

# Chapter 12

## Robustness Analysis 1 Small Gain Principle

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# Small Gain Theorem

## Theorem 1 (Small Gain Theorem)

*Assume that  $M(s), \Delta(s)$  are stable. The closed-loop system is robustly stable iff one of the following conditions is true.*

- *When  $\|\Delta\|_{\infty} \leq 1$ , there holds  $\|M(s)\|_{\infty} < 1$ .*
- *When  $\|\Delta\|_{\infty} < 1$ , there holds  $\|M(s)\|_{\infty} \leq 1$ .*

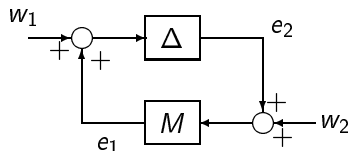


Figure: Small Gain Theorem

# Small Gain Theorem: proof

Sufficiency:

under the given condition, the loop gain satisfies

$$|M(j\omega)\Delta(j\omega)| = |M(j\omega)||\Delta(j\omega)| < 1.$$

It is contained inside the unit disk and does not encircle  $(-1, j0)$ . So, CLS is stable.

Necessity:

- Basic idea: find an uncertainty in the given set which destabilizes CLS if  $\|M\|_{\infty} \geq 1$ .
- Due to  $\|M\|_{\infty} \geq 1$  and the continuity of frequency response, there must be a frequency  $\omega_0 \in [0, \infty]$  at which  $|M(j\omega_0)| = 1$  holds, i.e.

$$M(j\omega_0) = e^{j\theta} \text{ or } -e^{j\theta}, \quad \theta \in [0, \pi).$$

- In the sequel,  $\Delta$  is constructed only for the case of positive sign.

# Small Gain Theorem: proof

- If we can find an uncertainty  $\Delta(s)$  satisfying

$$\Delta(j\omega_0) = e^{-j\theta}, \quad \|\Delta\|_\infty \leq 1,$$

- Then

$$1 - M(j\omega_0)\Delta(j\omega_0) = 1 - e^{j\theta}e^{-j\theta} = 0$$

holds and CLS has an unstable pole  $j\omega_0$ .

- Next, we construct stable rational uncertainties satisfying this condition case by case.

- 1  $M(j\omega_0) = 1$  when  $\theta = 0$ . Then,  $\Delta(s) = 1$ .
- 2 When  $\theta \in (0, \pi)$ ,

$$\Delta(s) = \frac{a-s}{a+s}, \quad a = \frac{\omega_0}{\tan \theta/2} > 0$$

satisfies  $\Delta(j\omega_0) = e^{-j\theta}$  and  $\|\Delta\|_\infty = 1$  simultaneously.

- 3 Uncertainties constructed are all stable and rational, and they belong to the given uncertainty set since their  $\mathcal{H}_\infty$  norms are 1.

# Remarks

- ➊ Physically, small gain theorem corresponds to the fact that the external input is attenuated every time it circulates in the closed loop.
- ➋ Small gain theorem is not only sufficient, but also necessary. That is, for a norm-bounded dynamic uncertainty set, small gain theorem is not conservative.
- ➌ However, we should understand this necessity correctly. It is true only when the phase of uncertainty can change freely, which seldom happens in practice.
- ➍ For real-world systems small gain theorem is simply sufficient, not necessary!

# Robust Stability Criteria

- Sensitivity  $S$  and Complementary sensitivity  $T$ :

$$S(s) = (I + PK)^{-1}, \quad T(s) = (I + PK)^{-1}PK. \quad (1)$$

Table: Robust Stability Criteria

$W(s)$ and $\Delta(s)$ are stable, $\ \Delta\ _\infty \leq 1$	
Plant set $\mathbb{P}$	Robust stability criterion
$(I + \Delta W)P$	Nominal stability and $\ WT\ _\infty < 1$
$(I + \Delta W)^{-1}P$	Nominal stability and $\ WS\ _\infty < 1$
$P + \Delta W$	Nominal stability and $\ WKS\ _\infty < 1$
$P(I + \Delta WP)^{-1}$	Nominal stability and $\ WSP\ _\infty < 1$

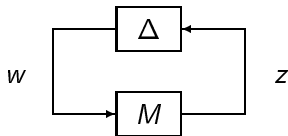
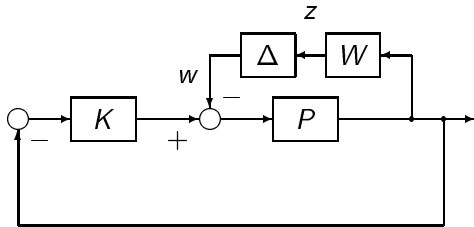
# Proof: feedback uncertainty case

- 1 When  $\Delta(s) = 0$ , the system must be nominally stable.
- 2 Denote the input and output of  $\Delta$  as  $z, w$  resp.
- 3 Compute the transfer matrix from  $w$  to  $z$

$$z = Mw, \quad M = -WSP.$$

- 4 Transform CLS to the left figure.
- 5 By small gain theorem the robust stability condition is

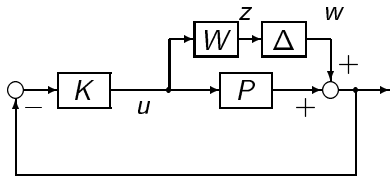
$$1 > \|M\|_{\infty} = \|-WSP\|_{\infty} = \|WSP\|_{\infty}.$$



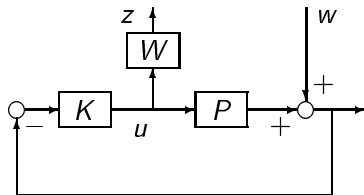
# Bridge over performance and robust stability

Example 1: Robust stability about additive uncertainty (Figure (a)),  $\|\Delta\|_\infty \leq 1$ .

Robust stability  $\Leftrightarrow (P, K)$  is internal stable and  $\|WKS\|_\infty < 1$   
 $\Leftrightarrow$  Suppress the influence of disturbance  $w$  on input  $u$  (Figure(b))



(a) Robust stability problem



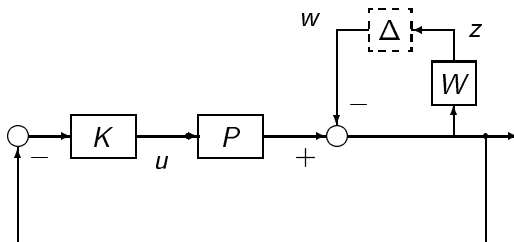
(b) Equivalent disturbance attenuation problem

# Bridge over performance and robust stability

Example 2: Sensitivity reduction:

$(P, K)$  is stable and  $\|WS\|_\infty < 1 \Leftrightarrow$

Robustly stabilize plant set  $\{\tilde{P} = \frac{P}{1 + \Delta W}, \|\Delta\|_\infty \leq 1\}$



- An  $\mathcal{H}_\infty$  performance problem is equivalent to a robust stability problem with a virtual uncertainty  $\Delta$  inserted between the input and output of the closed-loop transfer function.

# Bridge over performance and robust stability

## Theorem 2

- (1) *The CLS containing stable uncertainty  $\|\Delta\|_\infty \leq 1$  is stable iff the nominal CLS is stable and  $\|\mathcal{F}_\ell(G, K)\|_\infty < 1$ .*
- (2) *The nominal CLS is stable and  $\|\mathcal{F}_\ell(G, K)\|_\infty < 1$  iff the CLS formed by inserting an arbitrary virtual stable uncertainty  $\|\Delta\|_\infty \leq 1$  between its input  $w$  and output  $w$  is stable.*

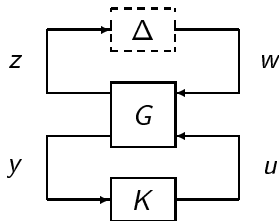


Figure: Equivalence between nominal performance and robust stability

# Analysis of Robust Performance

Example 1:

- 1 plant set

$$\tilde{P} = P + \Delta W, \quad \|\Delta\|_{\infty} \leq 1.$$

- 2 Spec: reduce the tracking error

$$\left\| W_S \frac{1}{1 + (P + \Delta W)K} \right\|_{\infty} < 1. \quad (2)$$

- 3 Nominal system must satisfy this spec first

$$\|W_S S\|_{\infty} < 1.$$

- 4 Robust stability

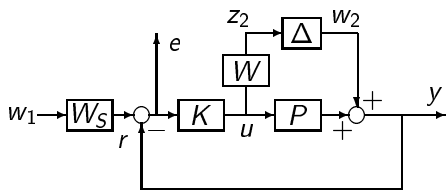
$$\|WKS\|_{\infty} < 1.$$

# Analysis of Robust Performance

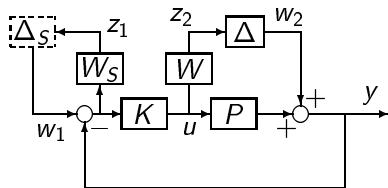
- 1 Examine if nominal performance and robust stability can guarantee the robust performance.

$$W_S \frac{1}{1 + PK} \times \frac{1}{1 + \Delta WKS} = W_S S(1 + \Delta WKS)^{-1}.$$

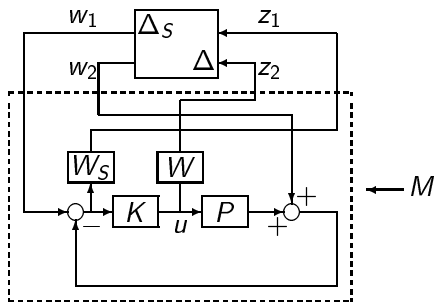
- 2  $\Delta$  ( $\|\Delta\|_\infty \leq 1$ ) can take any complex value. So even if  $\|WKS\|_\infty < 1$ , such a frequency still can be found at which  $|1 + \Delta WKS| \ll 1$  holds.
- 3 For this uncertainty  $\Delta$ , the tracking performance deteriorates significantly.
- 4 No matter how good the nominal performance ( $\|W_S S\|_\infty$ ) and robust stability are, the robust performance cannot be ensured!



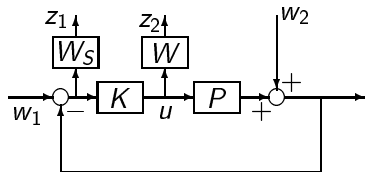
(a) Original problem



(b) Equivalent stability problem



(c) Separation of uncertainty



(d) Conversion to disturbance control problem

# Sufficient Condition for Robust Performance

Example 2: Derive a condition for the robust tracking.

- 1 Robust performance problem is equivalent to the robust stability problem when a virtual uncertainty  $\Delta_S$  ( $\|\Delta_S\|_\infty \leq 1$ ) (Figure (b))
- 2 After transforming Figure (b) into Figure (c), the problem is reduced to the robust stability problem of a CLS with a dilated uncertainty:

$$\begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} \Delta_S & \\ & \Delta \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}.$$

- 3 Since  $\left\| \begin{bmatrix} \Delta_S & \\ & \Delta \end{bmatrix} \right\|_\infty < 1$ , by small gain theorem a sufficient condition for robust stability is that, transfer matrix  $M(s)$  from  $[w_1 \ w_2]^T$  to  $[z_1 \ z_2]^T$  satisfies

$$\|M\|_\infty \leq 1, \quad M(s) = \begin{bmatrix} W_S S & -W_S S \\ W_K S & -W_K S \end{bmatrix}.$$

- 4 This is just a sufficient condition.

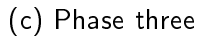
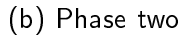
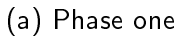
# Introduction of Scaling

## Example 2: continued

- 1 Introduction of minimum phase transfer function does not change the stability of CLS.
- 2 Equivalent block diagram transformations:  $(a) \rightarrow (b) \rightarrow (c)$
- 3 Scaled norm condition for robust performance

$$\|D^{-1}MD\|_{\infty} \leq 1, \quad D(s) = \begin{bmatrix} D_1(s) & \\ & D_2(s) \end{bmatrix}. \quad (3)$$

- 4 A suitable  $D$  may make  $\|D^{-1}MD\|_{\infty}$  less than  $\|M\|_{\infty}$ , thus lower the conservatism.



# Stability Radius of Norm-Bounded Parametric Systems

## Theorem 3 (Qiu's Theorem)

*The uncertain system with real uncertainty  $\Delta \in \mathbb{R}^{p \times q}$  is robustly stable iff the parameter uncertainty matrix  $\Delta$  satisfies*

$$\frac{1}{\|\Delta\|_2} > \sup_{\omega} \inf_{\gamma \in (0,1]} \sigma_2 \left( \begin{bmatrix} \Re(M(j\omega)) & -\gamma \Im(M(j\omega)) \\ \frac{1}{\gamma} \Im(M(j\omega)) & \Re(M(j\omega)) \end{bmatrix} \right). \quad (4)$$

$\sigma_2(X)$  denotes the second largest singular value of matrix  $X$ .

