

Spring 2019

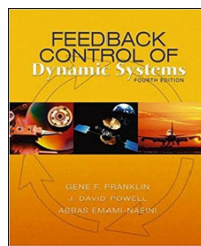
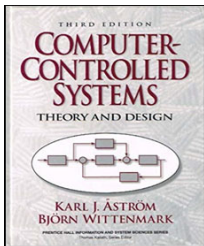
數位控制系統
Digital Control Systems

DCS-22
Stability

Feng-Li Lian

NTU-EE

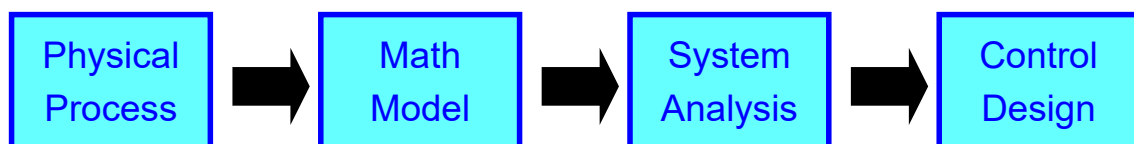
Feb19 – Jun19



Introduction: The Design Philosophy of Control Science

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DCS22-Stability-2

▪ The Research Procedure in Control Science



- Plant
- Sensor
- Actuator
- Computer
- Communication
- Noise
- Disturbance

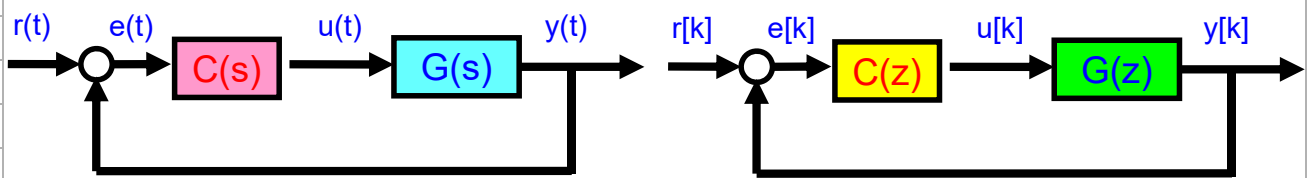
- Differential eqn
- Laplace transform
- Transfer function
- State space form

- Difference eqn
- z transform
- Transfer function
- State space form

- Root locus
- Bode diagram
- Nyquist plot
- Stability
- Robustness
- Sensitivity
- Controllability
- Observability

- Estimator
- Identification
- Regulation
- Tracking
- PID
- Pole placement
- Optimal Control LQR/LQG
- Adaptive control
- Robust control
- Decentralized (or Multi-person) Control

Introduction: From CT Plant to DT Plant



Plant (CT):

- Input-Output Model:

$$\begin{matrix} u(t) \\ y(t) \end{matrix}$$

$$G(s) = \frac{Y(s)}{U(s)}$$

- State-Space Model:

$$\begin{aligned} \dot{\mathbf{x}}(t) &= \mathbf{A}\mathbf{x}(t) + \mathbf{B}u(t) \\ y(t) &= \mathbf{C}\mathbf{x}(t) + \mathbf{D}u(t) \end{aligned}$$

Plant (DT):

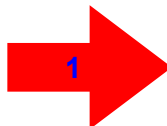
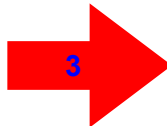
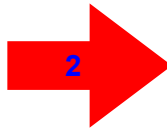
- Input-Output Model:

$$\begin{matrix} u[k] \\ y[k] \end{matrix}$$

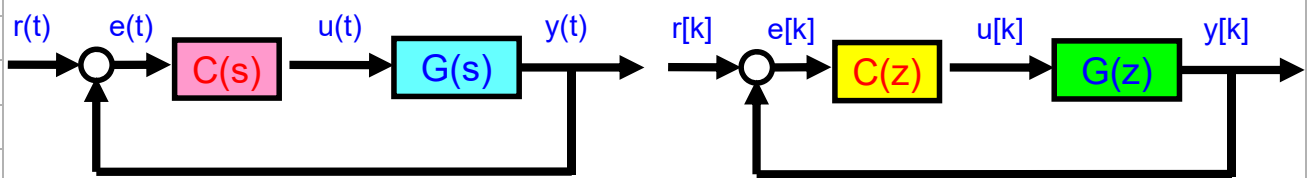
$$G(z) = \frac{Y(z)}{U(z)}$$

- State-Space Model:

$$\begin{aligned} \mathbf{x}[k+1] &= \mathbf{F}\mathbf{x}[k] + \mathbf{H}u[k] \\ y[k] &= \mathbf{C}\mathbf{x}[k] + \mathbf{D}u[k] \end{aligned}$$



Introduction: Model and Analysis



Plant (CT):

- Input-Output Model:

$$\frac{Y_c(s)}{U_c(s)} = G_c(s) = \frac{B_c(s)}{A_c(s)}$$

- State-Space Model:

$$\begin{aligned} \dot{\mathbf{x}}(t) &= \mathbf{A}\mathbf{x}(t) + \mathbf{B}u(t) \\ y(t) &= \mathbf{C}\mathbf{x}(t) + \mathbf{D}u(t) \end{aligned}$$

Plant (DT):

- Input-Output Model:

$$\frac{Y_d(z)}{U_d(z)} = G_d(z) = \frac{B_d(z)}{A_d(z)}$$

- State-Space Model:

$$\begin{aligned} \mathbf{x}[k+1] &= \mathbf{F}\mathbf{x}[k] + \mathbf{H}u[k] \\ y[k] &= \mathbf{C}\mathbf{x}[k] + \mathbf{D}u[k] \end{aligned}$$

System Properties:

- Stability
- Controllability and Reachability
- Observability and Detectability

- Solution of a System
- Stability and Asymptotic Stability
- Input-Output Stability
- Stability Tests:
 - Jury's Stability Criterion
 - Nyquist and Bode Diagrams
 - Nyquist Criterion
 - Relative Stability
- Lyapunov's Second Stability

Solution of a System

- CT: $\frac{d}{dt}x(t) = f(x(t), t)$
 - linear or nonlinear
 - time-invariant or time-varying
- DT: $x(k + 1) = f(x(k), k)$
- Initial Condition:
$$x_{10} = x_1(k_0)$$
$$x_{20} = x_2(k_0)$$
- Solution:
$$x_1(k) = p_1(k, k_0, x_{10})$$
$$x_2(k) = p_2(k, k_0, x_{20})$$

Definition 3.1: Stability

- $x_1(k)$ is stable

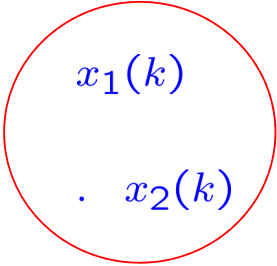
if for a given $\epsilon > 0$

there exists a $\delta(\epsilon, k_0)$

such that

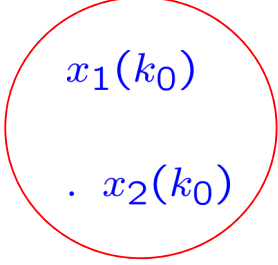
all solutions with $\|x_2(k_0) - x_1(k_0)\| < \delta$

$$\Rightarrow \|x_2(k) - x_1(k)\| < \epsilon, \quad \forall k \geq k_0$$



$x_1(k)$

$x_2(k)$



$x_1(k_0)$

$x_2(k_0)$

Definition 3.2: Asymptotic Stability

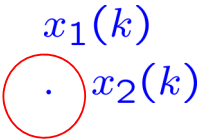
- $x_1(k)$ is asymptotic stable

if it is stable, and

if δ can be chosen

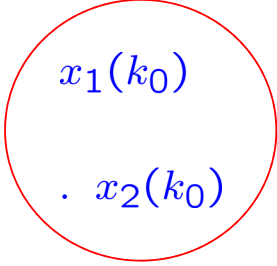
such that $\|x_2(k_0) - x_1(k_0)\| < \delta$

$$\Rightarrow \|x_2(k) - x_1(k)\| \rightarrow 0, \quad \text{when } k \rightarrow \infty$$



$x_1(k)$

$x_2(k)$



$x_1(k_0)$

$x_2(k_0)$

Stability of Linear Discrete-Time Systems

$$x_1(k+1) = \mathbf{F} x_1(k), \quad x_1(0) = a_1$$

$$x_2(k+1) = \mathbf{F} x_2(k), \quad x_2(0) = a_2$$

$$\Rightarrow \tilde{x} = x_1 - x_2$$

$$\Rightarrow x_1(k+1) - x_2(k+1) = \mathbf{F} x_1(k) - \mathbf{F} x_2(k)$$

$$\Rightarrow \tilde{x}(k+1) = \mathbf{F} \tilde{x}(k), \quad \tilde{x}(0) = a_1 - a_2$$

$$\Rightarrow \text{If } x_1 \text{ is stable}$$

$$\Rightarrow \text{every other solution is also stable}$$

$$\Rightarrow \text{Hence, for LTI systems,}$$

stability is a property of the system and not of a special solution

Solution of LTI DT Systems

$$\tilde{x}(k+1) = \mathbf{F} \tilde{x}(k), \quad \tilde{x}(0) = a_1 - a_2$$

$$\Rightarrow \tilde{x}(k) = \mathbf{F}^k \tilde{x}(0)$$

$$\text{Let } \lambda_i = \text{eig}(\mathbf{F})$$

$$\mathbf{F} = \mathbf{U} \begin{bmatrix} \lambda_1 & & * \\ & \dots & \\ 0 & & \lambda_n \end{bmatrix} \mathbf{U}^{-1}$$

$$\mathbf{F}^k = \mathbf{U} \begin{bmatrix} \lambda_1^k & & * \\ & \dots & \\ 0 & & \lambda_n^k \end{bmatrix} \mathbf{U}^{-1}$$

$$\text{Asymptotic stable} \Rightarrow |\lambda_i| < 1, \quad i = 1, \dots, n$$

■ Theorem 3.1: Asymptotic Stability of Linear Systems

- A DT LTI system is asymptotic stable
 - ⇔ all eig(**F**) are strictly inside the unit disc

■ Stability of Linear Continuous-Time Systems

$$\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t), \quad \mathbf{x}(t_0) = \mathbf{x}_0$$

$$\Rightarrow \mathbf{x}(t) = e^{(\mathbf{A}(t-t_0))} \mathbf{x}(t_0)$$

$$\text{Let } \lambda_i = \text{eig}(\mathbf{A}) \quad \Rightarrow \mathbf{A} = \mathbf{U} \begin{bmatrix} \lambda_1 & \cdots & * \\ 0 & & \lambda_n \end{bmatrix} \mathbf{U}^{-1}$$

$$\Rightarrow \mathbf{x}(t) = \mathbf{U} \begin{bmatrix} e^{\lambda_1(t-t_0)} & \cdots & * \\ 0 & & e^{\lambda_n(t-t_0)} \end{bmatrix} \mathbf{U}^{-1} \mathbf{x}(t_0)$$

$$\text{Asymptotic stable} \Rightarrow \text{Real}(\lambda_i) < 0, \quad i = 1, \dots, n$$

Definition 3.3: Bounded-Input-Bounded-Output Stability

- A LTI system is defined as **BIBO stable**
if a **bounded input** gives a **bounded output**
for every initial value

Theorem 3.2: Relation between Stability Concept

Asymptotic stable \Rightarrow stable and BIBO stable

Example 3.1: Harmonic Oscillator

$$x(k+1) = \begin{bmatrix} \cos wh & \sin wh \\ -\sin wh & \cos wh \end{bmatrix} x(k) + \begin{bmatrix} 1 - \cos wh \\ \sin wh \end{bmatrix} u(k)$$

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

- $\text{mag}(\text{eig}(\mathbf{F})) = 1$
- if $u(k) = 0 \Rightarrow \|x(k+1)\| = \|x(0)\|$
 \Rightarrow the system is **stable**
- But, if input is **a cos or sin signal with w rad/s**
 \Rightarrow the output contains a **sinusoidal** function
with **growing amplitude**
 \Rightarrow the system is **not BIBO stable**

- Eigenvalues of F

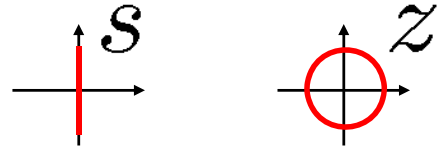
$$\lambda_i = \text{eig}(F)$$

- Characteristic Polynomials

$$A(z) = a_0 z^n + a_1 z^{n-1} + \dots + a_n$$

$$a_i \Leftrightarrow \lambda_i$$

- Root locus method



- Nyquist criterion

- Lyapunov's method

Stability Test: Jury's Stability Criterion

(1918) (1922) (1961)

Schur-Cohn-Jury

$$A(z) = a_0 z^n + a_1 z^{n-1} + \dots + a_n = 0$$

$$a_0 \quad a_1 \quad \dots \quad a_{n-1} \quad a_n$$

$$a_n \quad a_{n-1} \quad \dots \quad a_1 \quad a_0$$

$$\alpha_n = \frac{a_n}{a_0}$$

$$a_0^{n-1} \quad a_1^{n-1} \quad \dots \quad a_{n-1}^{n-1}$$

$$a_{n-1}^{n-1} \quad a_{n-2}^{n-1} \quad \dots \quad a_0^{n-1}$$

$$\alpha_{n-1} = \frac{a_{n-1}^{n-1}}{a_0^{n-1}}$$

⋮

$$a_0^0$$

$$a_i^{k-1} = a_i^k - \alpha_k a_{k-i}^k$$

$$\alpha_k = a_k^k / a_0^k$$

3.6.2 Routh's Stability Criterion

- Routh in 1874
- Hurwitz in 1895

$$a(s) = s^n + a_1s^{n-1} + a_2s^{n-2} + \dots + a_{n-1}s + a_n. \quad (3.65)$$

A necessary (but not sufficient) condition for stability is that *all* the coefficients of the characteristic polynomial be positive.

A system is stable if and only if *all* the elements in the first column of the Routh array are positive.

We then add subsequent rows to complete the **Routh array**:

Row	n	s^n :	1	a_2	a_4	\dots
Row	$n-1$	s^{n-1} :	a_1	a_3	a_5	\dots
Row	$n-2$	s^{n-2} :	b_1	b_2	b_3	\dots
Row	$n-3$	s^{n-3} :	c_1	c_2	c_3	\dots
	\vdots	\vdots	\vdots	\vdots	\vdots	
Row	2	s^2 :	*	*		
Row	1	s :	*			
Row	0	s^0 :	*			

$$b_1 = -\frac{\det \begin{bmatrix} 1 & a_2 \\ a_1 & a_3 \end{bmatrix}}{a_1} = \frac{a_1a_2 - a_3}{a_1},$$

$$b_2 = -\frac{\det \begin{bmatrix} 1 & a_4 \\ a_1 & a_5 \end{bmatrix}}{a_1} = \frac{a_1a_4 - a_5}{a_1},$$

$$b_3 = -\frac{\det \begin{bmatrix} 1 & a_6 \\ a_1 & a_7 \end{bmatrix}}{a_1} = \frac{a_1a_6 - a_7}{a_1},$$

$$c_1 = -\frac{\det \begin{bmatrix} a_1 & a_3 \\ b_1 & b_2 \end{bmatrix}}{b_1} = \frac{b_1a_3 - a_1b_2}{b_1},$$

$$c_2 = -\frac{\det \begin{bmatrix} a_1 & a_5 \\ b_1 & b_3 \end{bmatrix}}{b_1} = \frac{b_1a_5 - a_1b_3}{b_1},$$

$$c_3 = -\frac{\det \begin{bmatrix} a_1 & a_7 \\ b_1 & b_4 \end{bmatrix}}{b_1} = \frac{b_1a_7 - a_1b_4}{b_1}.$$

Franklin, Powell, Emami-Naeini 2002

Stability Test: Jury's Stability Criterion

▪ Theorem 3.3: Jury's Stability Test

- If $a_0 > 0$,
then, $A(z) = 0$ has all roots **inside unit disc**
 \iff all $a_0^k > 0, k = 0, 1, \dots, n-1$
- If **no** a_0 is zero,
then, the number of **negative** a_0^k
= the number of roots **outside the unit disc**
- **Remark:**
- If all $a_0^k > 0$,
then,

$$a_0^0 > 0 \iff \begin{cases} A(1) > 0 \\ (-1)^n A(-1) > 0 \end{cases}$$

Example: Jury's Stability Test

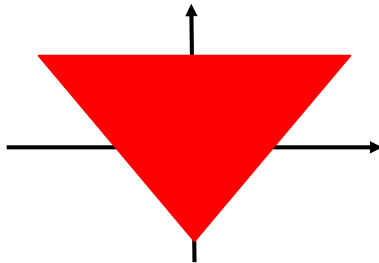
$$A(z) = z^2 + a_1z + a_2$$

$$\begin{array}{ccc} \boxed{1} & a_1 & a_2 \\ a_2 & a_1 & 1 \end{array}$$

$$\begin{array}{ccc} \boxed{1 - a_2^2} & a_1(1 - a_2) & \\ a_1(1 - a_2) & 1 - a_2^2 & \end{array}$$

$$\boxed{1 - a_2^2 - \frac{a_1^2(1 - a_2)}{1 + a_2}}$$

All the roots are inside the unit circle if



$$\alpha_n = \frac{a_n}{a_0} \quad \alpha_i^{k-1} = a_i^k - \alpha_k a_{k-i}^k$$

$$\alpha_k = a_k/a_0^k$$

$$\alpha_2 = a_2$$

$$\alpha_1 = \frac{a_1}{1 + a_2}$$

$$\begin{aligned} 1 - a_2^2 &> 0 \\ 1 - a_2^2 - \frac{a_1^2(1 - a_2)}{1 + a_2} &> 0 \\ \Rightarrow a_2 &< 1 \\ a_2 &> -1 + a_1 \\ a_2 &> -1 - a_1 \end{aligned}$$

Stability Test: Nyquist and Bode Diagrams

- DT pulse-transfer function: $G(z)$
- Nyquist or Frequency curve

$$G(e^{jwh}), \text{ for } wh \in [0, \pi]$$

upto to the Nyquist frequency, $w_N = \pi/h$

- Note that it is sufficient to consider the map in $wh \in [-\pi, \pi]$
- Because $G(e^{jwh})$ is periodic with period $2\pi/h$

▪ Example 3.3: Frequency Responses

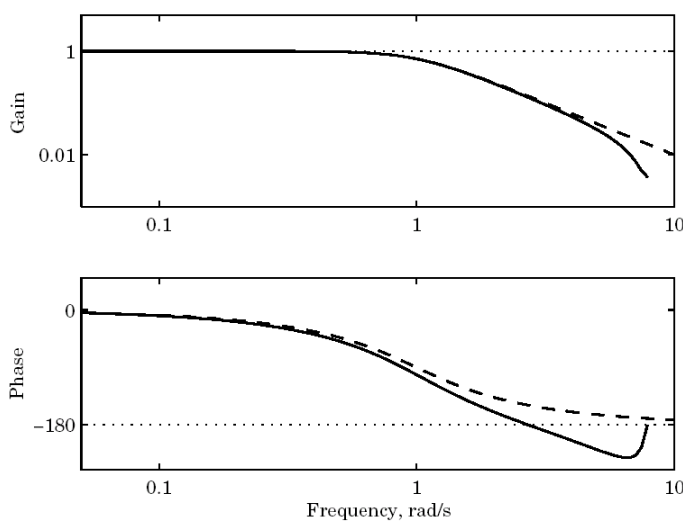
$$G(s) = \frac{1}{s^2 + 1.4s + 1}$$

Zero-order hold sampling $h = 0.4$

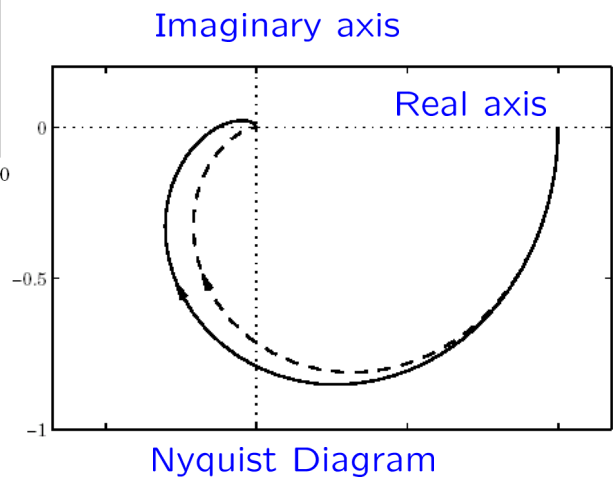
$$G(z) = \frac{0.066z + 0.055}{z^2 - 1.450z + 0.571}$$

▪ Example 3.3: Frequency Responses

Bode Diagram

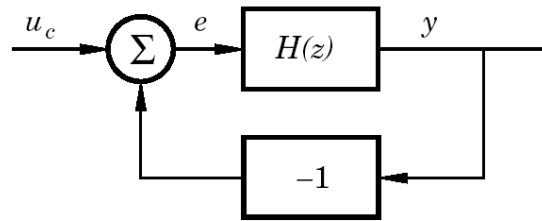


- CT: - - -
- DT: —



Nyquist Diagram

▪ Nyquist Criterion



Closed-loop system

$$Y(z) = H_{cl}(z)U_c(z) = \frac{H(z)}{1 + H(z)} U_c(z)$$

Closed-loop system characteristic equation

$$1 + H(z) = 0$$

Franklin, Powell, Emami-Naeini 2002

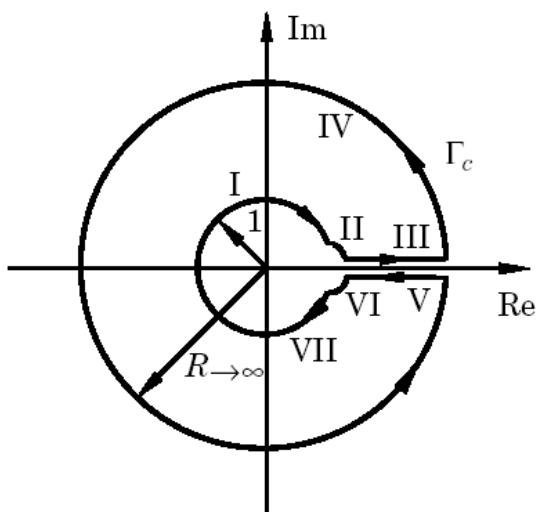
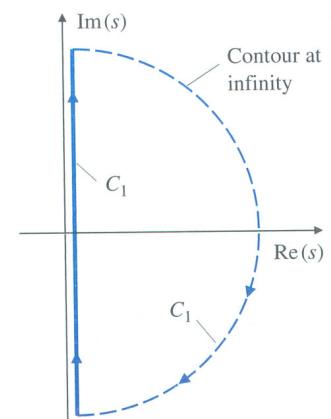


Figure 6.17

An *s*-plane plot of a contour C_1 that encircles the entire RHP



Principle of arguments states

$$N = Z - P$$

Z and P are the number of zeros and poles of $1 + H(z)$ outside the unit disc.

▪ Example 3.4: A Second-order system

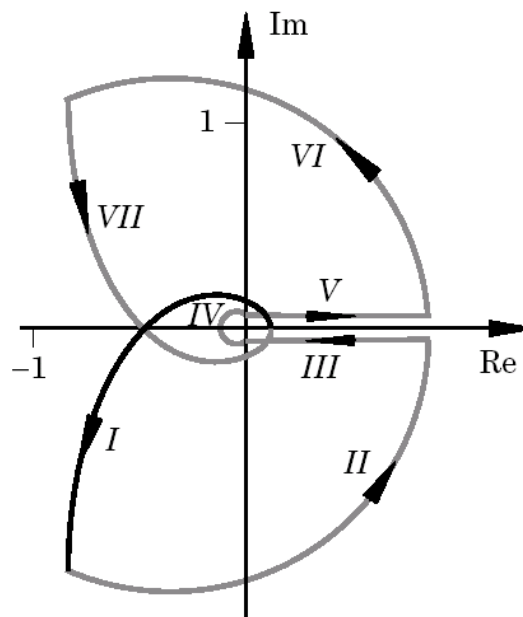
$$h = 1$$

$$H(z) = \frac{0.25K}{(z - 1)(z - 0.5)}$$

then

$$H(e^{i\omega}) = \frac{0.25K (1.5(1 - \cos \omega) - 2 \sin^2 \omega - i \sin \omega(2 \cos \omega - 1.5))}{(2 - 2 \cos \omega)(1.25 + \cos \omega)}$$

▪ Example 3.4: A Second-order system



$$H(e^{j\omega}), \text{ for } \omega \in [0, \pi]$$

- At some ω , phase shift $> 180^\circ$
- Stable if $K > 2$

- Definitions 3.4 & 3.5: Gain & Phase Margins

- The **amplitude** or **gain margin**:

$$\arg G(e^{i\omega_0 h}) = -\pi \quad A_{\text{marg}} = \frac{1}{|G(e^{i\omega_0 h})|}$$

- The **phase margin**:

$$|G(e^{i\omega_c h})| = 1 \quad \phi_{\text{marg}} = \pi + \arg G(e^{i\omega_c h})$$

Lyapunov's Second Stability

- Definition 3.6: Lyapunov Function

- $V(x)$ is a **Lyapunov function** for

$$x(k+1) = f(x(k)) \quad f(0) = 0$$

- If:

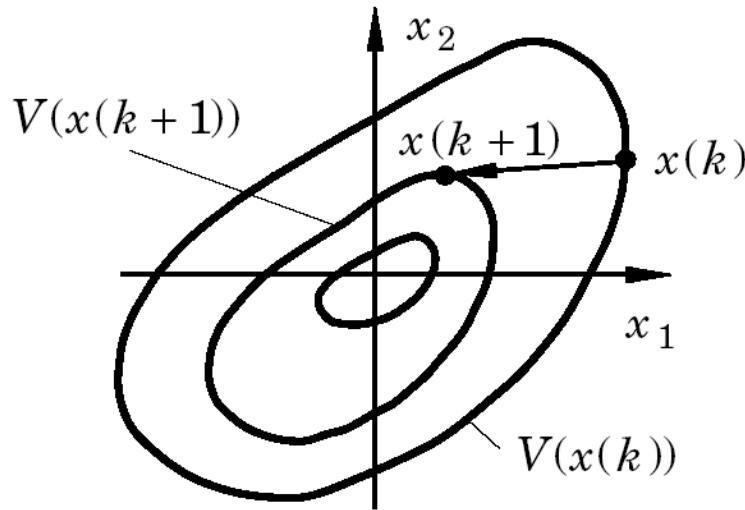
1. $V(x)$ is **continuous** in x and $V(0) = 0$
2. $V(x)$ is **positive definite**
3. $\Delta V(x) = V(f(x)) - V(x)$
is **negative definite**
4. $V(x) \rightarrow \infty$ as $|x| \rightarrow \infty$

- **Existence** of Lyapunov function implies **asymptotic stability** for the solution $x = 0$

▪ Geometric Illustration

1. $V(x)$ is continuous in x and $V(0) = 0$
2. $V(x)$ is positive definite
3. $\Delta V(x) = V(f(x)) - V(x)$ is negative definite
4. $V(x) \rightarrow \infty$ as $|x| \rightarrow \infty$

$$x(k + 1) = f(x(k)), \quad f(0) = 0$$



▪ Example 3.6: Lyapunov function

1. $V(x)$ is continuous in x and $V(0) = 0$
2. $V(x)$ is positive definite
3. $\Delta V(x) = V(f(x)) - V(x)$ is negative definite
4. $V(x) \rightarrow \infty$ as $|x| \rightarrow \infty$

$$x(k + 1) = \mathbf{F}x(k)$$

$$V(x) = x^T \mathbf{P}x \quad \mathbf{P} > 0$$

$$\Delta V(x) = V(x(k + 1)) - V(x(k))$$

$$= V(\mathbf{F}x(k)) - V(x(k))$$

$$= (\mathbf{F}x(k))^T \mathbf{P}(\mathbf{F}x(k)) - x^T \mathbf{P}x$$

$$= x^T \mathbf{F}^T \mathbf{P} \mathbf{F} x - x^T \mathbf{P} x$$

$$= x^T (\mathbf{F}^T \mathbf{P} \mathbf{F} - \mathbf{P}) x = x^T (-\mathbf{Q}) x = -x^T (\mathbf{Q}) x$$

V is a Lyapunov function

iff there exists a $\mathbf{P} > 0$

$$\mathbf{F}^T \mathbf{P} \mathbf{F} - \mathbf{P} = -\mathbf{Q} \quad \mathbf{Q} > 0$$

that satisfies the Lyapunov equation

- Example 3.6: Lyapunov function for CT case

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

$$V(\mathbf{x}) = \mathbf{x}^T \mathbf{P} \mathbf{x} \quad \mathbf{P} > 0$$

$$\begin{aligned} \dot{V}(\mathbf{x}) &= \mathbf{x}^T \mathbf{P} \dot{\mathbf{x}} + \dot{\mathbf{x}}^T \mathbf{P} \mathbf{x} \\ &= \mathbf{x}^T \mathbf{P} \mathbf{A} \mathbf{x} + (\mathbf{A} \mathbf{x})^T \mathbf{P} \mathbf{x} \end{aligned}$$

$$= \mathbf{x}^T \mathbf{P} \mathbf{A} \mathbf{x} + \mathbf{x}^T \mathbf{A}^T \mathbf{P} \mathbf{x}$$

$$= \mathbf{x}^T (\mathbf{P} \mathbf{A} + \mathbf{A}^T \mathbf{P}) \mathbf{x}$$

$$= \mathbf{x}^T (-\mathbf{Q}) \mathbf{x}$$

$$= -\mathbf{x}^T (\mathbf{Q}) \mathbf{x}$$

$$\mathbf{P} \mathbf{A} + \mathbf{A}^T \mathbf{P} = -\mathbf{Q}$$

Khalil 2002

- Example 3.6: Lyapunov function

$$\Phi^T \mathbf{P} \Phi - \mathbf{P} = -\mathbf{Q} \quad \mathbf{Q} > 0$$

$$\Phi = \begin{bmatrix} 0.4 & 0 \\ -0.4 & 0.6 \end{bmatrix} \quad \mathbf{Q} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \Rightarrow \mathbf{P} = \begin{bmatrix} 1.19 & -0.25 \\ -0.25 & 2.05 \end{bmatrix}$$

