

Linear System and State Space Formulation

Chen Ch. 2-5

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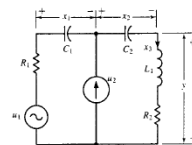
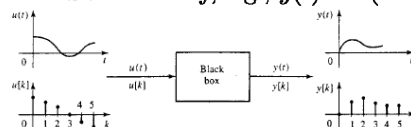
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Basic Concepts

- **System:** a collection of elements, that interacts with its environment via a set of input $u \in \mathbb{R}^m$ and output $y \in \mathbb{R}^p$.
 - Causal system (current output depends only on past/current inputs, not future (e.g., $y = \frac{1}{s}u$) vs acausal system (e.g., $y = su$).
 - Static (memoryless) system $y(t) = h(u(t))$ vs dynamic (with memory) system $y(t) = h(t, x(t_o), u([t_o; t]))$.
- **State:** $x(t_1) \in \mathbb{R}^n$ of a causal system at t_1 is the information needed, together with the input $u : [t_1, t_2]$, to uniquely define the output y at t_2 .
 - State $x(t_1)$ at t_1 contains all the information related to the past input history of $u : [-\infty, t_1]$ to define $y(t)$, $\forall t \geq t_1$.
 - State is not unique (e.g., $x_1 + x_2, x_1 - x_2, x_3$).
 - Dimension of system = dimension of minimal state vector $x \in \mathbb{R}^n$.
- Finite-dimensional system (only finite SV needed) vs infinite-dimensional system (infinite SV necessary, e.g., $y(t) = u(t - \tau)$).



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Dynamical Systems

- State-space system representation:

(Continuous-time): $\dot{x}(t) = f(x(t), u(t), t), \quad y(t) = h(x(t), u(t), t)$

(Discrete-time): $x(k+1) = f(x(k), u(k), k), \quad y(k) = h(x(k), u(k), k)$

where $t \in \mathbb{R}$ is time, $k \in \mathbb{Z}$ is discrete-time index, $x \in \mathbb{R}^n$ is state vector, $u \in \mathbb{R}^p$ is input vector, and $y \in \mathbb{R}^m$ output vector, with f, h being state dynamics map and output map.

- **State transition map:** given $x_o = x(t_o)$ and $u([t_o, t_1])$, $x(t_1)$ is uniquely determined, i.e.,

$$x(t_1) = s(t_1, t_o, x_o, u([t_o; t_1]))$$

- Semi-group property: $\forall t_2 \geq t_1 \geq t_o, x_o \in \mathbb{R}^n$ and $u([t_o; t_2])$,

$$\begin{aligned} s(t_2, t_o, x_o, u([t_o; t_2])) &= s(t_2, t_1, s(t_1, t_o, x_o, u([t_o; t_1])), u([t_1; t_2])) \\ &= s(t_2, t_1, x_1, u([t_1; t_2])) \end{aligned}$$

- Ex) solution by integration: $x(t_1) = x(t_o) + \int_{t_o}^{t_1} f(x(\tau), u(\tau), \tau) d\tau$.

- Output map: $y(t) = h(s(t, t_o, x_o, u([t_o; t])), u(t), t) = h'(t, t_o, x_o, u([t_o; t]))$.

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Linear Dynamical System

- Linear system property: given $(x_i(t_o), u_i([t_o; t])) \rightarrow (x_i(t), y_i(t)), t \geq t_o$,

- $\alpha(x_i(t_o), u_i([t_o; t])) \rightarrow \alpha(x_i(t), y_i(t))$ for constant α (homogeneity).

- $(x_1(t_o) + x_2(t_o), u_1 + u_2) \rightarrow (x_1(t) + x_2(t), y_1(t) + y_2(t))$ (additivity).

- **Superposition property:** for any real constants α, β ,

$$(\alpha x_1(t_o) + \beta x_2(t_o), \alpha u_1 + \beta u_2) \rightarrow (\alpha x_1 + \beta x_2, \alpha y_1 + \beta y_2)$$

- Note the scaling applies both to IC and input.

- For **linear dynamics** $\dot{x} = f(x, u, t)$ (i.e., $f(\alpha x_1 + \beta x_2, \alpha u_1 + \beta u_2, t) = \alpha f(x_1, u_1, t) + \beta f(x_2, u_2, t)$), we have linear state-transition map:

$$s(t, t_o, \alpha x_1^o + \beta x_2^o, \alpha u_1 + \beta u_2) = \alpha s(t, t_o, x_1^o, u_1) + \beta s(t, t_o, x_2^o, u_2)$$

Further, with **linear output map** (i.e., $h(\alpha x_1 + \beta x_2, \alpha u_1 + \beta u_2, t) = \alpha h(x_1, u_1, t) + \beta h(x_2, u_2, t)$),

$$\begin{aligned} h(s(t, t_o, \alpha x_1^o + \beta x_2^o, \alpha u_1 + \beta u_2), t) &= y(t, t_o, \alpha x_1^o + \beta x_2^o, \alpha u_1 + \beta u_2) \\ &= \alpha y(t, t_o, x_1^o, u_1) + \beta y(t, t_o, x_2^o, u_2) \end{aligned}$$

- LTV or LTI system: $\dot{x} = A(t)x + B(t)u, \quad y = C(t)x + D(t)u$.

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LTV and LTI Systems

- State-space representation of linear systems:

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

where $A(t) \in \mathbb{R}^{n \times n}$, $B(t) \in \mathbb{R}^{n \times p}$, $C(t) \in \mathbb{R}^{m \times n}$, $D(t) \in \mathbb{R}^{m \times p}$ are state, input, output and direct feedthrough matrices. If these are constants \Rightarrow linear time-invariant (LTI); else \Rightarrow linear time-varying (LTV).

- Jacobian linearization** of $\dot{x} = f(x, u, t)$, $y = h(x, u, t)$ around nominal trajectory $\bar{x} = f(\bar{x}, \bar{u}, t)$, $\bar{y} = h(\bar{x}, \bar{u}, t)$ or equilibrium (with $\dot{\bar{x}} = 0$).

– Define deviations $\tilde{x} := x - \bar{x}$, $\tilde{u} := u - \bar{u}$, $\tilde{y} := y - \bar{y}$. Then,

$$\dot{\tilde{x}} = \dot{x} - \dot{\bar{x}} = f(\bar{x} + \tilde{x}, \bar{u} + \tilde{u}, t) - f(\bar{x}, \bar{u}, t) = \left. \frac{\partial f}{\partial x} \right|_{\bar{x}(t), \bar{u}(t)} \tilde{x} + \left. \frac{\partial f}{\partial u} \right|_{\bar{x}(t), \bar{u}(t)} \tilde{u}$$

$$\tilde{y} = y - \bar{y} = h(\bar{x} + \tilde{x}, \bar{u} + \tilde{u}, t) - h(\bar{x}, \bar{u}, t) = \left. \frac{\partial h}{\partial x} \right|_{\bar{x}(t), \bar{u}(t)} \tilde{x} + \left. \frac{\partial h}{\partial u} \right|_{\bar{x}(t), \bar{u}(t)} \tilde{u}$$

– Ex) $\dot{x}_1 = x_2$, $\dot{x}_2 = -\frac{g}{L} \sin x_2$; $\dot{x} = t \sin x + u^2 \sin t$, $y = \cos tx + e^{-t}u$.

- Realization of $Y(s) = H(s)U(s)$. Then, $H(s) = H_{sp}(s) + H(\infty)$ with $D = H(\infty)$. Further, $H_{sp}(s) = \frac{1}{D(s)}[N_1 s^{n-1} \dots + N_n]$. Then, can obtain A, B, C, D via $Y(s) = [N_1 s^{n-1} + \dots + N_n]X(s)$ and $X(s) = \frac{1}{D(s)}U(s)$.

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Linear System Response

- Consider linear system $\dot{x} = A(t)x + B(t)u$ with superposition property:

$$s(t, t_o, \alpha x_1^o + \beta x_2^o, \alpha u_1 + \beta u_2) = \alpha s(t, t_o, x_1^o, u_1) + \beta s(t, t_o, x_2^o, u_2)$$

- Then, we can write the state response $x(t)$ as composed of:

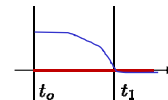
$$x(t) = s(t, t_o, x_o, u) = s(t, t_o, x_o + 0, 0 + u) = s(t, t_o, x_o, 0) + s(t, t_o, 0, u)$$

- **zero-input response** $x_{u=0, x_o}(t) = s(t, t_o, x_o, 0)$ (i.e., response $x(t)$ only due to IC x_o with no input) and
- **zero-state response** $x_{u, x_o=0}(t) = s(t, t_o, 0, u)$ (i.e., response only due to input u with zero IC).
- Recall SISO TF case: $Y(s) = H(s)U(s) + H(s)[IC]$.

- Fundamental matrix:** A matrix $X(t) \in \mathbb{R}^{n \times n}$ is a fundamental matrix of $\dot{x} = A(t)x$, if

$$\dot{X}(t) = A(t)X(t)$$

with non-singular $X(t_o)$ for some t_o .



- $X(t)$ is non-singular $\forall t$: if singular at $t_1 > t_o \rightarrow \exists$ non-zero ν s.t., $x(t_1) := X(t_1)\nu = 0$. Then, $x(t) = X(t)\nu$ is a solution, thus, w/ $t \rightarrow t_o$, $x(t_o) = X(t_o)\nu = 0 \rightarrow$ contradiction (no information collapse).

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Transition Matrix

- Consider the LTV system

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

with zero-input response $s(t, t_o, x_o, 0)$. Then, since linear (e.g., using $(X(t), X(t_o))$), it can be represented by **transition matrix** $\Phi(t, t_o) \in \mathbb{R}^{n \times n}$:

$$s(t, t_o, x_o, 0) = \Phi(t, t_o)x_o$$

- $\Phi(t, t_o) = X(t)X^{-1}(t_o)$ with $\Phi(t_o, t_o) = I$.
 - $\frac{\partial}{\partial t}\Phi(t, t_o) = A(t)\Phi(t, t_o)$ (i.e., Φ is a fundamental matrix).
 - $\Phi(t, t_o) = \Phi(t, t_1)\Phi(t_1, t_o)$.
 - $\Phi^{-1}(t, t_o) = \Phi(t_o, t)$.
 - $\Phi(t, t_o)$ is non-singular $\forall t \geq t_o$ (i.e., no information collapse).
 - $\frac{\partial}{\partial t_o}\Phi(t, t_o) = -\Phi(t, t_o)A(t_o)$.
- LTI system: $\frac{d}{dt}\Phi(t, t_o) = A\Phi(t, t_o), \Phi(t_o, t_o) = I \rightarrow \Phi(t, t_o) = e^{A(t-t_o)}$.
- LTV system: $\frac{d}{dt}\Phi(t, t_o) = A(t)\Phi(t, t_o) \rightarrow \Phi(t, t_o) \neq e^{\int_{t_o}^t A(\tau)d\tau}$.
with $e^{At} = I + At + \frac{A^2 t^2}{2!} + \dots$

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Formulas for Transition Matrix

- Consider transition matrix $\Phi(t, t_o)$ of LTV system $\dot{x} = A(t)x + B(t)u$ with

$$\frac{\partial}{\partial t}\Phi(t, t_o) = A(t)\Phi(t, t_o), \quad \Phi(t_o, t_o) = I$$

- Picard iteration:**

- Implicit integration equation: $\Phi(t, t_o) = I + \int_{t_o}^t A(\tau)\Phi(\tau, t_o)d\tau$.
- k -th iteration: $\Phi_{k+1}(t, t_o) = I + \int_{t_o}^t A(\tau)\Phi_k(\tau, t_o)d\tau$ w/ $\Phi_1(t, t_o) = I$.
- Peano-Baker formula:

$$\Phi(t, t_o) = I + \int_{t_o}^t A(\tau_1)d\tau_1 + \int_{t_o}^t A(\tau_1) \left[\int_{t_o}^{\tau_1} A(\tau_2)d\tau_2 \right] d\tau_1 + \dots$$

- For the LTI system with $\dot{x} = Ax + Bu$,

- Peano-Baker formula:

$$\Phi(t, t_o) = I + \frac{A(t-t_o)}{1!} + \frac{A^2(t-t_o)^2}{2!} + \dots = e^{A(t-t_o)}$$

- Laplace transform: $s\Phi(s) - I = A\Phi(s) \rightarrow \Phi(s) = (sI - A)^{-1}$.

$$-A = \begin{bmatrix} -1 & 0 \\ 1 & -2 \end{bmatrix} \rightarrow (sI - A)^{-1} = \begin{bmatrix} \frac{1}{s+1} & 0 \\ \frac{1}{(s+1)(s+2)} & \frac{1}{s+2} \end{bmatrix} \rightarrow \Phi(t, 0)$$

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Jordan Form and Multiplicity

- Consider LTI system $\dot{x} = Ax + Bu$ with transition matrix

$$\Phi(t, 0) = e^{At} = I + \frac{At}{1!} + \frac{A^2 t^2}{2!} + \frac{A^3 t^3}{3!} + \dots$$

- Recall the eigen-problem of A , i.e., $A\nu_i = \lambda_i \nu_i$, where $\lambda_i \in \mathbb{C}$ and $\nu_i \in \mathbb{C}^n$ are eigenvalue and eigenvector. Then, for each λ_i ,

algebraic multiplicity of $\lambda_i \geq$ geometric multiplicity of λ_i

i.e., order of $(s - \lambda_i)$ in CE \geq number of independent eigenvectors.

- Jordan form:** for $A \in \mathbb{R}^{4 \times 4}$ with deficient λ_2 ,

$$A = T \begin{bmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 1 & 0 \\ 0 & 0 & \lambda_2 & 0 \\ 0 & 0 & 0 & \lambda_4 \end{bmatrix} T^{-1} \Rightarrow e^{At} = T \begin{bmatrix} e^{\lambda_1 t} & 0 & 0 & 0 \\ 0 & e^{\lambda_2 t} & t e^{\lambda_2 t} & 0 \\ 0 & 0 & e^{\lambda_2 t} & 0 \\ 0 & 0 & 0 & e^{\lambda_4 t} \end{bmatrix} T^{-1}$$

from $AT = \Lambda T$, where $T \in \mathbb{R}^{4 \times 4}$ is collection of (generalized) eigenvectors.

- Stable if all $\lambda_i \in \text{LHP}$; marginally stable if all $\lambda_i \in \text{LHP}$ except some non-deficient λ_i on jw -axis; unstable if some $\lambda_i \in \text{RHP}$ or deficient λ_i on jw -axis (if deficiency = 2 \rightarrow growth w/ t^2 : $[0, 0; 0, 0]$, $[0, 1; 0, 0]$).

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Similarity Transformation

- Consider the LTI system:

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

- Define coordinate transformation via $x = T\bar{x}$ with non-singular matrix T . We may then think of \bar{x}_i as the component of x along T_i .

- Similarity transformation:** with $x = T\bar{x}$, can transform the dynamics s.t.,

$$\dot{\bar{x}} = \bar{A}\bar{x}(t) + \bar{B}u(t), \quad y = \bar{C}\bar{x}(t) + Du$$

where $\bar{A} = T^{-1}AT$, $\bar{B} = T^{-1}B$, $\bar{C} = CT$. This similarity transformation is equivalent with:

- Same eigenvalues: $\text{CE}(\bar{A}) = \det[\lambda I - \bar{A}] = \det[T^{-1}] \det[\lambda I - A] \det[T] = \det[\lambda I - A] = \text{CE}(A)$
- Same transfer function: $\bar{H}(s) = \bar{C}(sI - \bar{A})^{-1}\bar{B} + D = \bar{C}T(sI - T^{-1}AT)^{-1}T^{-1}B + D = C(SI - A)^{-1}B + D = H(s)$.
- Other properties (e.g., controllability, observability).
- If all eigenvalues are distinct, we may then achieve **modal decomposition**: with $x = Tz$, T collection of eigenvectors,

$$\dot{z}_i = \lambda_i z_i + \bar{B}_i u, \quad \text{with mode } T_i \in \mathbb{R}^n \text{ and modal freq. } \lambda_i$$

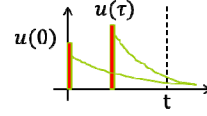
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Zero-State Response and Convolution

- Consider LTV system:

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



with state response given by zero-input and zero-state responses:

$$x(t) = \Phi(t, t_o)x(t_o) + s(t, t_o, 0, u([t_o; t]))$$

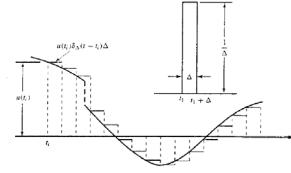
- To compute the zero-state response, using the unit impulse $\delta_\Delta(t - t_k)$, write

$$u(t) \approx \sum_k u(t_k) \delta_\Delta(t - t_k) \Delta$$

- Then, from the linearity of state-transition map s ,

$$s(t, t_o, 0, u) \approx \sum_k s(t, t_o, 0, u(t_k) \delta_\Delta(t - t_k)) \Delta$$

$$= \sum_k \Phi(t, t_k) B(t_k) u(t_k) \Delta \rightarrow \int_{t_o}^t \Phi(t, \tau) B(\tau) u(\tau) d\tau \quad (\text{as } \Delta \rightarrow 0)$$



since, with zero-state condition (i.e., $x(t_k) = 0$), $x(t_{k+1}) = \int_{t_k}^{t_{k+1}} [A(t)x(t) + B(t)u(t_k)\delta_\Delta(t - t_k)] dt = B(t_k)u(t_k)\Delta$, which will propagate to t via state transition matrix $\Phi(t, t_{k+1})$.

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Zero-State Response: Proof

- Consider LTV system:

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

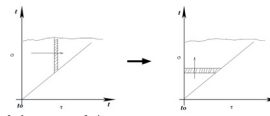
and denote by $x(t)$ as the zero-state response under $u(t)$ with $x(t_o) = 0$.

We want to show this $x(t) = s(t, t_o, 0, u)$ is the same as

$$z(t) := \int_{t_o}^t \Phi(t, \tau) B(\tau) u(\tau) d\tau$$

- Note first that $z(t_o) = 0$. Also, using $\Phi(t, \tau) = I + \int_\tau^t A(\sigma) \Phi(\sigma, \tau) d\sigma$,

$$\begin{aligned} z(t) &= \int_{t_o}^t \left[I + \int_\tau^t A(\sigma) \Phi(\sigma, \tau) d\sigma \right] B(\tau) u(\tau) d\tau \\ &= \int_{t_o}^t B(\tau) u(\tau) d\tau + \int_{t_o}^t \int_\tau^t A(\sigma) \Phi(\sigma, \tau) d\sigma B(\tau) u(\tau) d\tau \\ &= \int_{t_o}^t B(\tau) u(\tau) d\tau + \int_{t_o}^t A(\sigma) \int_{t_o}^\sigma \Phi(\sigma, \tau) B(\tau) u(\tau) d\tau d\sigma \\ &= \int_{t_o}^t B(\tau) u(\tau) d\tau + \int_{t_o}^t A(\sigma) z(\sigma) d\sigma \end{aligned}$$



implying $\dot{z} = A(t)z(t) + B(t)u(t)$ with $z(t_o) = 0 \Rightarrow z(t) = x(t)$.

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Output Response

- Consider LTV system:

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

with total state-response given by:

$$x(t) = s(t, t_o, x_o, u) = \Phi(t, t_o)x_o + \int_{t_o}^t \Phi(t, \tau)B(\tau)u(\tau)d\tau$$

- Output response is then given by:

$$y(t) = C(t)\Phi(t, t_o)x_o + C(t) \int_{t_o}^t \Phi(t, \tau)B(\tau)u(\tau)d\tau + D(t)u(t)$$

which is linear (i.e., $\alpha(x_1 + u_1) + \beta(x_2 + u_2) \rightarrow \alpha y_1 + \beta y_2$), and consists of free-response (zero-input output) and force-response (zero-state output).

- LTI system: with $\Phi(t, t_o) = e^{A(t-t_o)}$, and $t_o = 0$ and $x(0) = x_o$,

$$y(t) = Ce^{At}x_o + C \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau + Du(t)$$

$$Y(s) = C(sI - A)^{-1}x_o + [C(sI - A)^{-1}B + D]U(s)$$

where $H(s) = [C(sI - A)^{-1}B + D]$ is the **transfer matrix**.

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Impulse Response and Transfer Matrix

- Consider LTV system with its output response:

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

$$y(t) = C(t)\Phi(t, t_o)x_o + C(t) \int_{t_o}^t \Phi(t, \tau)B(\tau)u(\tau)d\tau + D(t)u(t)$$

- Impulse response matrix:** with zero initial state and unit-impulse input through the j -th input channel $u_j(t) = [0; \dots; \delta(t - t'); 0; \dots 0]$,

$$y_j(t) = [C(t)\Phi(t, t')B(t') + D(t)\delta(t - t')]_j \in \Re^m, \quad t \geq t'$$

i.e., $y_j(t)$ is a m -dimensional column vector. By collecting this, we can construct $m \times p$ impulse response matrix:

$$h(t, t') = C(t)\Phi(t, t')B(t') + \delta(t - t') \cdot D(t)$$

whose j -th column is output from j -th channel unit-impulse input.

- LTI system:** with $t' = 0$, $h_{\text{impulse}}(t) = Ce^{At}B + \delta(t) \cdot D$. Taking Laplace TF,

$$H_{\text{impulse}}(s) = C(sI - A)^{-1}B + D$$

which is transfer matrix computed before (or directly from $\dot{x} = Ax + Bu, y = Cx + Du$) \Rightarrow transfer matrix is Laplace TF of IR-matrix.

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DT-LTV System and Exact Discretization

- Consider LTV discrete-time system:

$$x(k+1) = A(k)x(k) + B(k)u(k), \quad y(k) = C(k)x(k) + D(k)u(k)$$

where $k \in \mathcal{Z}$ is the time-index, $x(k) \in \mathbb{R}^n$, $u(k) \in \mathbb{R}^p$, $y(k) \in \mathbb{R}^m$, $A(k) \in \mathbb{R}^{n \times n}$, $B(k) \in \mathbb{R}^{n \times p}$, $C(k) \in \mathbb{R}^{m \times n}$ and $D(k) \in \mathbb{R}^{m \times p}$.

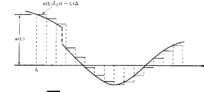
- Simple discretization: with $\dot{x}(t) \approx (x((k+1)T) - x(kT))/T$,

$$x(k+1) = (1 + TA)x(k) + TB(k)u(k), \quad y(k) = Cx(k) + Du(k)$$

- Exact discretization w/ piecewise-continuous input $u(k)$:

– Response of CT-LTI system: $x(k) = e^{A_k T} x(0) + \int_0^{kT} e^{A(kT-\tau)} B u(\tau) d\tau$.

$$\begin{aligned} x(k+1) &= e^{A(k+1)T} x(0) + \int_0^{(k+1)T} e^{A((k+1)T-\tau)} B u(\tau) d\tau \\ &= e^{AT} x(k) + \int_{kT}^{(k+1)T} e^{A((k+1)T-\tau)} B u(\tau) d\tau = e^{AT} x(k) + \int_0^T e^{A\tau} d\tau B u(k) \end{aligned}$$



– $x(k+1) = A_d x(k) + B_d u(k)$, $y(k) = Cx(k) + Du(k)$, with $A_d = e^{AT}$, $B_d = \int_0^T e^{A\tau} d\tau B$. Also, $B_d = A^{-1}(A_d - I)B$ if A is nonsingular.

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DT-LTV System State Response

- LTV discrete-time system:

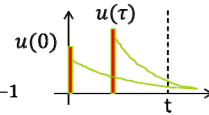
$$x(k+1) = A(k)x(k) + B(k)u(k), \quad y(k) = C(k)x(k) + D(k)u(k)$$

- The system is still linear with

$$x(k_1) = s(k_1, k_o, x_o, u([k_o; k_1-1])) = s(k_1, k_o, x_o, 0) + s(k_1, k_o, 0, u([k_o; k_1-1]))$$

- State-propagation of LTV-DT system:

$$\begin{aligned} x_k &= A_{k-1}x_{k-1} + B_{k-1}u_{k-1} \\ &= A_{k-1}A_{k-2}x_{k-2} + A_{k-1}B_{k-2}u_{k-2} + B_{k-1}u_{k-1} \\ &\vdots \\ &= [\prod_{i=k_o}^{k-1} A(i)] x(k_o) + \sum_{i=k_o}^{k-1} [\prod_{j=i+1}^{k-1} A(j)] B(i)u(i) \\ &= \Phi(k, k_o)x(k_o) + \sum_{i=k_o}^{k-1} \Phi(k, i+1)B(i)u(i) \end{aligned}$$



consisting of zero-input response w/ state transition matrix $\Phi(k, k_o) = A(k-1)A(k-2)\dots A(k_o) \in \mathbb{R}^{n \times n}$ and zero-state responses (i.e., convolution).

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DT-LTV System Response

- LTV discrete-time system:

$$x(k+1) = A(k)x(k) + B(k)u(k), \quad y(k) = C(k)x(k) + D(k)u(k)$$

- LTV-DT system state response:

$$x(k) = \Phi(k, k_o)x(k_o) + \sum_{i=k_o}^{k-1} \Phi(k, i+1)B(i)u(i), \quad \Phi(k, k_o) := \prod_{i=k_o}^{k-1} A(i)$$

- LTV-DT system output response:

$$\begin{aligned} y(k) &= C(k)\Phi(k, k_o)x_o + C(k) \sum_{i=k_o}^{k-1} \Phi(k, i+1)B(i)u(i) + D(k)u(k) \\ &= C(k) \sum_{i=k_o}^{k-1} \Phi(k, i+1)B(i)u(i) + D(k)u(k) \quad (\text{if } x_o = 0) \end{aligned}$$

$$= \sum_{i=k_o}^k [C(k)\Phi(k, i+1)B(i) + \delta(k-i)D(i)] u(i) = \sum_{i=k_o}^k H(k, i)u(i)$$

where $H(k, i) = 0 \forall k < i$, and $H(k, i) := C(k)\Phi(k, i+1)B(i) + D(i)\delta(k-i)$ is impulse response matrix with unit-impulse $\delta(k-i)$ at $i = k$.

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DT-LTV System Transition Matrix

- DT LTV system:

$$x(k+1) = A(k)x(k) + B(k)u(k), \quad y(k) = C(k)x(k) + D(k)u(k)$$

- LTV-DT system state response:

$$x(k) = \Phi(k, k_o)x(k_o) + \sum_{i=k_o}^{k-1} \Phi(k, i+1)B(i)u(i), \quad \Phi(k, k_o) := \prod_{i=k_o}^{k-1} A(i)$$

- DT-LTV state-transition matrix:

$$\Phi(k, k_o) := \prod_{i=k_o}^{k-1} A(i) = A(k-1)A(k-2)\dots A(k_o+1)A(k_o)$$

- $\Phi(k_1, k_o)$ exists and is unique $\forall k_1 \geq k_o$.
- $\Phi(k+1, k_o) = A(k)\Phi(k, k_o)$, $\Phi(k_1, k-1) = \Phi(k_1, k)A(k-1)$.
- If $k_1 < k_o$, $\Phi(k_1, k_o)$ exists & unique iff $A(k)$ invertible $\forall k_1 \leq k < k_o$.
- $\Phi^{-1}(k_1, k_o)$ exists and is given by

$$\Phi^{-1}(k_1, k_o) = A^{-1}(k_o)A^{-1}(k_o+1)\dots A^{-1}(k_1-1)$$

iff $A(k)$ invertible $\forall k_o \leq k \leq k_1-1$.

- $\Phi(k_2, k_o) = \Phi(k_2, k_1)\Phi(k_1, k_o)$ only for $k_o \leq k_1 \leq k_2$; $\forall k_o, k_1, k_2$ iff $A(k)$ invertible.

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DT-LTI System Response

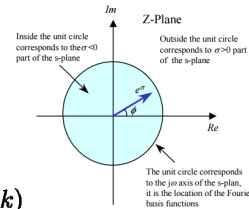
- DT-LTI system:

$$x(k+1) = Ax(k) + Bu(k), \quad y(k) = Cx(k) + Du(k)$$

- DT-LTI system response: with $\Phi(k,0) := \prod_{i=0}^{k-1} A = A^k$,

$$x(k) = A^k x(0) + \sum_{i=0}^{k-1} A^{k-i-1} B(i) u(i)$$

$$y(k) = CA^k x(0) + \sum_{i=0}^{k-1} CA^{k-i-1} Bu(i) + Du(k)$$



- Jordan form:

$$A = T \begin{bmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 1 & 0 \\ 0 & 0 & \lambda_2 & 0 \\ 0 & 0 & 0 & \lambda_4 \end{bmatrix} T^{-1} \Rightarrow A^k = T \begin{bmatrix} \lambda_1^k & 0 & 0 & 0 \\ 0 & \lambda_2^k & k\lambda_2^{k-1} & 0 \\ 0 & 0 & \lambda_2^k & 0 \\ 0 & 0 & 0 & \lambda_4^k \end{bmatrix} T^{-1}$$

stable if all $|\lambda_i| < 1$; marginally stable if all $|\lambda_i| < 1$ with some $|\lambda_i| = 1$, yet, not deficient; unstable if some $|\lambda_i| > 1$ or some deficient $|\lambda_i| = 1$.

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DT-LTI System Response and z-Transform

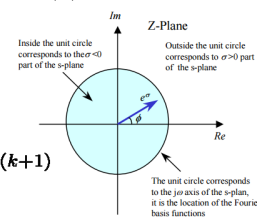
- DT-LTI system:

$$x(k+1) = Ax(k) + Bu(k), \quad y(k) = Cx(k) + Du(k)$$

- z-transform: $X(z) := \mathcal{Z}[x(k)] = \sum_{k=0}^{\infty} x(k)/z^k$.

- z-transform of $x(k+1)$:

$$\begin{aligned} \mathcal{Z}[x(k+1)] &= \sum_{k=0}^{\infty} x(k+1)z^{-k} = z \sum_{k=0}^{\infty} x(k+1)z^{-(k+1)} \\ &= z \left[\sum_{j=0}^{\infty} x(j)z^{-j} + x(0) - x(0) \right] = zX(z) - zx(0) \end{aligned}$$



- z-transform of DT-LTI system (w/ $(zI - A)^{-1}z = \mathcal{Z}[A^k]$, $\frac{z}{z-a} = \mathcal{Z}[a^k]$),

$$\begin{aligned} X(z) &= (zI - A)^{-1}zx(0) + (zI - A)^{-1}BU(z) \\ Y(z) &= C(zI - A)^{-1}zx(0) + C(zI - A)^{-1}BU(z) + DU(z) \end{aligned}$$

- Transfer matrix $H(z) := C(zI - A)^{-1}B + D$ (i.e., z-transform of impulse response): stable if all poles are within unit circle, unstable if some are outside unit circle, marginally stable if on the circle.

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Stability and Norms

- **IO-stability (or BIBO-stability):** any bounded input produces bounded output \Rightarrow related to zero-state response and transfer function.
- **Internal stability:** state evolution is bounded for any initial conditions with zero-input \Rightarrow related to zero-input response.
- For LTI systems, internal stability implies IO-stability, but not vice versa (e.g., pz-cancellation: $\dot{x}_1 = x_2, \dot{x}_2 = 4x_1, y = x_2 - 2x_1 \Rightarrow \lambda(A) = \{2, -2\}, H(s) = 1/(s+2)$).
- For LTV systems, internal stability not necessarily implies IO-stability.
- Signal ∞ -norm of $u(t) \in \mathbb{R}^n$: $\|u(\cdot)\|_\infty := \sup_t \|u(t)\|_\infty = \sup_t (\max_i |u_i(t)|)$, where $\|x(t)\|_\infty := \max_i |x_i|$ is vector ∞ -norm.
- Matrix induced ∞ -norm of $A \in \mathbb{R}^{n \times n}$:

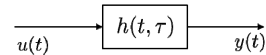
$$\|A\|_{i,\infty} := \sup_x \frac{\|Ax\|_\infty}{\|x\|_\infty} = \max_j \sum_i |A_{ji}|$$

$$\text{i.e., } \|Ax\|_\infty \leq \max_j \sum_i (|A_{ji}| \max_k |x_k|) =: \|A\|_{i,\infty} \|x\|_\infty.$$

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IO-Stability



- **Def:** We say a system $H : u \mapsto y$ is IO-stable (or BIBO-stable) if $\exists M \geq 0$ s.t., for **any bounded** $u(t)$,

$$\|y(\cdot)\|_\infty \leq M \cdot \|u(\cdot)\|_\infty$$

- For LTV system: $\dot{x} = A(t)x + B(t)u, y = C(t)x + D(t)u$, zero-state response is given by:

$$y(t) = C(t) \int_{t_o}^t \Phi(t, \tau) B(\tau) u(\tau) d\tau + D(t) u(t) = \int_{t_o}^t h(t, \tau) u(\tau) d\tau$$

where $h(t, \tau) := C(t)\Phi(t, \tau)B(\tau) + \delta(t-\tau) \cdot D(t)$ is impulse response matrix.

- **Th. 5:** CT-LTV system is IO-stable iff

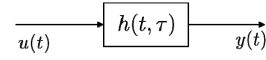
$$\sup_t \int_{t_o}^t \|h(t, \tau)\|_{i,\infty} d\tau \leq M, \quad \forall t \geq t_o$$

- For SISO LTI system: $\|y(\cdot)\|_\infty = \sup_t \left| \int_0^t h(t-\tau) u(\tau) d\tau \right| = \sup_t \left| \int_0^t h(\alpha) u(t-\alpha) d\alpha \right| \leq u_{\max} \int_0^\infty |h(\alpha)| d\alpha \Rightarrow$ i.e., BIBO stable if $\int_0^\infty |h(\alpha)| d\alpha \leq M < \infty$ (i.e., impulse response $h(t)$ is absolutely integrable).

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IO-Stability Proof



Th. 5: CT-LTV system is IO-stable iff

$$\sup_t \int_{t_o}^t \|h(t, \tau)\|_{i, \infty} d\tau \leq M, \quad \forall t \geq t_o$$

- Sufficiency (\Leftarrow): from $y(t) = \int_{t_o}^t h(t, \tau)u(\tau)d\tau$, for all $t \geq t_o$,

$$\begin{aligned} \|y(\cdot)\|_{\infty} &= \sup_t \left\| \int_{t_o}^t h(t - \tau)u(\tau)d\tau \right\|_{\infty} \leq \sup_t \int_{t_o}^t \|h(t - \tau)u(\tau)\|_{\infty} d\tau \\ &\leq \sup_t \int_{t_o}^t \|h(t, \tau)\|_{i, \infty} \|u(\tau)\|_{\infty} d\tau \leq \|u(\cdot)\|_{\infty} \sup_t \int_{t_o}^t \|h(t, \tau)\|_{i, \infty} d\tau \end{aligned}$$

definition of induced norm definition of signal norm

- Necessity (\Rightarrow): Suppose not. Consider SISO LTI for simplicity. Then,

$$|y(t)| = \left| \int_0^t h(\alpha)u(t - \alpha)d\alpha \right| \rightarrow \infty$$

$$\text{if we choose bounded input } u(t) \text{ s.t., } u(t - \alpha) = \begin{cases} 1 & \text{if } h(\alpha) \geq 0 \\ -1 & \text{if } h(\alpha) < 0 \end{cases}.$$

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IO-Stability of CT-LTI System

- **Th. 5:** CT-LTV system is IO-stable iff

$$\sup_t \int_{t_o}^t \|h(t, \tau)\|_{i, \infty} d\tau \leq M, \quad \forall t \geq t_o$$

- **Th. 5-M1:** CT-LTI system is IO-stable (or BIBO-stable) iff every $h_{ij}(t)$ is absolutely integrable.

- The above condition can be rewritten by

$$\int_0^{\infty} \|h(t)\|_{i, \infty} dt = \int_0^{\infty} \max_j \sum_i |h_{ji}(t)| dt \leq M$$

with the remaining arguments going similarly as before.

- **Th. 5-M2:** CT-LTI system with a proper rational transfer matrix $H(s)$ is BIBO-stable iff every pole of every component $H_{ij}(s)$ has a negative real part.

- For BIBO-stability, $H(s)$ should be strictly stable (i.e., poles strictly within LHS).
- Poles of $H(s)$ is a subset of eigenvalues of A .

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IO-Stability and CT-LTI Response

- **Th. 5.2:** Suppose a SISO CT-LTI system with impulse response $h(t)$ is BIBO-stable. Then,

$$\begin{aligned} y(t) &\rightarrow aH(0) \text{ if } u(t) = a; \text{ and} \\ y(t) &\rightarrow |H(j\omega_o)| \sin(\omega_o t + \angle H(j\omega_o)) \text{ if } u(t) = \sin \omega_o t \end{aligned}$$

- If $u(t) = a$, with $H(s) = \int_0^\infty h(\tau)e^{-s\tau}d\tau$,

$$y(t) = \int_0^t h(\tau)u(t-\tau)d\tau = a \int_0^t h(\tau)d\tau \Rightarrow y(t) \rightarrow a \int_0^\infty h(\tau)d\tau = aH(0)$$

- From $H(j\omega_o) = \int_0^\infty h(\tau)e^{-j\omega_o\tau}d\tau = \int_0^\infty h(\tau)[\cos \omega_o\tau - j \sin \omega_o\tau]d\tau$,
 $\text{Re}[H(j\omega_o)] = \int_0^\infty h(\tau) \cos \omega_o\tau d\tau$ and $\text{Im}[H(j\omega_o)] = -\int_0^\infty h(\tau) \sin \omega_o\tau d\tau$.

Also, if $u(t) = \sin \omega_o t$, output is given by $y(t) = \int_0^t h(\tau) \sin \omega_o(t - \tau)d\tau = \sin \omega_o t \int_0^t h(\tau) \cos \omega_o\tau d\tau - \cos \omega_o t \int_0^t h(\tau) \sin \omega_o\tau d\tau$. Thus, from absolute integrability of $h(t)$, the integrands exist and

$$\begin{aligned} y(t) &\rightarrow \sin \omega_o t \int_0^\infty h(\tau) \cos \omega_o\tau d\tau - \cos \omega_o t \int_0^\infty h(\tau) \sin \omega_o\tau d\tau \\ &= \sin \omega_o t \text{Re}[H(j\omega_o)] + \cos \omega_o t \text{Im}[H(j\omega_o)] = |H(j\omega_o)| \sin(\omega_o t + \angle H(j\omega_o)) \end{aligned}$$

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IO-Stability of CT-LTV and DT-LTI Systems

- **Th. 5-M3:** CT-LTV system with $h(t, \tau) = C(t)\Phi(t, \tau)B(\tau) + \delta(t - \tau)D(t)$ is IO-stable (or BIBO-stable) iff $\|D(\cdot)\|_\infty \leq M_1$ and

$$\int_{t_o}^t \|C(t)\Phi(t, \tau)B(\tau)\|_{i,\infty} d\tau \leq M_2, \quad \forall t \geq t_o$$

- Consider DT-LTV system, with its zero-state output response given by

$$y(k) = C(k) \sum_{i=k_o}^{k-1} \Phi(k, i+1)B(i)u(i) + D(k)u(k) = \sum_{i=k_o}^k h(k, i)u(i)$$

with impulse response $h(k, i) = C(k)\Phi(k, i+1)B(i) + \delta(k - i)D(i)$. BIBO-stability would then have something with absolute summability of $h(k, i)$, i.e., $\sum_{i=0}^k \|h(k, i)\|_{i,\infty} \leq M, \forall k \geq i$.

- **Th. 5-MD1:** DT-LTI system with impulse response matrix $h(k) \in \mathbb{R}^{m \times p}$ is BIBO-stable iff $h_{ij}(k)$ is absolutely summable.
- **Th. 5-MD2:** DT-LTI system with proper rational transfer matrix $H(z)$ is BIBO-stable iff every pole of every $H_{ij}(z)$ has a magnitude less than 1.

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IO-Stability and DT-LTI Response

- **Th. 5-D2:** Suppose a SISO DT-LTI system with impulse response sequence $h(k)$ is BIBO-stable. Then,

$$\begin{aligned} y(k) &\rightarrow aH(1) \text{ if } u(k) = a; \text{ and} \\ y(k) &\rightarrow |H(e^{-j\omega_o})| \sin(\omega_o k + \angle H(e^{-j\omega_o})) \text{ if } u(k) = \sin \omega_o k \end{aligned}$$

- If $u(k) = a$, with $H(z) = \sum_{i=1}^{\infty} h(i)z^{-i}$,

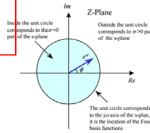
$$y(k) = \sum_{m=0}^k h_{k-m} u_m = \sum_{m=0}^k h_m u_{k-m} = a \sum_{m=0}^k h_m \Rightarrow y(k) \rightarrow a \sum_{m=0}^{\infty} h_m = aH(1)$$

- From $H(e^{-j\omega_o}) = \sum_{m=0}^{\infty} h_m e^{-j\omega_o m} = \sum_{m=0}^{\infty} h_m [\cos \omega_o m - j \sin \omega_o m]$,
 $\text{Re}[H(e^{-j\omega_o})] = \sum_{m=0}^{\infty} h_m \cos \omega_o m$ and $\text{Im}[H(e^{-j\omega_o})] = -\sum_{m=0}^{\infty} h_m \sin \omega_o m$

Also, if $u(k) = \sin \omega_o k$, output is given by $y(k) = \sum_{m=0}^k h_m \sin \omega_o (k-m)$
 $= \sin \omega_o k \sum_{m=0}^k h_m \cos \omega_o m - \cos \omega_o k \sum_{m=0}^k h_m \sin \omega_o m$.

Thus, from absolute summability of $h(k)$, summations exist (w/ $k \rightarrow \infty$) and

$$\begin{aligned} y(k) &\rightarrow \sin \omega_o k \text{Re}[H(e^{-j\omega_o})] + \cos \omega_o k \text{Im}[H(e^{-j\omega_o})] \\ &= |H(e^{-j\omega_o})| \sin(\omega_o k + \angle H(e^{-j\omega_o})) \end{aligned}$$



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Internal Stability

- Recall the zero-input responses of FD linear systems:

$$x(t) = \Phi(t, t_o)x(t_o), \quad x(k) = \Phi(k, k_o)x(k_o)$$

where $\Phi(t, t_o) = e^{A(t-t_o)}$ and $\Phi(k, k_o) = \Pi_{i=k_o}^{k-1} A$ for CT/DT LTI systems.

- **Def. 5.1:** Linear systems is **Lyapunov stable (or marginally stable)** if every bounded IC x_o produces bounded zero-input response x ; **asymptotically stable** if, for all bounded ICs x_o , MS and $x \rightarrow 0$.

– For MS $\Rightarrow \Phi$ should be bounded; For AS $\Rightarrow \Phi$ should converge to zero.

- **Th. 5-4:** CT-LTI system $\dot{x} = Ax$ is MS iff all $\lambda_i(A)$ are in LHP with some $j\omega$ -axis being non-deficient; AS iff all $\lambda_i(A)$ are strictly within LHP.
- **Th. 5-D4:** DT-LTI system $x_{k+1} = Ax_k$ is MS iff all $\lambda_i(A)$ are in UC with some on UC being non-deficient; AS iff all $\lambda_i(A)$ strictly within UC.

$$e^{At} = T \begin{bmatrix} e^{\lambda_1 t} & 0 & 0 & 0 \\ 0 & e^{\lambda_2 t} & t e^{\lambda_2 t} & 0 \\ 0 & 0 & e^{\lambda_2 t} & 0 \\ 0 & 0 & 0 & e^{\lambda_4 t} \end{bmatrix} T^{-1}, \quad A^k = T \begin{bmatrix} \lambda_1^k & 0 & 0 & 0 \\ 0 & \lambda_2^k & k \lambda_2^{k-1} & 0 \\ 0 & 0 & \lambda_2^k & 0 \\ 0 & 0 & 0 & \lambda_4^k \end{bmatrix} T^{-1}$$

- Not applicable to LTV systems...

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Lyapunov Stability - Definition

Def. 2.1: Consider an **autonomous** system

$$\dot{x} = f(x), \quad f(0) = 0$$

$f : \mathcal{D} \rightarrow \mathbb{R}^n$ locally Lipschitz on \mathcal{D} and $0 \in \mathcal{D}$. Then, equilibrium $x = 0$ is

- Lyapunov stable, if, $\forall \epsilon > 0, \exists \delta(\epsilon) > 0$ s.t.,

$$\|x(0)\| < \delta \implies \|x(t)\| < \epsilon, \quad \forall t \geq 0$$

- unstable, if it is not stable

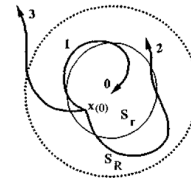
- asymptotically stable, if it is stable and we can find $\delta' > 0$ s.t.

$$\|x(0)\| < \delta' \implies \|x(t)\| \rightarrow 0$$

- exponentially stable if $\exists \alpha, \gamma, \delta' > 0$ s.t.,

$$\|x(0)\| < \delta' \implies \|x(t)\| \leq \alpha \|x(0)\| e^{-\gamma t}$$

- globally asymptotically stable, if asymptotically stable for any $\forall x(0) \in \mathbb{R}^n$.



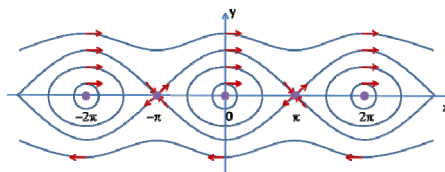
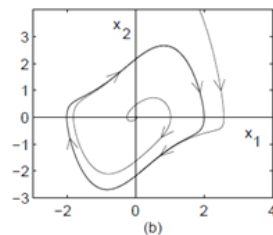
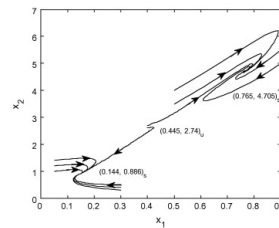
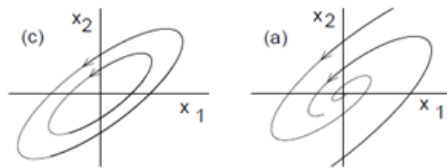
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Lyapunov Stability - Examples

Lyapunov stable, if, for any $\epsilon > 0$, there exists $\delta(\epsilon) > 0$ s.t.,

$$\|x(0)\| < \delta \implies \|x(t)\| < \epsilon, \quad \forall t \geq 0$$



satisfy definition: 1) for some ϵ or 2) $\forall \delta, \exists \epsilon$

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CT-LTI Lyapunov Theorem

Th. 5.5: $A \in \mathbb{R}^{n \times n}$ is Hurwitz (AS) iff, for any $Q \succ 0$, \exists a unique $P \succ 0$ s.t.

$$PA + A^T P = -Q \quad (\text{Lyapunov equation})$$

- Lyapunov analysis: for $\dot{x} = Ax$, define $V = \frac{1}{2}x^T P x \Rightarrow \dot{V} = -x^T Q x < 0$ unless $V = 0 \Rightarrow V(t) \rightarrow 0$ (AS, in fact, ES).
- Can be used to find Lyapunov function: given $A = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix}$, choose any $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \succ 0$. Then, solve for $P = \begin{bmatrix} 1.5 & -0.5 \\ -0.5 & 1.0 \end{bmatrix} \succ 0$.
- (Sufficiency \Rightarrow): Given $Q \succ 0$, define $P := \int_0^\infty e^{A^T t} Q e^{A t} dt \succ 0$, which exists (from AS) and PD. Then,

$$\begin{aligned} PA + A^T P &= \int_0^\infty e^{A^T t} Q e^{A t} A dt + \int_0^\infty A^T e^{A^T t} Q e^{A t} dt \\ &= \int_0^\infty \frac{d}{dt} \left(e^{A^T t} Q e^{A t} \right) dt = e^{A^T t} Q e^{A t} \Big|_0^\infty = -Q \end{aligned}$$

- (Necessity \Leftarrow): Define λ, v s.t., $A v = \lambda v \Rightarrow v^* A^* = v^* \bar{A}^T = v^* A^T = \bar{\lambda} v^*$. Further, $-v^* Q v = v^* (PA + A^T P) v = (\lambda + \bar{\lambda}) v^* P v$ where $v^* Q v > 0$ and $v^* P v > 0 \Rightarrow \lambda(A)$ in LHP (i.e., A is Hurwitz).

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CT-LTV Lyapunov Theorem

Consider linear time-varying system

$$\dot{x} = A(t)x, \quad x(t_o)$$

with $x = 0$ equilibrium. Then, $x(t) = \Phi(t, t_o)x(t_o)$, where $\Phi(t, t_o)$ is the state transition matrix (e.g., $\Phi(t, t_o) = e^{A(t-t_o)}$ for LTI system).

Ex) Consider $\dot{x} = A(t)x$ with $A(t) = \begin{bmatrix} -1 + 1.5 \cos^2 t & 1 - 1.5 \sin t \cos t \\ -1 - 1.5 \sin t \cos t & -1 + 1.5 \sin^2 t \end{bmatrix}$.

Then, $\text{eig}(A(t)) = -0.25 \pm 0.25\sqrt{7}j$. Yet, $\Phi(t, 0) = \begin{bmatrix} e^{0.5t} \cos t & e^{-t} \sin t \\ -e^{0.5t} \sin t & e^{-t} \cos t \end{bmatrix}$

Th. 5.5V: Suppose \exists smooth bounded $P(t) \succ 0$ s.t.

$$\dot{P}(t) + P(t)A(t) + A^T(t)P(t) = -Q(t)$$

with $Q(t) \succ 0$. Then, $x = 0$ is AS. Also, if $A(t)$ is continuous and bounded and $x \rightarrow 0$ (i.e., AS), for any $Q(t) \succ 0$, $\exists P(t) \succ 0$ satisfying Lyapunov equation.

- (\Rightarrow) Lyapunov analysis with $V = \frac{1}{2}x^T P(t)x \rightarrow \dot{V} = -x^T Q(t)x < 0$ unless $V(t) = 0 \rightarrow x \rightarrow 0$ (i.e., AS, in fact, ES).
- (\Leftarrow) Choose $V(x, t) = x^T P(t)x$. Also, given $Q(t)$, $P(t) = \int_t^\infty \Phi^T(\tau, t) Q(\tau) \Phi(\tau, t) d\tau$.

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CT-LTV Lyapunov Analysis

Th. 8.2 (R): Consider CT-LTV system $\dot{x} = A(t)x$. Denote, at each t , the largest and smallest eigenvalues of $A(t) + A^T(t)$ by $\lambda_{\max}(t)$ and $\lambda_{\min}(t)$. Then,

$$\|x_o\| e^{\frac{1}{2} \int_{t_o}^t \lambda_{\min}(\sigma) d\sigma} \leq \|x(t)\| \leq \|x_o\| e^{\frac{1}{2} \int_{t_o}^t \lambda_{\max}(\sigma) d\sigma}, \quad \forall t \geq t_o$$

where $\|\cdot\|$ is the vector (i.e., Euclidean) 2-norm.

- Define $V(t) = \|x(t)\|^2$. Then,

$$\lambda_{\min}(t) \|x(t)\|^2 \leq \frac{dV}{dt} = x^T(t)(A^T(t) + A(t))x(t) \leq \lambda_{\max}(t) \|x(t)\|^2$$

i.e., $V(t_o) e^{\int_{t_o}^t \lambda_{\min}(\sigma) d\sigma} \leq V(t) \leq V(t_o) e^{\int_{t_o}^t \lambda_{\max}(\sigma) d\sigma}$, for all $t \geq t_o$.

- **Cor. 8.2-1:** CT-LTV system is stable if

$$\int_{\tau}^t \lambda_{\max}(\sigma) d\sigma \leq \gamma, \quad \forall t, \tau, \text{ s.t., } t \geq \tau$$

- **Cor. 8.2-2:** CT-LTV system is ES if

$$\int_{\tau}^t \lambda_{\max}(\sigma) d\sigma \leq -\lambda(t - \tau) + \gamma, \quad \forall t, \tau, \text{ s.t., } t \geq \tau$$

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CT-LTV Stability w/ Perturbation -I

Th. 8.6 (R): Suppose CT-LTV system $\dot{x} = A(t)x$ is ES with continuous/bounded $A(t)$. Then, \exists a small enough $\beta > 0$ s.t.,

$$\dot{z} = [A(t) + F(t)]z$$

is also ES if $\|F(t)\| \leq \beta$, where $\|\cdot\|$ is matrix 2-norm (other norms also work).

- From Lyapunov theorem, $\exists P(t), Q(t) \succ 0$ s.t.,

$$\dot{P}(t) + A(t)^T P(t) + A(t) P(t) = -Q(t)$$

where $P(t) := \int_t^\infty \Phi^T(\sigma, t) Q(t) \Phi(t, \sigma) d\sigma$. Then,

$$[A(t) + F(t)]^T P(t) + P(t)[A(t) + F(t)] + \dot{P}(t) = F^T(t) P(t) + P(t) F(t) - Q(t)$$

where LHS is PD if $\|F(t)\|$ small enough, since $P(t)$ bounded (from ES).

- ES is robust against bounded perturbation.
- This robustness also true for CT-LTI systems.
- If CT-LTV system is AS, stability is in general fragile.

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CT-LTV Stability w/ Perturbation - II

Th. 8.5 (R): Suppose CT-LTV system $\dot{x} = A(t)x$ is stable. Then, \exists a small enough $\beta > 0$ s.t.,

$$\dot{z} = [A(t) + F(t)]z$$

is also stable if $\int_{\tau}^{\infty} \|F(\sigma)\| d\sigma \leq \beta, \forall \tau \geq 0$.

- Using convolution: $z(t) = \Phi(t, t_o)z_o + \int_{t_o}^t \Phi(t, \sigma)F(\sigma)z(\sigma)d\sigma$, where, from stability, $\|\Phi(t, \sigma)\| \leq \gamma \forall t, \sigma, t \geq \sigma$.
- Taking the 2-norm: $\|z(t)\| \leq \gamma\|z_o\| + \int_{t_o}^t \gamma\|F(\sigma)\|\|z(\sigma)\|d\sigma$, i.e., an implicit inequality w.r.t., $\|z(t)\|$.
- From Gronwall-Bellman inequality: $\|z(t)\| \leq \gamma\|z_o\|e^{\int_{t_o}^t \gamma\|F(\sigma)\|d\sigma} \leq \gamma\|z_o\|e^{\gamma\beta}, \forall t \geq t_o$, i.e., stable.
- **Lem. 3.2 (Gronwall-Bellman):** For continuous $\phi(t), v(t)$ with $v(t) \geq 0, \forall t \geq t_o$,

$$\phi(t) \leq \psi + \int_{t_o}^t v(\sigma)\phi(\sigma) d\sigma \Rightarrow \phi(t) \leq \psi e^{\int_{t_o}^t v(\sigma)d\sigma}$$
 - $r(t) = \psi + \int_{t_o}^t v(\sigma)d\sigma \Rightarrow \dot{r}(t) = v(t)\phi(t) \leq v(t)r(t) \Rightarrow$ multiplying $e^{-\int_{t_o}^t v(\sigma)d\sigma} \Rightarrow \frac{d}{dt} \left[r(t)e^{-\int_{t_o}^t v(\sigma)d\sigma} \right] \leq 0 \Rightarrow \phi(t) \leq r(t) \leq \psi e^{\int_{t_o}^t v(\sigma)d\sigma}$.
- However small persistent perturbation can destabilize CT-LTV systems.

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DT-LTI Lyapunov Theorem

Th. 5-D5: $A \in \mathbb{R}^{n \times n}$ has $\rho(A) < 1$ (i.e., DT-LTI AS) iff, for any $Q \succ 0$, \exists a unique $P \succ 0$ s.t.

$$A^T P A - P = -Q \quad (\text{Lyapunov equation})$$

- Lyapunov analysis: for $x_{k+1} = Ax_k$, define $V_k = \frac{1}{2}x_k^T P x_k \Rightarrow V_{k+1} - V_k = \frac{1}{2}x_k^T (A^T P A - P)x_k = -\frac{1}{2}x_k^T Q x_k < 0$ unless $V_k = 0 \Rightarrow V_k \rightarrow 0$ (AS).
- (Sufficiency \Rightarrow): Given $Q \succ 0$, define $P := \sum_{m=0}^{\infty} (A^T)^m Q A^m \succ 0$, which exists (from AS) and PD. Then,

$$\begin{aligned} A^T P A - P &= \sum_{m=0}^{\infty} A^T (A^T)^m Q A^m A - \sum_{m=0}^{\infty} (A^T)^m Q A^m \\ &= \sum_{m=1}^{\infty} (A^T)^m Q A^m - \sum_{m=0}^{\infty} (A^T)^m Q A^m = -Q \end{aligned}$$

- (Necessity \Leftarrow): Define λ, v s.t., $Av = \lambda v \Rightarrow v^* A^* = v^* \bar{A}^T = v^* A^T = \bar{\lambda} v^*$. Further,

$$-v^* Q v = v^* (A^T P A - P) v = (|\lambda|^2 - 1) v^* P v$$

where $v^* Q v > 0, v^* P v > 0 \Rightarrow |\lambda(A)|^2 < 1$ (i.e., $\rho(A) < 1 \Rightarrow$ DT-LTI AS).

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IO-Stability and Internal Stability

- LTI system stability conditions:

$$\int_0^{\infty} \|C e^{At} B + \delta(t) D\|_{i,\infty} dt \leq M \quad (\text{BIBO}) \quad \text{and} \quad \lambda_i(A) \in \text{LHP} \quad (\text{AS})$$

- AS implies BIBO, but not vice versa (e.g., pz-cancellation).
- BIBO excludes marginal stability.

- LTV system stability conditions:

$$\int_0^{\infty} \|C(t)\Phi(t,\tau)B(\tau) + \delta(t-\tau)D(t)\|_{i,\infty} dt \leq M \quad (\text{BIBO})$$

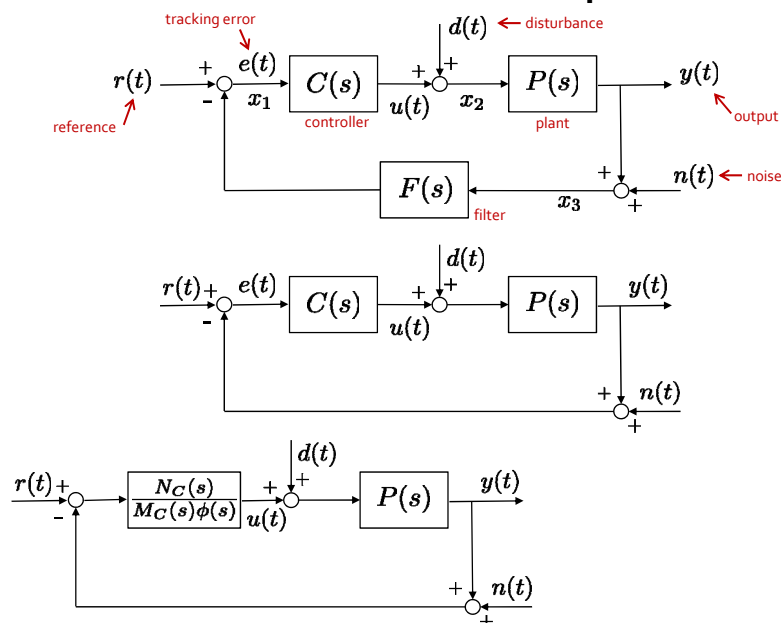
$$\|\Phi(t,\tau)\|_{i,\infty} \rightarrow 0 \quad \text{as } t \rightarrow \infty \quad (\text{AS})$$

- AS may not even imply BIBO (since only AS, not ES as for LTI).
- AS implies BIBO, if ES and $C(t), B(t)$ bounded.

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Basic Feedback Loop



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Basic Feedback Loop

