

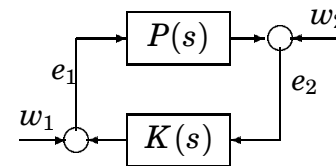
## Lecture 2

- Well-posedness and internal stability.
- Coprime factorization over  $H_\infty$ .
- Performance specifications in terms of  $H_2$  and  $H_\infty$  norms.

## Well-Posedness

Even for a matrix equation  $Ax = b$ , the solution  $x$  does not always exist.

Feedback gives a linear equation in an infinite-dimensional space. Solvability?

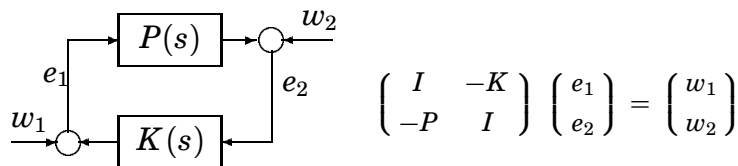


$$e_1 = Ke_2 + w_1$$

$$e_2 = Pe_1 + w_2$$

**Example:** Let  $P(s) = \frac{s+1}{s+2}$  and  $K(s) = 1$ . The closed-loop system is not proper

$$\frac{1}{1 - \frac{s+1}{s+2}} = \frac{s+2}{s+2-s-1} = s+2.$$



$$\begin{pmatrix} I & -K \\ -P & I \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} = \begin{pmatrix} w_1 \\ w_2 \end{pmatrix}$$

The system is solvable if the matrix of the system is invertible for almost all  $s$ . Then

$$\begin{pmatrix} e_1 \\ e_2 \end{pmatrix} = \begin{pmatrix} I & -K \\ -P & I \end{pmatrix}^{-1} \begin{pmatrix} w_1 \\ w_2 \end{pmatrix}$$

**Definition:** The closed-loop system is called *well-posed* if

$$\begin{pmatrix} I & -K \\ -P & I \end{pmatrix}^{-1}$$

exists for almost all  $s$  and is a proper function.

**Lemma:** Let  $G$  be proper and square. Then  $G^{-1}$  exists for almost all  $s$  and is proper if and only if  $G(\infty)$  is nonsingular.

**Proof:** Let  $G(s) = C(sI - A)^{-1}B + D$ . Hence  $G(\infty) = D$ .

“ $\Rightarrow$ ”

$G^{-1}$  exists and is proper  $\Rightarrow G(\infty)^{-1}$  exists and is bounded  $\Rightarrow G(\infty)$  is nonsingular.

“ $\Leftarrow$ ”

Calculate the inverse by [Zhou, p. 14]

$$\begin{aligned} G(s)^{-1} &= (D + C(sI - A)^{-1}B)^{-1} = \\ &= D^{-1} - D^{-1}C(sI - A + BD^{-1}C)^{-1}BD^{-1}. \end{aligned}$$

Hence, the inverse exists for almost all  $s$  (except the eigenvalues of the matrix  $A - BD^{-1}C$ ) and is proper.

**Corollary:** The following statements are equivalent

1. The closed-loop system  $(P, K)$  is well-posed,
2.  $\begin{pmatrix} I & -K(\infty) \\ -P(\infty) & I \end{pmatrix}$  is invertible,
3.  $I - K(\infty)P(\infty)$  is invertible,
4.  $I - P(\infty)K(\infty)$  is invertible.

**Proof:** Due to [Zhou, p. 14] and  $\det(I) = 1$  we have

$$\det \begin{pmatrix} I & -K \\ -P & I \end{pmatrix} = \det(I - KP) = \det(I - PK)$$

*Remark:* Very often in practical cases we have  $P(\infty) = 0$  (no direct feed-through). This gives well-posedness automatically

**Corollary 1:** Let  $K \in RH_\infty$ . Then  $(P, K)$  is internally stable iff it is well-posed and  $P(I - KP)^{-1} \in RH_\infty$ .

**Corollary 2:** Let  $P \in RH_\infty$ . Then  $(P, K)$  is internally stable iff it is well-posed and  $K(I - PK)^{-1} \in RH_\infty$ .

**Corollary 3:** Let  $P$  and  $K \in RH_\infty$ . Then  $(P, K)$  is internally stable iff it is well-posed and  $(I - PK)^{-1} \in RH_\infty$ .

See [Zhou, p.69] for proof (very easy).

## Internal Stability

Well-posedness guarantees solvability. What about stability?

**Definition:** The closed-loop system is called *internally stable* if

$$\begin{pmatrix} I & -K \\ -P & I \end{pmatrix}^{-1} \in RH_\infty$$

The  $H_\infty$ -norm of this operator is the  $L_2$ -gain from disturbances  $w$  to loop signals  $e$ . Using the formula in [Zhou, p. 14] we get the equivalent condition

$$\begin{pmatrix} (I - KP)^{-1} & K(I - PK)^{-1} \\ P(I - KP)^{-1} & (I - PK)^{-1} \end{pmatrix} \in RH_\infty.$$

## Theorem

The system is internally stable if and only if it is well-posed and

1. There is no unstable pole-zero cancellation in  $PK$ ,
2.  $(I - PK)^{-1} \in RH_\infty$ .

**Proof sketch:** If there is an unstable pole-zero cancellation in  $PK$ , then this means that there is an unstable mode of  $K$  which does not show up in the output of  $P$  and hence can not be stabilized by feedback from  $P$ .

Stability of  $PK(I - PK)^{-1} = -I + (I - PK)^{-1} = (I - PK)^{-1}PK$  and unstable pole-zero cancellation in the product  $PK$  directly implies stability of  $K(I - PK)^{-1}$  and  $(I - PK)^{-1}P$ .

To prove stability of  $(I - KP)^{-1}$  is harder.

**Definition:** Let  $m, n \in RH_\infty$ . Then  $m$  and  $n$  are said to be *coprime over  $RH_\infty$*  if there exist  $x, y \in RH_\infty$  such that  $xm + yn = 1$ .

**Definition:** Two matrices  $M, N \in RH_\infty$  are said to be

- *right coprime over  $RH_\infty$*  if there exist  $X, Y \in RH_\infty$  such that

$$\begin{pmatrix} X & Y \end{pmatrix} \begin{pmatrix} M \\ N \end{pmatrix} = XM + YN = I.$$

- *left coprime over  $RH_\infty$*  if there exist  $X, Y \in RH_\infty$  such that

$$\begin{pmatrix} M & N \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix} = MX + NY = I.$$

The right hand equations are called *Bezout identities*

## Coprime Factorization over $RH_\infty$

Let  $P$  be a proper real rational matrix. A right coprime factorization (rcf) of  $P$  is a factorization  $P = NM^{-1}$  where  $N$  and  $M$  are right coprime over  $RH_\infty$ .

Similarly, a left coprime factorization (lcf) of  $P$  has the form  $P = \tilde{M}^{-1}\tilde{N}$  and  $\tilde{N}$  and  $\tilde{M}$  are left coprime over  $RH_\infty$ . Of course,  $M$  and  $\tilde{M}$  are square.

- Coprimeness means there is no cancellation in the fraction (no nontrivial common right/left divisors).
- For scalar plant rcf=lcf.
- For real rational matrices both factorizations always exist.
- They are not unique.
- There is a state space method to calculate them.

## Feedback Interpretation

Let  $P(s) = C(sI - A)^{-1}B + D$ , that is

$$\begin{aligned} \dot{x} &= Ax + Bu, \\ y &= Cx + Du \end{aligned}$$

Introduce a change of control  $v = u - Fx$  where  $A + BF$  is stable. Then we get

$$\begin{aligned} \dot{x} &= (A + BF)x + Bv & u &= Fx + v \\ y &= (C + DF)x + Dv \end{aligned}$$

Denote by  $M(s)$  the transfer function from  $v$  to  $u$  and by  $N(s)$  the transfer function from  $v$  to  $y$

$$\begin{aligned} M(s) &= F(sI - A - BF)^{-1}B + I, \\ N(s) &= (C + DF)(sI - A - BF)^{-1}B + D. \end{aligned}$$

Therefore,  $u = Mv$ ,  $y = Nv$  and, finally,  $y = NM^{-1}u$

## Coprime Factorization and Internal Stability

Consider a plant  $P$  and a controller  $K$  with some rcf and lcf

$$P = NM^{-1} = \tilde{M}^{-1}\tilde{N} \quad K = UV^{-1} = \tilde{V}^{-1}\tilde{U}$$

**Theorem:** The following conditions are equivalent:

1. The closed-loop system  $(P, K)$  is internally stable.
2.  $\begin{pmatrix} M & U \\ N & V \end{pmatrix}$  is invertible in  $RH_\infty$ .
3.  $\begin{pmatrix} \tilde{V} & -\tilde{U} \\ -\tilde{N} & \tilde{M} \end{pmatrix}$  is invertible in  $RH_\infty$ .
4.  $\tilde{M}V - \tilde{N}U$  is invertible in  $RH_\infty$ .
5.  $\tilde{V}M - \tilde{U}N$  is invertible in  $RH_\infty$ .

*Proof:* See [Zhou, p. 74].

## Double Coprime Factorization

A double coprime factorization (dcf) of  $P$  over  $RH_\infty$  is a factorization

$$P = NM^{-1} = \tilde{M}^{-1}\tilde{N}$$

such that there exist  $X_r, X_l, Y_r, Y_l \in RH_\infty$  and it holds

$$\begin{pmatrix} X_r & Y_r \\ -\tilde{N} & \tilde{M} \end{pmatrix} \begin{pmatrix} M & -Y_l \\ N & X_l \end{pmatrix} = I.$$

- The only difference between the dcf and a couple of some rcf and lcf is in additional condition  $X_r Y_l = Y_r X_l$
- The controller  $K = -Y_l X_l^{-1} = -X_r^{-1} Y_r$  is internally stabilizing.
- There is a state space method to calculate dcf explicitly (see [Zhou]).

## Performance Specifications

Introduce the following notations

$$\begin{aligned} L_i &= KP, & L_o &= PK, \\ S_i &= (I + L_i)^{-1}, & S_o &= (I + L_o)^{-1}, \\ T_i &= I - S_i, & T_o &= I - S_o. \end{aligned}$$

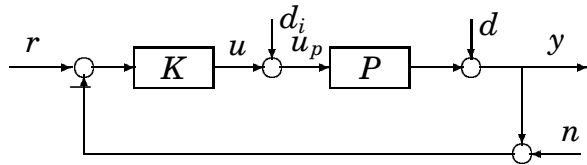
$L_i$  — the input loop transfer function,

$L_o$  — the output loop transfer function,

$S_i$  — the input sensitivity ( $u_p = S_i d_i$ ).

$S_o$  — the output sensitivity ( $y = S_o d$ ).

$T$  — the complementary sensitivity.



$$\begin{aligned} y &= T_o(r - n) + S_o P d_i + S_o d, \\ r - y &= S_o(r - d) + T_o n - S_o P d_i, \\ u &= K S_o(r - n) - K S_o d - T_i d_i, \\ u_p &= K S_o(r - n) - K S_o d + S_i d_i \end{aligned}$$

1) Good performance requires

$$\underline{\sigma}(L_o) \gg 1, \quad \underline{\sigma}(L_i) \gg 1, \quad \underline{\sigma}(K) \gg 1.$$

2) Good robustness and good sensor noise rejection requires

$$\overline{\sigma}(L_o) \ll 1, \quad \overline{\sigma}(L_i) \ll 1, \quad \overline{\sigma}(K) \leq M.$$

Conflict!!! Separate frequency bands!

## $H_2$ and $H_\infty$ Performance.

For good rejection of  $d$  at  $y$  and  $u$  both  $\|S_o\|$  and  $\|K S_o\|$  should be small at low-frequency range. It can be captured by the norm specification

$$\left\| \begin{pmatrix} W_e S_o W_d \\ \rho W_u K S_o W_d \end{pmatrix} \right\|_{2 \text{ or } \infty} \leq 1$$

where  $W_d$  reflects the frequency contents of  $d$  or models the disturbance power spectrum,  $W_e$  reflects the requirement on the shape of  $S_o$  and  $W_u$  reflects restriction on the control.

For robustness to high frequency uncertainty, the complementary sensitivity has to be limited

$$\left\| \begin{pmatrix} W_e S_o W_d \\ \rho W_u T_o W_d \end{pmatrix} \right\|_{\infty} \leq 1$$

## What have we learned today?

- Well-posedness to guarantee solvability.
- Internal stability — stability of a feedback loop
- Coprime factorization and internal stability.
- State space formula to calculate coprime factors.
- Performance specifications
- Using norms to capture loop requirements.