



## 5. Stability

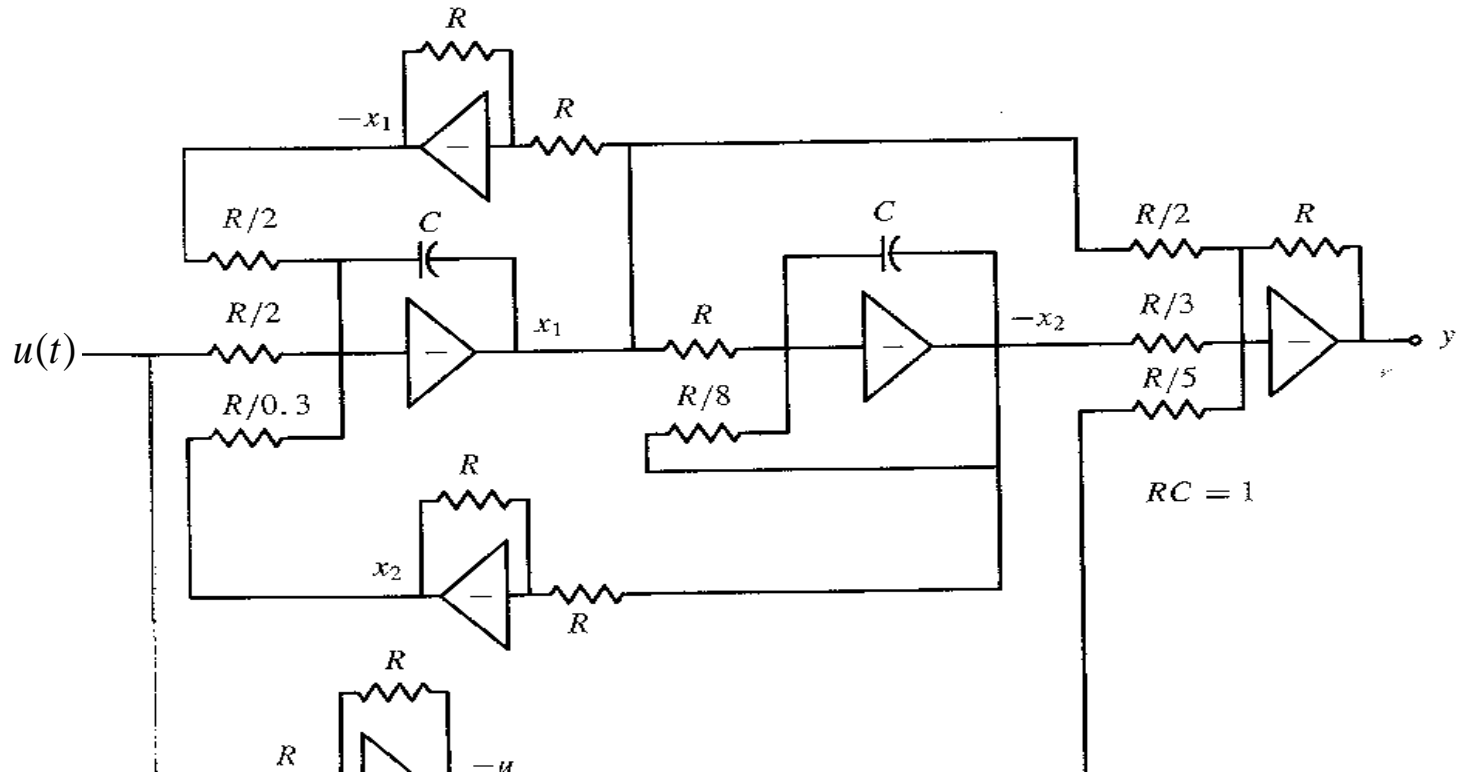
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- ✓ Motivations
- ✓ Input Output Stability
- ✓ Internal Stability for LTI System
- ✓ General Definition of Internal Stability
- ✓ Lyapunov Stability Theory
- ✓ Stability of Linear Time Varying System

# Motivations

Is  $y(t)$  bounded for bounded  $u(t)$ ?  $\rightarrow$  IO stability

Is  $x_1(t)(x_2(t), \dots)$  bounded?  $\rightarrow$  Internal stability





# Input Output Stability

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Input Output Stability of LTI System

$$y(t) = \int_0^t g(t-\tau)u(\tau)d\tau = \int_0^t g(\tau)u(t-\tau)d\tau \dots\dots (*)$$

$u(t)$ : *bounded*

$$|u(t)| \leq u_m < \infty \quad \forall t > 0$$

## Theorem 5.1

(\*) is BIBO stable iff

$$\int_0^{\infty} |g(t)| dt \leq M < \infty.$$



# Input Output Stability

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Pf)

$$\begin{aligned}(\Leftrightarrow) \quad |y(t)| &= \left| \int_0^\infty g(\tau)u(t-\tau)d\tau \right| \\ &\leq \int_0^\infty |g(\tau)||u(t-\tau)|d\tau \\ &\leq u_m \int_0^\infty |g(\tau)|d\tau \leq u_m M < \infty.\end{aligned}$$

( $\Rightarrow$ ) Assume  $\exists t_1$  such that

$$\int_0^{t_1} |g(\tau)|d\tau = \infty.$$

Let us choose

$$u(t_1 - \tau) = \begin{cases} 1 & \text{if } g(\tau) > 0 \\ -1 & \text{if } g(\tau) \leq 0 \end{cases}$$

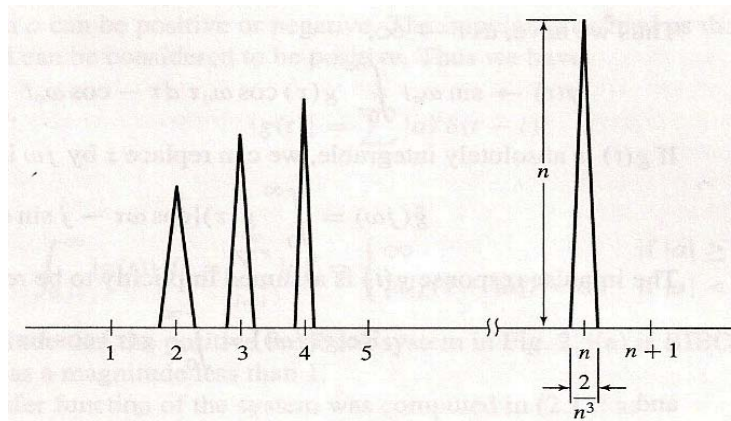
$$y(t_1) = \int_0^{t_1} g(\tau)u(t_1 - \tau)d\tau = \int_0^{t_1} |g(\tau)|d\tau = \infty$$

# Input Output Stability

Note) Even if  $\int |g(\tau)| d\tau < \infty$ ,

$g(\tau)$  may not be bounded or may not converge to zero.

Ex)



$$\sum_{n=2}^{\infty} \frac{1}{n^2} < \infty, \text{ but } \lim_{t \rightarrow \infty} g(t) = \infty$$

**Lemma:** The uniformly continuous function

satisfying  $\int |g(\tau)| d\tau < \infty$ , converge to zero.

*i.e.*  $\lim_{t \rightarrow \infty} g(t) = 0$



# Input Output Stability

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## Theorem 5.2

If  $g(t)$  is BIBO stable,

for  $u(t) = a$ ,  $\lim_{t \rightarrow \infty} y(t) = \mathbf{g}(0)a$

for  $u(t) = \sin w_0 t$ ,  $\lim_{t \rightarrow \infty} y(t) = |\mathbf{g}(jw_0)| \sin(w_0 t + \angle \mathbf{g}(jw_0))$ ,

where  $\mathbf{g}(s)$  is Laplace transform of  $g(t)$ .

Pf)

$$\mathbf{g}(s) = \int_0^{\infty} g(\tau) e^{-s\tau} d\tau$$

$$u(t) = a, y(t) = \int_0^t g(\tau) u(t-\tau) d\tau = a \int_0^t g(\tau) d\tau$$

$$\lim_{t \rightarrow \infty} y(t) = a\mathbf{g}(0)$$



# Input Output Stability

Pf\_cont)

$$u(t) = \sin w_0 t$$

$$y(t) = \int_0^t g(\tau) \sin w_0 (t - \tau) d\tau$$

$$= \int_0^t g(\tau) [\sin w_0 t \cos w_0 \tau - \cos w_0 t \sin w_0 \tau] d\tau$$

$$= \sin w_0 t \int_0^t g(\tau) \cos w_0 \tau d\tau - \cos w_0 t \int_0^t g(\tau) \sin w_0 \tau d\tau$$

$$\mathbf{g}(jw) = \int_0^\infty g(\tau) [\cos w\tau - j \sin w\tau] d\tau$$

$$\text{Re}[\mathbf{g}(jw)] = \int_0^\infty g(\tau) \cos w\tau d\tau$$

$$\text{Im}[\mathbf{g}(jw)] = -\int_0^\infty g(\tau) \sin w\tau d\tau$$

$$\lim_{t \rightarrow \infty} y(t) = \sin w_0 t \text{Re}[\mathbf{g}(jw_0)] + \cos w_0 t \text{Im}[\mathbf{g}(jw_0)]$$

$$= |\mathbf{g}(jw)| \sin(w_0 t + \angle \mathbf{g}(jw))$$



# Input Output Stability

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## Theorem 5.3

A SISO system with proper rational transfer function  $\mathbf{g}(s)$  is BIBO stable iff every pole of  $\mathbf{g}(s)$  has a negative real part.

Note) MIMO is BIBO.

iff  $g_{ij}(t)$  is BIBO stable (Theorem 5.M1, Theorem 5.M3).

Note)

Time varying system is BIBO stable iff

$$\int_{t_0}^t |g(t, \tau)| d\tau \leq M < \infty \text{ for all } t, t_0 > 0$$



# Input Output Stability

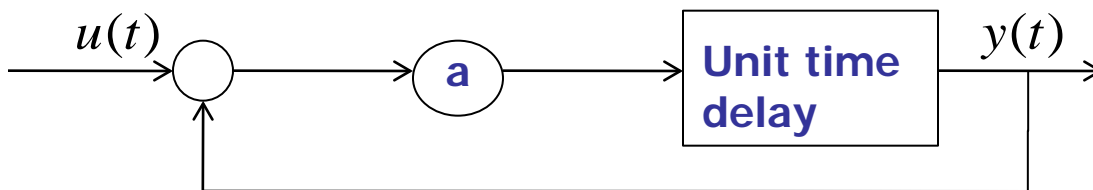
Example) unity-feedback system

$$g(t) = \sum_{i=1}^{\infty} a^i \delta(t-i) \rightarrow |g(t)| = \sum_{i=1}^{\infty} |a|^i \delta(t-i)$$

$$\int_0^{\infty} |g(t)| dt = \sum_{i=1}^{\infty} |a|^i = \begin{cases} \infty & \text{if } |a| \geq 1 \\ |a|/(1-|a|) < \infty & \text{if } |a| < 1 \end{cases}$$

This system is BIBO stable iff the gain  $a$  has a magnitude less than 1.

$\mathbf{g}(s) = \frac{se^{-s}}{1-ae^{-s}}$  is not rational function and Theorem 5.3 is not applicable.



# Input Output Stability

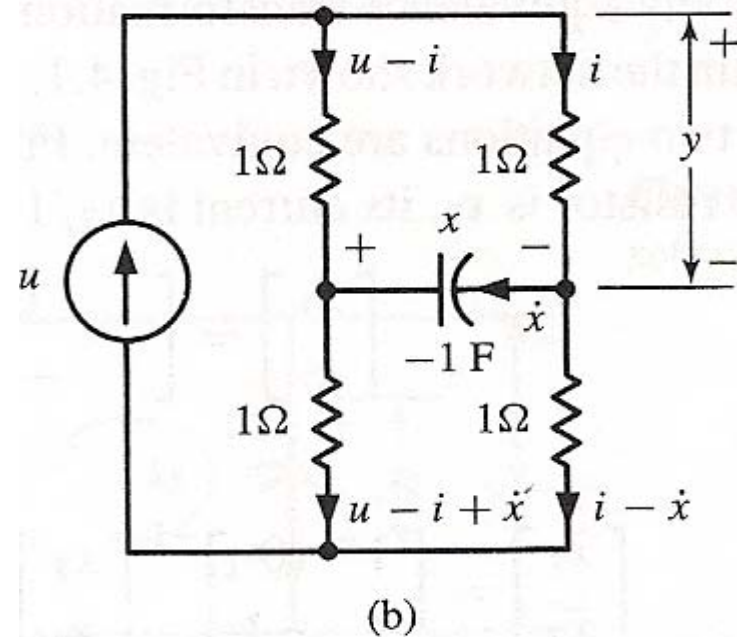
## Example

$$\dot{x} = x(t), \quad y = 0.5x(t) + 0.5u(t),$$

$$\bar{A} = 1, \quad \bar{B} = 0, \quad \bar{C} = 0.5, \quad \bar{D} = 0.5$$

$$\mathbf{g}(s) = \mathbf{C}^T (s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D}$$

$$= 0.5(s-1)^{-1} \cdot 0 + 0.5 = 0.5$$



→ BIBO stable even if it has positive real part eigenvalue.

→ Internal stability (state stability is needed).



# Input Output Stability

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Discrete-Time Case

$$y[k] = \sum_{m=0}^k g[k-m]u[m]$$

**Theorem 5.D.1**

Discrete-time SISO system is BIBO  
iff  $g[k]$  is absolutely summable, i.e.,

$$\sum_{k=0}^{\infty} |g[k]| \leq M < \infty$$



# Input Output Stability

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## Theorem 5.D.2

If a discrete-time system is BIBO stable, then

1.  $u[k] = a$

$$\lim_{k \rightarrow \infty} y[k] = \mathbf{g}(1)a$$

2.  $u[k] = \sin w_0 k$

$$\lim_{k \rightarrow \infty} y[k] \rightarrow \left| \mathbf{g}(e^{jw_0}) \right| \sin(w_0 k + \angle \mathbf{g}(e^{jw_0}))$$

where  $\mathbf{g}(z) = \sum_{m=0}^{\infty} g[m]z^{-m}$

## Theorem 5.D.3

Discrete-time SISO system with proper rational transfer function  $\mathbf{g}(z)$  is BIBO stable iff every pole has a magnitude less than 1.



# Internal Stability for LTI System

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## Internal Stability (State Stability)

$$\dot{x}(t) = Ax(t)$$

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

**Definition** : Zero-input Response of  $\dot{x} = Ax$  is marginally stable or stable in the sense of Lyapunov if  $|x(t)| < \infty$  for all  $t > 0$  & all  $x_0$ , and is asymptotically stable if  $x(t) \rightarrow 0$  as  $t \rightarrow \infty$

# Linear System for LTI System

## Theorem 5.4

1. The state equation is marginally stable iff all eigenvalues of  $A$  have zero or negative real part and those with zero real part are simple roots of minimal polynomial of  $A$ .
2. The state equation is asymptotically stable iff all eigenvalues of  $A$  have negative real part.

- ①  $\begin{cases} \text{Re}(\lambda_i) \leq 0 \text{ and} \\ \text{Re}(\lambda_i) = 0 \ \& \ \bar{n}_i = 1 \end{cases}$   
 $\Rightarrow$  marginally stable
- ② All  $\text{Re } \lambda_i < 0 \rightarrow \text{A.S}$

$$e^{At} = Qe^{\hat{A}t}Q^{-1}$$

$$= Q \begin{bmatrix} e^{\lambda_1 t} & te^{\lambda_1 t} & \frac{1}{2}t^2 e^{\lambda_1 t} & & \\ & e^{\lambda_1 t} & te^{\lambda_1 t} & & \\ & & e^{\lambda_1 t} & & \\ & & & e^{\lambda_2 t} & \\ & & & & \ddots \end{bmatrix} Q^{-1}$$



# Internal Stability for LTI System

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## Example 5.4

$$\dot{x} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix} x \rightarrow x(t) = c_1 + c_2 e^{-t} \leq c_1$$

Characteristic polynomial is  $\lambda^2(\lambda + 1)$

Minimal polynomial is  $\lambda(\lambda + 1)$

$\lambda = 0$  is simple root  $\rightarrow$  marginally stable

$$\dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix} x \rightarrow x(t) = c_1 + c_2 t + c_3 e^{-t} \rightarrow \infty$$

Minimal polynomial is  $\lambda^2(\lambda + 1)$

$\rightarrow$  not marginally stable



# Internal Stability for LTI System

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Discrete-Time Case

$$x[k + 1] = Ax[k]$$

$$x[k] = A^k x_0$$

## Theorem 5.D.4

1.  $|\lambda_i| \leq 1$  and  $|\lambda_i| = 1$  with  $\bar{n}_i = 1 \rightarrow$  marginally stable
2.  $|\lambda_i| < 1 \rightarrow$  asymptotically stable





# General Definition of Internal Stability

## Definition (Equilibrium Point)

$x_e$  is said to be equilibrium point at  $t_0$  iff

$$x(t) = \Phi(t, t_0)x_e = x_e \quad \forall t \geq t_0.$$

Note)

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

$$x(t) = \Phi(t, t_0)x(t_0)$$

$$\Phi(t, t_0) = e^{\int_{t_0}^t A(\tau) d\tau} \quad \text{or} \quad X(t)X^{-1}(t_0)$$

$$x_e = \Phi(t, t_0)x_e$$

$$[\mathbf{I} - \Phi(t, t_0)]x_e = 0 \quad \forall t \geq t_0$$

If all columns of  $[\mathbf{I} - \Phi(t, t_0)]$  are linearly independent

$$x_e = 0,$$

otherwise it may be

$$x_e \neq 0.$$



# General Definition of Internal Stability

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## Definition

$x_e$  is stable i.s.L at  $t_0$  iff

for every  $\varepsilon > 0$ ,  $\exists \delta(\varepsilon, t_0) > 0$  such that

$$\|x_0 - x_e\| \leq \delta(\varepsilon, t_0) \rightarrow \|x(t) - x_e\| \leq \varepsilon \quad \forall t \geq t_0.$$

$x_e$  is uniformly stable i.s.L  $[t_0, \infty)$  iff

for every  $\varepsilon > 0$ ,  $\exists \delta(\varepsilon) > 0$  such that

$$\|x_0 - x_e\| \leq \delta(\varepsilon) \rightarrow \|x(t) - x_e\| \leq \varepsilon \quad \forall t \geq t_0.$$



# General Definition of Internal Stability

## Example

$$\dot{x} = (6t \sin t - 2t)x(t)$$

$$x(t) = x(t_0) \exp \int_{t_0}^t (6\tau \sin \tau - 2\tau) d\tau$$

$$= x(t_0) \exp(6 \sin t - 6t \cos t - t^2 - 6 \sin t_0 + 6t_0 \cos t_0 + t_0^2)$$

Define  $c(t_0) = \sup_{t \geq t_0} \exp(6 \sin t - 6t \cos t - t^2 - 6 \sin t_0 + 6t_0 \cos t_0 + t_0^2)$

$$< \sup_{t \geq t_0} \exp(12 + 6(t + t_0) - (t^2 - t_0^2)),$$

$$< \sup_{t \geq t_0} \exp(12 + 6T + 12t_0 - T^2), \quad T := t - t_0$$

$$< \infty$$

$$|x(t)| < |x(t_0)| c(t_0)$$



# General Definition of Internal Stability

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## Example (cont)

For any given  $\varepsilon > 0$ , if we choose

$$\delta(\varepsilon, t_0) = \varepsilon / c(t_0),$$

then

$$\begin{aligned} |x(t_0) - x_e| &\leq \delta(\varepsilon, t_0), \quad x_e = 0 \\ \rightarrow |x(t)| &< |x(t_0)| c(t_0) \leq \varepsilon \quad \forall t \geq t_0. \end{aligned}$$

This implies the system is stable i.s.L.



# General Definition of Internal Stability

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## Example (cont)

On the other hand, if we choose

$$t_0 = 2n\pi, \quad t = (2n+1)\pi$$

$$x[(2n+1)\pi] = x(2n\pi) \exp^{(4n+1)(6-\pi)\pi}$$

for

$$x[(2n+1)\pi] < \varepsilon$$

$$x(t_0) = x(2n\pi) < \varepsilon \cdot \exp^{-((4n+1)(6-\pi)\pi)} = \delta(\varepsilon, t_0 = 2n\pi) .$$

It is not possible to choose a single  $\delta(\varepsilon)$   
independent of  $t_0 = 2n\pi$ .

That is  $|x(t_0)| \leq \delta(\varepsilon, t_0) \rightarrow 0$  as  $t_0 \rightarrow \infty$ .

This implies the system is not uniformly stable.

# General Definition of Internal Stability

## Example: Pendulum

Let  $x_1 = \theta$

$$x_2 = \dot{\theta}$$

$$\dot{x}_1 = x_2$$

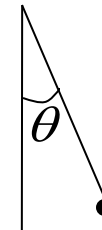
$$\dot{x}_2 = \left( -\frac{g}{l} \right) \sin x_1 + \frac{\cos x_1}{ml} u(t)$$

$$\dot{x}_1 = x_2 = 0 \text{ with } u(t) = 0$$

$$x_e = \begin{bmatrix} k\pi \\ 0 \end{bmatrix}, k = 0, \pm 1, \pm 2 \dots$$

$$\begin{cases} x_e = [k\pi \ 0]^T, & k = 0, \pm 2, \pm 4 \text{ uniformly stable eq. pt.} \end{cases}$$

$$\begin{cases} x_e = [k\pi \ 0]^T, & k = \pm 1, \pm 3, \quad \text{unstable eq. pt.} \end{cases}$$





# General Definition of Internal Stability

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## Definition

(1)  $x_e$  is asymptotically stable at  $t_0$  if

- $x_e$  is stable i.s.L at  $t_0$ , and
- $\|x(t) - x_e\| \rightarrow 0$  as  $t \rightarrow \infty$

i.e.) for any  $\bar{\varepsilon}$ ,  $\exists \gamma > 0$  and  $T(\bar{\varepsilon}, \gamma, t_0) > 0$  such that

$\|x(t_0) - x_e\| \leq \gamma$  yields

$$\|x(t) - x_e\| \leq \bar{\varepsilon} \quad \forall t \geq t_0 + T$$

(2)  $x_e$  is uniformly asymptotically stable if

- $x_e$  is uniformly stable i.s.L. over  $[t_0, \infty)$
- $T$  is independent to  $t_0$



# General Definition of Internal Stability

## Example

$$\dot{x} = -\frac{1}{1+t}x(t) \rightarrow x(t) = \frac{x_0(1+t_0)}{1+t}$$

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

2)  $\|x_0\| < \varepsilon \rightarrow \|x(t)\| < \frac{\varepsilon(1+t_0)}{1+t} \leq \varepsilon \rightarrow$  uniformly stable

3) for  $\|x(t_0 + T)\| = \frac{x_0(1+t_0)}{1+t_0+T} < \varepsilon$

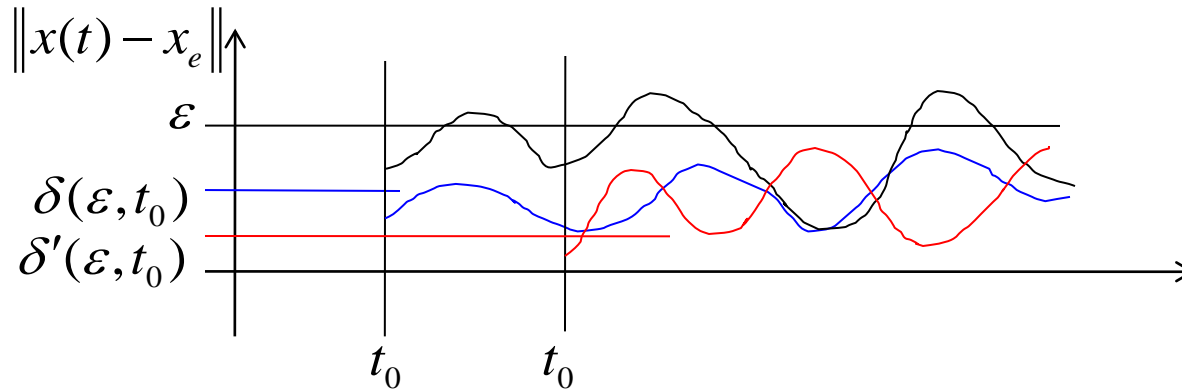
$$T > \frac{x_0(1+t_0)}{\varepsilon} - 1 - t_0 = \left( \left( \frac{x_0}{\varepsilon} - 1 \right) (1+t_0) \right)$$

$\Rightarrow$  not uniformly asymptotically stable.

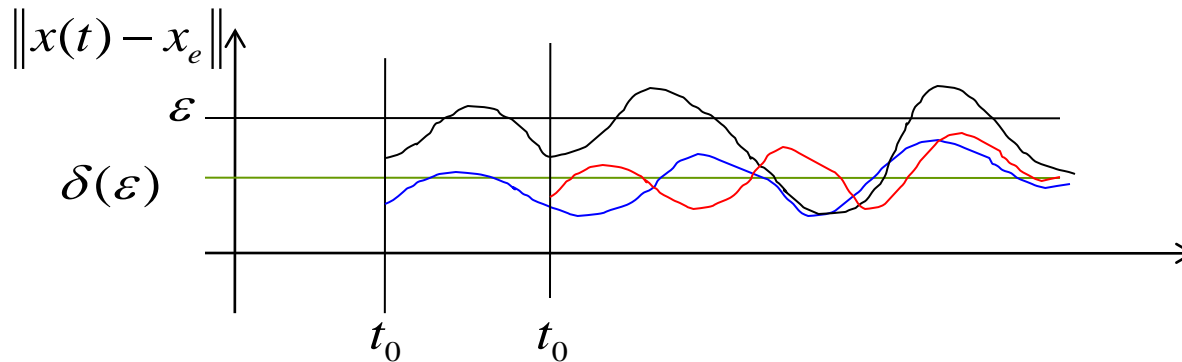


# Review

## Stable i.s.L.

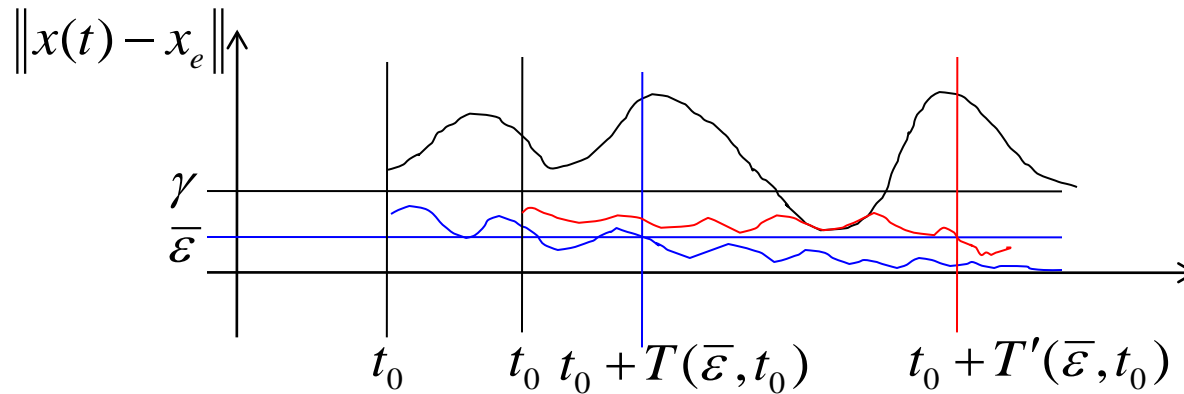


## Uniformly Stable i.s.L.

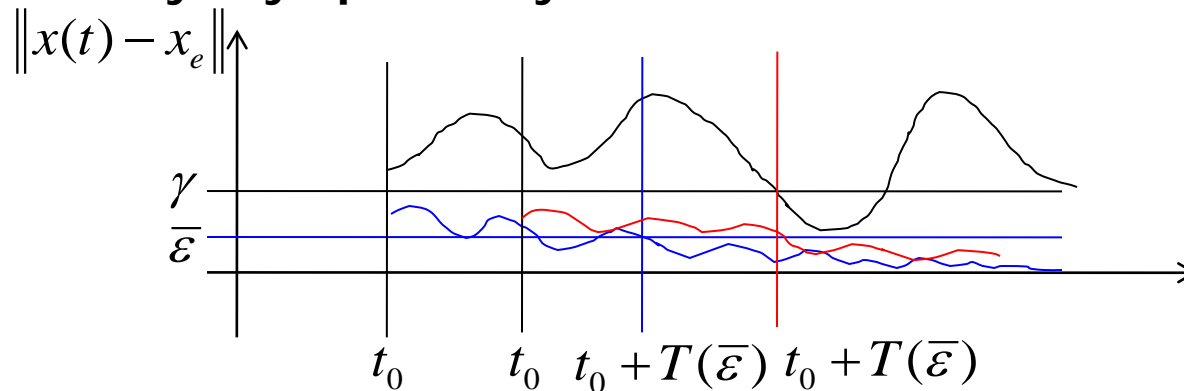


# Review

## Asymptotically Stable i.s.L.



## Uniformly Asymptotically Stable i.s.L.





## HW 5-1

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**Problem:** Determine the eigenvalues and stability of the eq. and discuss the relation between eigenvalues and stability of the time varying system.

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

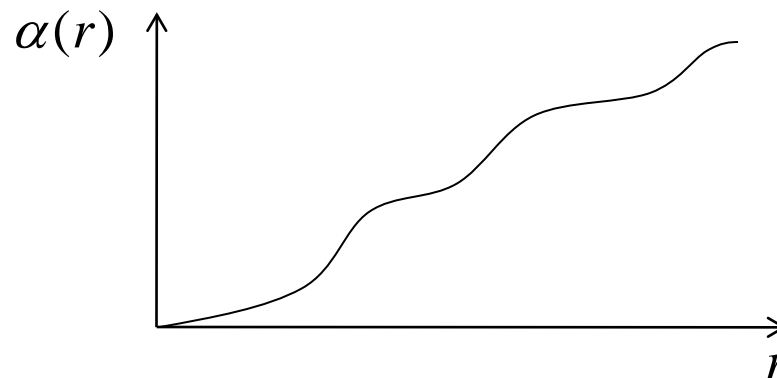
# Lyapunov Stability Theory

## Lyapunov Theory

**Definition:** class  $K$  functions [Hahn, 1967]

Function  $\alpha(r) : R^+ \rightarrow R^+$  belongs to class  $K$  if

- $\alpha(0) = 0$
- continuous
- strictly increasing.





# Lyapunov Stability Theory

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**Definition:** locally positive definite functions

Continuous function  $V(x, t) : B_h \subset \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is

*l.p.d.f.* if for some  $h, \alpha \in \mathbb{K}$

$$\begin{cases} V(x, t) \geq \alpha(\|x\|) \quad \forall x \in B_h \subset \mathbb{R}^n, \quad B_h := \{x \mid x \in \mathbb{R}^n, \|x\| < h\} \\ V(o, t) = 0. \end{cases}$$

In case of  $B_h = \mathbb{R}^n$ ,  $V(x, t)$  is *globally p.d.f.*



# Lyapunov Stability Theory

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## **Definition:** Descrescent Function

$V(x, t)$  is descrescent if

$\exists \alpha(\cdot) \in K$  such that

$$V(x, t) \leq \alpha(\|x\|), \forall x \in R^n, t \geq 0$$

# Lyapunov Stability Theory

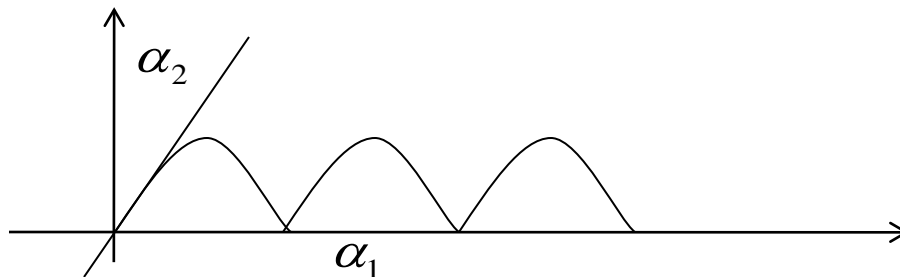
## Example

$$\left. \begin{aligned} V(x,t) &= \|x\|^2 = x^T x \\ V(x,t) &= x^T Mx, \quad M > 0 \end{aligned} \right\} \text{g.p.d.f. descrescent}$$

$$V(x,t) = (t+1)\|x\|^2 \quad : \text{g.p.d.f.}$$

$$V(x,t) = e^{-t}\|x\|^2 \quad : \text{descrescent}$$

$$V(x,t) = \sin^2(\|x\|^2) : \text{l.p.d.f. , descrescent in } B_h = \{x \mid \|x\| < \pi\}$$



(not K-class ftn bound because not strictly increasing)



# Lyapunov Stability Theory

## Theorem (Lyapunov Stability Theorem)

If  $\exists$  a continuously differentiable function  $V(x, t)$  such that

$\dot{x} = f(x, t)$ ,  $f(0, t) = 0$  ( $x(t) = 0$  is equilibrium point)

$$\alpha_1(\|x\|) \leq V(x, t) \leq \alpha_2(\|x\|) \quad \forall x \in B_\gamma \text{ (or } x \in R^n)$$

$$\dot{V} = \frac{dV}{dt} + \frac{dV}{dx} f(x, t) \leq -\alpha_3(\|x\|)$$

$$(\equiv \alpha_3(\|x\|) \leq -\dot{V})$$

then  $x(t) = 0$  is (globally) uniformly asymptotically stable.

*Note :*

$$\alpha_1(\|x\|) \leq V(x, t), \quad \alpha_3(\|x\|) \leq -\dot{V} \Rightarrow \text{asymptotically stable}$$

$$\alpha_1(\|x\|) \leq V(x, t), \quad 0 \leq -\dot{V} \Rightarrow \text{marginally stable (i.s.L)}$$



# Lyapunov Stability Theory

$V(x, t)$	$-\dot{V}(X, t)$	stability
p.d.f., descrescent	p.d.f.	G. U. A. S
l.p.d.f., descrescent	l.p.d.f.	U. A. S
l.p.d.f.,	l.p.d.f	A. S
l.p.d.f., descrescent	$\geq 0$ locally	U. . S i.s.L.
l.p.d.f.,	$\geq 0$ locally	S i.s.L.
p.d.f.,	p.d.f.	G. A. S
p.d.f.,	$\geq 0$	G. S i.s.L.
p.d.f., descrescent	$\geq 0$	G. U. . S i.s.L.



# Lyapunov Stability Theory

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Lemma : Barbalat's Lemma: (convergence)

If a real valued function  $g(t)$  is uniformly continuous  $\forall t \geq 0$  &

$$\lim_{t \rightarrow \infty} \int_0^t g(\tau) d\tau < \infty,$$

then  $\lim_{t \rightarrow \infty} g(t) = 0$ .



# Lyapunov Stability Theory

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**Theorem** : boundedness & convergence set

Suppose  $f(x, t)$  is locally Lipschitz on  $B_r \times \mathbb{R}_+$

Let  $V(x, t)$  be continuously differentiable function such that

$$\alpha_1(\|x\|) \leq V(x, t) \leq \alpha_2(\|x\|) \text{ (l.p.d.f. \& descrescent)}$$

and  $\dot{V} \leq -W(x) \leq 0$

Assume  $\dot{V}$  is uniformly continuous ( $\equiv \ddot{V}$  is bounded)

Then solutions of

$$\dot{x} = f(x, t), \quad \|x(t_0)\| \leq \alpha_2^{-1}(\alpha_1(r))$$

are bounded, i.e.

$$\|x\| \leq r,$$

and  $W(x(t)) \rightarrow 0$  as  $t \rightarrow \infty$ .



# Lyapunov Stability Theory

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Brief bf)

$$l = \min_{\|x\|=r} (V(x)) = \max_{\|x\|=\delta} (V(x))$$

$$\Rightarrow \|x(t_0)\| \leq \delta \rightarrow \|x\| \leq r$$

$$\alpha_1(r) = \min_{\|x\|=r} (V(x, t)) = \max_{\|x\|=\delta} (V(x, t)) = \alpha_2(\delta)$$

$$\Rightarrow \delta = \alpha_2^{-1}(\alpha_1(r))$$

$$\therefore \|x(t_0)\| \leq \delta = \alpha_2^{-1}(\alpha_1(r))$$

$$\Rightarrow \|x\| \leq r$$



# Lyapunov Stability Theory

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Brief bf(continued)

Since  $V(x, t)$  is decrescent,

$V(x, t)$  is bounded for bounded  $x$ , hence

$$\lim_{t \rightarrow \infty} V(x, t) = \lim_{t \rightarrow \infty} \int_0^t \dot{V}(\tau) d\tau < \infty.$$

Since  $\dot{V}$  is uniformly continuous

$$\lim_{t \rightarrow \infty} \dot{V}(t) = 0 \leq -\lim_{t \rightarrow \infty} W(x) \leq 0$$

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

If  $W(x) \in K$ ,  $W(x) = 0 \rightarrow x = 0$ .

Hence  $\lim_{t \rightarrow \infty} x(t) = 0$  ( $x(t) \rightarrow 0$ , A. S.)



# Lyapunov Stability Theory

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Example

$$\dot{x}_1 = x_2 + cx_1(x_1^2 + x_2^2)$$

$$\dot{x}_2 = -x_1 + cx_2(x_1^2 + x_2^2)$$

Is this system stable ? How can we determine the stability?

The candidate for Lyapunov function

$$V(x) = x_1^2 + x_2^2, \leftarrow \text{p.d.f and decrescent.}$$

$$\dot{V} = 2c(x_1^2 + x_2^2)^2.$$

If  $c = 0$ ,  $\dot{V} = 0$ , and therefore  $x_e = 0$  u.s..

If  $c < 0$ ,  $\dot{V} \leq -\alpha(\|x\|)$ , and therefore  $x_e = 0$  g.u.a.s..

If  $c > 0$ ,  $\dot{V} \geq \alpha(\|x\|)$ , and therefore  $x_e = 0$  unstable.



# Lyapunov Stability Theory

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Example

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -x_2 - e^{-t} x_1$$

Is this system stable ? How can we determine the stability?

The candidate for Lyapunov function

$$V(x) = x_1^2 + x_2^2,$$

$$\dot{V} = -2x_2^2 + 2x_1x_2(1 - e^{-t}).$$

We can not say anything.

$$V(x) = x_1^2 + e^t x_2^2,$$

$$\dot{V} = -e^t x_2^2.$$

If  $c = 0$ ,  $\dot{V} \leq 0$ , and therefore  $x_e = 0$  stable i.s.L.



## HW5-2

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Problem

$$\dot{x}_1 = -2x_1 + x_1x_2 + x_2$$

$$\dot{x}_2 = -x_1^2 - x_1$$

Determine the stability and discuss the convergence property.





# Lyapunov Stability Theory

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## Theorem 5.5

$\operatorname{Re}\{\lambda_i(A)\} < 0$  iff

for any  $N' = N > 0$ ,  $\exists$  unique  $M = M' > 0$

such that  $A'M + MA = -N$ .



# Lyapunov Stability Theory

Pf)

$$(\Leftarrow) \quad M > 0 \rightarrow \operatorname{Re}\{\lambda_i\} < 0$$

$$V(x) = x'Mx > 0, \text{ decrescent}$$

$$\begin{aligned} \dot{V}(x) &= \dot{x}'MX + x'M\dot{x} \\ &= x'A'Mx + x'MAx \\ &= -x'Nx \leq -\lambda_m(N)\|x\|^2 \end{aligned}$$

By Lyapunov Theorem,

$$\lim_{t \rightarrow \infty} x = 0.$$

By Theorem 5.4,  $\operatorname{Re}\{\lambda_i(A)\} < 0$ .

$$(\Rightarrow) \quad \operatorname{Re}\{\lambda_i\} < 0 \rightarrow M > 0$$

$$\operatorname{Re}\{\lambda_i(A)\} + \operatorname{Re}\{\lambda_j(A)\} \neq 0 < 0$$



# Lyapunov Stability Theory

Pf\_continued)

$\Rightarrow A'M + MA = -N$  has unique solution (see Section 3.7)

$M = \int_0^{\infty} e^{A't} N e^{At} dt$  is solution.

$$\begin{aligned} (\because A'M + MA &= \int_0^{\infty} A' e^{A't} N e^{At} dt + \int_0^{\infty} e^{A't} N e^{At} A dt \\ &= \int_0^{\infty} \frac{d}{dt} (e^{A't} N e^{At}) dt = e^{A't} N e^{At} \Big|_{t=0}^{\infty} = -N) \end{aligned}$$

and  $N = \bar{N}'\bar{N}$  ( $\because N$  is symmetric)

$$\rightarrow x'Mx = \int_0^{\infty} x' e^{A't} \bar{N}'\bar{N} e^{At} x dt = \int_0^{\infty} \left\| \bar{N} e^{At} x \right\|_2^2 dt > 0.$$

$\Rightarrow M > 0$ .



# Lyapunov Stability Theory

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Corollary 5.5

$\text{Re}\{\lambda_i(A)\} < 0$  iff

for any given  $m \times n$  matrix  $\bar{N}$  with  $m < n$ , together with the property

$$\text{rank}(\mathbf{O}) := \text{rank} \begin{bmatrix} \bar{N} \\ \bar{N}A \\ \dots \\ \bar{N}A^{n-1} \end{bmatrix} = n \text{ (full column rank)}$$

where  $\mathbf{O}$  is an  $nm \times n$  matrix, the Lyapunov equation

$$A'M + MA = -\bar{N}'\bar{N} = -N.$$

has an unique  $M = M' > 0$ .



# Lyapunov Stability Theory

Pf) Since  $N = \bar{N}'\bar{N}$  is positive semidefinite,

$\exists x \neq 0$  such that  $Ne^{At}x = 0$  in  $[0, \infty)$ . Hence

$$x'Mx = \int_0^\infty x'e^{A't}\bar{N}'\bar{N}e^{At}xdt = \int_0^\infty \|Ne^{At}x\|_2^2 dt \geq 0.$$

$\Rightarrow M \geq 0$ .

By derivative of  $Ne^{At}x = 0$ ,

$$\begin{bmatrix} \bar{N} \\ \bar{N}A \\ \bar{N}A^{n-1} \end{bmatrix} e^{At}x = Oe^{At}x = 0.$$

If  $O$  has full rank,  $x = 0$ . This implies

there is no  $x \neq 0$  such that  $Ne^{At}x = 0$  in  $[0, \infty)$ .

$\Rightarrow M > 0$  if  $O$  has full rank.



# HW5-3

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Problem 5.21 in the Text P. 142



# Lyapunov Stability Theory

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## Discrete-Time Case

### Theorem 5.D5

$|\lambda_i(A)| < 1$  iff for any  $N > 0$  or  $N = \bar{N}'\bar{N} \geq 0$  &  $\bar{N} = m \times n$   
with full rank  $O$ , the discrete Lyapunov equation

$$M - A'MA = N$$

has unique symmetric solution  $M > 0$ .



# Lyapunov Stability Theory

b.pf)  $\Rightarrow$  If  $|\lambda_i| < 1$ , the solution

$$M = \sum_{m=0}^{\infty} (A')^m N A^m \text{ is well defined.}$$

$$(\because \sum_{m=0}^{\infty} (A')^m N A^m - A' \sum_{m=0}^{\infty} (A')^m N A^{m+1}$$

$$= N + \sum_{m=1}^{\infty} (A')^m N A^m - A' \sum_{m=0}^{\infty} (A')^m N A^{m+1} = N)$$

Since  $N > 0$ , for  $A v = \lambda v$

$$\begin{aligned} v^* N v &= v^* M v - v^* A' M A v \\ &= v^* M v - \lambda^* v^* M v \lambda \\ &= (1 - \lambda^2) v^* M v > 0 \end{aligned}$$

$$v^* M v > 0 \rightarrow M > 0.$$





# Stability of Linear Time Varying System

Theorem

$x_e$  of  $\dot{x} = A(t)x$  is stable i.s.L at  $t_0$  iff

$\exists K(t_0) \ni$

$$\|\Phi(t, t_0)\| \leq K(t_0) < \infty \quad \forall t \geq t_0,$$

and  $x_e$  is uniformly stable i.s.L if  $K$  is independent of  $t_0$ .

Pf) ( $\Leftarrow$ )  $x_e = \Phi(t, t_0)x_e, \forall t \geq t_0$

$$x(t) - x_e = \Phi(t, t_0)(x_0 - x_e)$$

$$\|x(t) - x_e\| \leq \|\Phi(t, t_0)\| \|x_0 - x_e\| \leq K(t_0) \|x_0 - x_e\|$$

for any  $\varepsilon$ ,  $\exists \delta(t_0) > 0$  such that

$$\|x_0 - x_e\| \leq \frac{\varepsilon}{K(t_0)} = \delta(t_0) \rightarrow \|x(t) - x_e\| \leq \varepsilon.$$

( $\Rightarrow$ ) by contradiction, it can be easily shown.



# Stability of Linear Time Varying System

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## Theorem

Zero state of  $\dot{x} = A(t)x$  is asymptotically stable at  $t_0$

iff  $\|\Phi(t, t_0)\| \leq K(t_0) < \infty$  &  $\|\Phi(t, t_0)\| \rightarrow 0$  as  $t \rightarrow \infty$ .

And it is uniformly asymptotically stable over  $[t_0, \infty]$

iff  $\exists K_1 > K_2 > 0$  such that

$$\|\Phi(t, t_0)\| \leq K_1 e^{-K_2(t-t_0)} \quad \forall t \geq t_0.$$

## Note

$$AS \not\leftrightarrow BIBO \leftarrow y(t) = \int_{t_0}^t C(\tau)\Phi(\tau, t_0)B(\tau)u(\tau)d\tau$$



# Stability of Linear Time Varying System

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## Theorem 5.7

Marginal and asymptotical stabilities are invariant under any Lyapunov transformation.

Pf)

By Lyapunov transformation,

$$\bar{X}(t) = P(t)X(t)$$

$$\begin{aligned}\bar{\Phi}(t, \tau) &= \bar{X}(t)\bar{X}^{-1}(\tau) = P(t)X(t)X^{-1}(\tau)P^{-1}(\tau) \\ &= P(t)\Phi(t, \tau)P^{-1}(\tau)\end{aligned}$$

Since  $P(t)$  and  $P^{-1}(t)$  is bounded,  $\bar{\Phi}(t, \tau)$  is bounded.