quad \frac{\partial h_2}{\partial x_2}\Big|_0 = 0, \quad \frac{\partial h_2}{\partial u}\Big|_0 = \frac{1}{ml}\cos x_1\Big|_0 = \frac{1}{30}\frac{1}{ml}$$

$$\frac{\partial h}{\partial x} = \begin{bmatrix} \frac{\partial h_1}{\partial x_1} & \frac{\partial h_1}{\partial x_2} \\ \frac{\partial h_2}{\partial x_1} & \frac{\partial h_2}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & 0 \end{bmatrix}, \quad \frac{\partial h}{\partial u} = \begin{bmatrix} \frac{\partial h_1}{\partial u} \\ \frac{\partial h_2}{\partial l_2} \\ \frac{\partial h_2}{\partial u} \end{bmatrix} = \begin{bmatrix} 0 \\ \frac{1}{ml} \end{bmatrix}$$
Linearized system:

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ \frac{1}{ml} \end{bmatrix} u$$
Exercise: Linearize the following system at x =0, u=0.

$$\begin{aligned} \dot{x}_1 &= x_2, \\ \dot{x}_2 &= -\frac{g}{l_1} \sin x_1 + \frac{m_2 g}{m_1 l_1} \cos x_3 \sin(x_3 - x_1) + \frac{1}{m_1 l_1} \sin x_3 \sin(x_3 - x_1) u \\ \dot{x}_3 &= x_4; \\ \dot{x}_4 &= -\frac{g}{l_2} \sin x_3 + \frac{1}{m_2 l_2} \cos x_3 u \end{aligned}$$



Let $\bar{i}(t) = \frac{1}{T} \int_{t}^{t+T} i(\tau) d\tau, \quad \bar{v}(t) = \frac{1}{T} \int_{t}^{t+T} v(\tau) d\tau, \quad \bar{x} = \begin{bmatrix} \bar{i} \\ \bar{v} \end{bmatrix}, \quad v = \begin{bmatrix} v_g \\ v_D \end{bmatrix}$ The averaged model is: $\dot{\bar{x}} = (DA_1 + (1-D)A_2)\bar{x} + (DB_1 + (1-D)B_2)v$ $\bar{y} = C\bar{x}$

Let the nominal working point be $D = D_0$, $\overline{x} = \overline{x}_0$, $\overline{y} = \overline{y}_0$

At steady state,
$$0 = (D_0 A_1 + (1 - D_0) A_2) \overline{x}_0 + (D_0 B_1 + (1 - D_0) B_2) v$$
(1)
$$\overline{y}_0 = C \overline{x}_0$$



At nominal working condition:

$$0 = (D_0 A_1 + (1 - D_0) A_2) \overline{x}_0 + (D_0 B_1 + (1 - D_0) B_2) v$$
(1)
$$\overline{y}_0 = C \overline{x}_0$$

To achieve robust stability and tracking, so that the same output y_0 is produced when parameters have changed, we obtain a perturbation model around the nominal working point:

Define $x_p = \overline{x} - \overline{x}_0$, $u = D - D_0$, $y = \overline{y} - \overline{y}_0$

$$\dot{x}_p = \overline{A}x_p + \overline{A}_b x_p u + \overline{B}u, \quad y = Cx_p \quad \overline{A} = D_0 A_1 + (1 - D_0) A_2, \overline{A}_b = A_1 - A_2$$
$$\overline{B} = (A_1 - A_2)\overline{x}_0 + (B_1 - B_2)v$$

This is obtained by subtracting (1) from the averaged model:

$$\dot{\overline{x}} = (DA_1 + (1-D)A_2)\overline{x} + (DB_1 + (1-D)B_2)v$$

$$\overline{y} = C\overline{x}$$

If the perturbation is small, $x_p u$ can be ignored as a second-order term The approximate linear model is

$$\dot{x}_p = \overline{A}x_p + \overline{B}u, \ y = Cx_p$$
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Linear Differential Inclusion (LDI)

An LTI system:

$\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t)$	 [x]	_	A	В	x	
$\mathbf{y}(t) = \mathbf{C}\mathbf{x}(t) + \mathbf{D}\mathbf{u}(t)$	у	=	С	D	u	

In many situations, A,B,C,D are not constant, but nonlinear time varying , and/or depend on a parameter α , such as,

 $\dot{\mathbf{x}}(t) = \mathbf{A}(\mathbf{x}, \alpha, t)\mathbf{x}(t) + \mathbf{B}(\mathbf{x}, \alpha, t)\mathbf{u}(t)$ $\mathbf{y}(t) = \mathbf{C}(\mathbf{x}, \alpha, t)\mathbf{x}(t) + \mathbf{D}(\mathbf{x}, \alpha, t)\mathbf{u}(t)$

We can find a set Ω such that $\begin{bmatrix} A(x, \alpha, t) & B(x, \alpha, t) \\ C(x, \alpha, t) & D(x, \alpha, t) \end{bmatrix} \in \Omega$

The system satisfies $\begin{bmatrix} \dot{x} \\ y \end{bmatrix} \in \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix}$: $\begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \Omega \right\}$

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\begin{bmatrix} \dot{x} \\ y \end{bmatrix} \in \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix} : \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \Omega \right\}
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- This is a linear differential inclusion (LDI)
- An LDI uses a set of linear systems to describe a complicated nonlinear system.
- In many cases Ω is a polytope: the behavior of an LDI can be characterized by finite many linear systems, e.g.,

 $\dot{\mathbf{x}}(t) = \mathbf{A}_i \mathbf{x}(t) + \mathbf{B}_i \mathbf{u}(t)$ $\mathbf{y}(t) = \mathbf{C}_i \mathbf{x}(t) + \mathbf{D}_i \mathbf{u}(t), \quad i = 1, \dots, N$

 Like a polygon, its properties are determined by finite many vertices.



Example: A model for a pendulum

$$\dot{x}_{1} = h_{1}(x) = x_{2}$$
$$\dot{x}_{2} = h_{2}(x) = -\frac{g}{l} \sin x_{1} + \frac{1}{ml} \cos x_{1} \quad u$$
$$\dot{x} = \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} \frac{\sin x_{1}}{x_{1}} & 0 \end{bmatrix} \begin{bmatrix} x_{1} \\ x_{2} \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{1}{ml} \cos x_{1} \end{bmatrix} u = A(x_{1})x + B(x_{1})u$$

If the angle is restricted between 0 and $\pi/4$, we can write

$$\dot{x} \in \left\{ \begin{bmatrix} A(x_1), & B(x_1) \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix} : x_1 \in [0, \pi/4] \right\}$$

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2.5 Modeling of Selected Systems

- We will briefly go over the following systems
 - Electrical Circuits
 - Operational Amplifiers
 - Mechanical Systems
 - Integrator/Differentiator Realization
- For any of the above system, we derive a state space description:

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

Different engineering systems are unified into the same framework, to be addressed by system and control theory.

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Electrical Circuits

- i of L and v of C

State variables?



• How to describe the evolution of the state variables?

Steps to obtain state and output equations:

Step 1: Pick $\{i_L, v_C\}$ as state variables

Step 2:
$$L \frac{di_L}{dt} = v_L = f_1(i_L, v_C, u)$$
 Linear functions
 $C \frac{dv_C}{dt} = i_C = f_2(i_L, v_C, u)$ By using KVL and KCL
Step 3: $\frac{di_L}{dt} = (1/L)f_1(i_C, v_L, u)$
 $\frac{dv_C}{dt} = (1/C)f_2(i_C, v_L, u)$

Step 4: Put the above in matrix form

Step 5: Do the same thing for y in terms of state variables and input, and put in matrix form







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Operational Amplifiers (Op Amps)

 $v_{0} = A (v_{a} - v_{b}), \text{ with } -V_{CC} \le v_{0} \le V_{CC}$ Non-inverting terminal, v_{a} $i_{a} \longrightarrow i_{o}$ Inverting terminal, v_{b} V_{cc} , -15V V_{cc} , -15V V_{cc} V_{cc} V_{cc}

- Usually, $A > 10^4$
- Ideal Op Amp:
 - A $\rightarrow \infty \sim$ Implying that $(v_a v_b) \rightarrow 0$, or $v_a \rightarrow v_b$
 - $i_a \rightarrow 0$ and $i_b \rightarrow 0$
- Problem: How to analyze a circuit with ideal Op Amps



Delineate the relationship between input and output

Input/Output description

• Key ideas:

Pure gain, no SVs

- Make effective use of $i_a = i_b = 0$ and $v_a = v_b$
- Do not apply the node equation to output terminals of op amps and ground nodes, since the output current and power supply current are generally unknown



> Today:

- Math. descriptions of systems
- Modeling of electric circuit

Next Time:

- Modeling of Selected Systems
 - Continuous-time systems (§2.5)
 - mechanical systems, integrator/differentiator realization
 - Discrete-Time systems (§2.6)
 - difference equations, simple financial systems
- Advanced Linear Algebra, Chapter 3

Problem Set #2:

- 1. Give examples for nonlinear systems and infinite dimensional systems respectively. What are the inputs, outputs and states?
- Suppose we have a linear time-invariant system. Its response to u₁ is y₁(t)= t+3, for t≥ 0, and its response to u₂ is y₂(t)= 2t, for t≥ 0. For t< 0, y₁(t) = y₂(t)=0. Assume zero initial conditions. What is the response to 2u₁(t-1)-u₂(t+1)? Plot the response for t∈[-2,4] with Matlab.
- 3. An LTI system is described by ÿ+4ÿ+3y = u, What is y(t) for u=0 and y(0)=1, y'(0)=-1? What is y(t) for a unit step u (u=1(t)) and y(0)=y'(0)=0? What is y(t) for u=1(t) and y(0)=2, y'(0)=-2? What is the state of the system?



