



ELEC 441: Control Systems

Lecture 8: Internal Stability

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Topics	CT	DT
Modeling	✓	✓
Stability	→ ● ←	→ ● ←
Controllability/Observability		
Realization		
State Feedback/Observers		
LQR/Kalman Filter		

- Last lecture we discussed BIBO stability
- BIBO stability cannot determine stability of state-space models with non-zero initial conditions
- Today, we study **internal stability** for:
 - ▶ LTI CT Systems: $\dot{x}(t) = Ax(t)$, $x(0) = x_0$
 - ▶ LTI DT Systems: $x[k+1] = Ax[k]$, $x[0] = x_0$
 - ▶ Notice that there is no input in the above systems
- Internal stability is assessed in two ways:
 - ▶ **Eigenvalue criteria** (today's lecture)
 - ▶ **Lyapunov theorem** (next lecture)

BIBO Stability Example (Review)

- Consider the following CT LTI system

$$\begin{cases} \dot{x}(t) = \begin{bmatrix} -1 & 10 \\ 0 & 1 \end{bmatrix} x(t) + \begin{bmatrix} -2 \\ 0 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} -2 & 3 \end{bmatrix} x(t) \end{cases}$$

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- The transfer function for the above SS model is given by

$$\begin{aligned} G(s) &:= C(sI - A)^{-1}B + D \\ &= \begin{bmatrix} -2 & 3 \end{bmatrix} \left(sI - \begin{bmatrix} -1 & 10 \\ 0 & 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} -2 \\ 0 \end{bmatrix} \\ &= \frac{4(s-1)}{(s+1)(s-1)} = \frac{4}{(s+1)} \rightarrow \text{Pole is } s = -1 \end{aligned}$$

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- System is BIBO stable after pole/zero cancellation ← **PROBLEM!**
This is problematic for non-zero initial conditions!

BIBO Stability Limitations (Review)



- Pole/zero cancellation in **unstable** region is **DANGEROUS!**
 - ▶ Plant model is **NEVER** exact! (cancellation does not occur in reality)
- Neither nonzero initial conditions nor input disturbances are considered (**very fragile!**)
- Although output signals are bounded, **does not guarantee** that internal signals are bounded

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- BIBO stability is not good enough for state-space models, we need a stability concept that accounts for:
 - ▶ Pole/zero cancellations
 - ▶ Nonzero initial conditions
 - ▶ Internal signal behaviour

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← **Internal Stability!**

Internal Stability for CT LTI Systems

- Consider the CT LTI system with no input

$$\begin{cases} \dot{x}(t) = Ax(t), & x(0) = x_0 \\ y(t) = Cx(t) \end{cases}$$

Internal Stability for CT LTI Systems

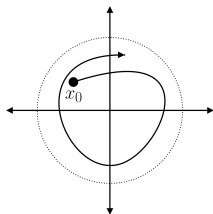
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- The system is **marginally stable (stable in the sense of Lyapunov)**, if for any x_0 the following holds for some $M > 0$:

$$\|x(t)\| \leq M < \infty, \quad \forall t > 0$$

where $\|x(t)\|$ is the **norm** of the state vector



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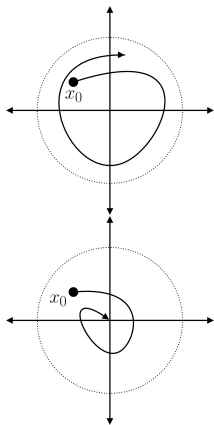
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- The system is **asymptotically stable** if the following holds for any x_0 :

$$x(t) \rightarrow 0 \quad \text{as} \quad t \rightarrow \infty$$



Eigenvalue Criteria for CT LTI Systems

Denote the i -th Eigenvalues of A matrix with λ_i

- **Asymptotically Stable:** $\text{Re}\{\lambda_i\} < 0, \forall i$
- **Marginally Stable:** $\text{Re}\{\lambda_i\} \leq 0, \forall i$
 - ▶ For $\text{Re}\{\lambda_i\} = 0 \rightarrow \text{rank}(\lambda_i I - A) = n - m_i$,
where n is the number of states and m_i is the **multiplicity** of λ_i
 - ▶ **Note:** if $m_i = 1$, the rank condition always holds!
- **Unstable:** two possibilities:
 - ▶ $\text{Re}\{\lambda_i\} > 0$ for some i
 - ▶ For $\text{Re}\{\lambda_i\} = 0 \rightarrow \text{rank}(\lambda_i I - A) \neq n - m_i$

Internal Stability for DT LTI Systems

- Consider the DT LTI system with no input

$$\begin{cases} x[k+1] = Ax[k], & x[0] = x_0 \\ y[k] = Cx[k] \end{cases}$$

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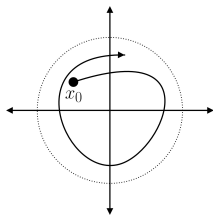
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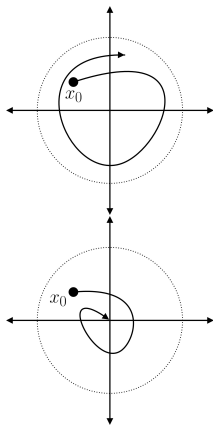
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Eigenvalue Criteria for DT LTI Systems

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Computing Matrix Eigenvalues

How do we compute the eigenvalues of a matrix?

- We can compute the eigenvalues of matrix A by solving

$$\det(\lambda I - A) = 0$$

- The above equation is referred to as the **characteristic equation**

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Eigenvalues for **special** matrices:

- Eigenvalues of **diagonal** matrices are the **diagonal elements**
- Eigenvalues of **triangular** matrices are the **diagonal elements**
- Eigenvalues of **block diagonal** matrices are the eigenvalues of the diagonal matrices

Idea Behind Internal Stability

Continuous-time (CT)

$$\begin{cases} \dot{x}(t) = Ax(t) \\ x(0) = x_0 \end{cases} \rightarrow x(t) = e^{At}x_0$$

- Example $A = \begin{bmatrix} -2 & 0 \\ 0 & -3 \end{bmatrix}$

$$e^{At} = \begin{bmatrix} e^{-2t} & 0 \\ 0 & e^{-3t} \end{bmatrix} \rightarrow \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

- Example $A = \begin{bmatrix} 2 & 0 \\ 0 & -3 \end{bmatrix}$

$$e^{At} = \begin{bmatrix} e^{2t} & 0 \\ 0 & e^{-3t} \end{bmatrix} \rightarrow \begin{bmatrix} \infty & 0 \\ 0 & 0 \end{bmatrix}$$

Discrete-time (DT)

$$\begin{cases} x[k+1] = Ax[k] \\ x[0] = x_0 \end{cases} \rightarrow x[k] = A^k x_0$$

- Example $A = \begin{bmatrix} 0.1 & 0 \\ 0 & -0.3 \end{bmatrix}$

$$A^k = \begin{bmatrix} 0.1^k & 0 \\ 0 & -0.3^k \end{bmatrix} \rightarrow \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

- Example $A = \begin{bmatrix} 1.1 & 0 \\ 0 & -0.3 \end{bmatrix}$

$$A^k = \begin{bmatrix} 1.1^k & 0 \\ 0 & -0.3^k \end{bmatrix} \rightarrow \begin{bmatrix} \infty & 0 \\ 0 & 0 \end{bmatrix}$$

Important Examples

- Consider the following system

$$\dot{x}(t) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} x(t)$$

Eigenvalues: $\lambda_1 = \lambda_2 = 0$, $m = 2$

$\text{rank}(\lambda I - A) = n - m = 0 \rightarrow$ Marginally Stable!

Important Examples

- Consider the following system

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$\text{rank}(\lambda I - A) = n - m = 0 \rightarrow$ Marginally Stable!

- Consider the following system

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x(t)$$

Eigenvalues: $\lambda_1 = \lambda_2 = 0$, $m = 2$

$\underbrace{\text{rank}(\lambda I - A)}_{=1} \neq \underbrace{n - m}_{=0} \rightarrow$ Unstable!

Remark on Stability of LTV Systems

- Consider the following system

$$\dot{x}(t) = \begin{bmatrix} -1 & e^{2t} \\ 0 & -1 \end{bmatrix} x(t), \quad x(0) = x_0$$

- Eigenvalues of $A(t)$ matrix are $\lambda_1 = \lambda_2 = -1, \forall t$.
However, this system is **unstable!**

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However, this system is **unstable!**
- Why? Solve the differential equations:

$$x(t) = \begin{bmatrix} e^{-t} & 0.5(e^t - e^{-t}) \\ 0 & e^{-t} \end{bmatrix} x_0$$

- For an initial state

$$x(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \implies x(t) = \begin{bmatrix} 0.5(e^t - e^{-t}) \\ e^{-t} \end{bmatrix} \rightarrow \begin{bmatrix} \infty \\ 0 \end{bmatrix}$$

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- Eigenvalue criteria does not work for LTV systems!**

- Internal stability and definitions of:
 - ▶ Asymptotic stability
 - ▶ Marginal stability
 - ▶ Instability
- Internal stability for CT and DT systems:
 - ▶ Eigenvalue criteria (today's lecture)
 - ▶ Lyapunov theorem (next lecture)
- Next, Lyapunov theorem for internal stability

Appendix A: Linear Algebra Review

Linear Independence:

- Vectors $\{v_1, \dots, v_n\}$ are **linearly independent** if:

$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = 0 \implies \alpha_1 = \alpha_2 = \dots = \alpha_n = 0$$

Matrix Rank:

- Definition: maximum number of linearly independent row (and column) vectors of a matrix
- Example:

$$\text{rank}(A) = \text{rank} \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix} = 2$$

- Example:

$$\text{rank}(A) = \text{rank} \begin{bmatrix} 1 & -1 \\ -5 & 5 \end{bmatrix} = 1$$