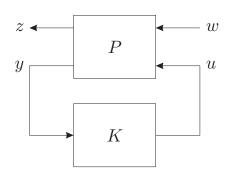
# **Optimal Output Feedback Control**

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Lecture 24: Optimal Output Feedback Control

### Recall: Linear Fractional Transformation



Plant:

$$\begin{bmatrix} z \\ y \end{bmatrix} = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \begin{bmatrix} w \\ u \end{bmatrix} \qquad \text{where} \qquad P = \begin{bmatrix} A & B_1 & B_2 \\ \hline C_1 & D_{11} & D_{12} \\ C_2 & D_{21} & D_{22} \end{bmatrix}$$

Controller:

$$u = Ky \qquad \text{where} \qquad K = \left[ \begin{array}{c|c} A_K & B_K \\ \hline C_K & D_K \end{array} \right]$$

## Optimal Control

Choose K to minimize

$$||P_{11} + P_{12}(I - KP_{22})^{-1}KP_{21}||$$

Equivalently choose  $\begin{vmatrix} A_K & B_K \\ C_K & D_K \end{vmatrix}$  to minimize

$$\left\| \begin{bmatrix} \begin{bmatrix} A & 0 \\ 0 & A_K \end{bmatrix} + \begin{bmatrix} B_2 & 0 \\ 0 & B_K \end{bmatrix} \begin{bmatrix} I & -D_K \\ -D_{22} & I \end{bmatrix}^{-1} \begin{bmatrix} 0 & C_K \\ C_2 & 0 \end{bmatrix} & B_1 + B_2 D_K Q D_{21} \\ B_K Q D_{21} \end{bmatrix} \right\|_{H^{\frac{1}{2}}}$$

$$\left[ \begin{bmatrix} C_1 & 0 \end{bmatrix} + \begin{bmatrix} D_{12} & 0 \end{bmatrix} \begin{bmatrix} I & -D_K \\ -D_{22} & I \end{bmatrix}^{-1} \begin{bmatrix} 0 & C_K \\ C_2 & 0 \end{bmatrix} & D_{11} + D_{12} D_K Q D_{21} \end{bmatrix} \right]_{H^{\frac{1}{2}}}$$

where  $Q = (I - D_{22}D_K)^{-1}$ .

# Optimal Control

Recall that

$$\begin{bmatrix} I & -D_K \\ -D_{22} & I \end{bmatrix}^{-1} = \begin{bmatrix} I + D_K Q D_{22} & D_K Q \\ Q D_{22} & Q \end{bmatrix}$$

where  $Q = (I - D_{22}D_K)^{-1}$ . Then

$$\begin{split} A_{cl} &:= \begin{bmatrix} A & 0 \\ 0 & A_K \end{bmatrix} + \begin{bmatrix} B_2 & 0 \\ 0 & B_K \end{bmatrix} \begin{bmatrix} I & -D_K \\ -D_{22} & I \end{bmatrix}^{-1} \begin{bmatrix} 0 & C_K \\ C_2 & 0 \end{bmatrix} \\ &= \begin{bmatrix} A & 0 \\ 0 & A_K \end{bmatrix} + \begin{bmatrix} B_2 & 0 \\ 0 & B_K \end{bmatrix} \begin{bmatrix} I + D_K Q D_{22} & D_K Q \\ Q D_{22} & Q \end{bmatrix} \begin{bmatrix} 0 & C_K \\ C_2 & 0 \end{bmatrix} \\ &= \begin{bmatrix} A + B_2 D_K Q C_2 & B_2 (I + D_K Q D_{22}) C_K \\ B_K Q C_2 & A_K + B_K Q D_{22} C_K \end{bmatrix} \end{split}$$

Likewise

$$\begin{split} C_{cl} &:= \begin{bmatrix} C_1 & 0 \end{bmatrix} + \begin{bmatrix} D_{12} & 0 \end{bmatrix} \begin{bmatrix} I + D_K Q D_{22} & D_K Q \\ Q D_{22} & Q \end{bmatrix} \begin{bmatrix} 0 & C_K \\ C_2 & 0 \end{bmatrix} \\ &= \begin{bmatrix} C_1 + D_{12} D_K Q C_2 & D_{12} (I + D_K Q D_{22}) C_K \end{bmatrix} \end{split}$$

Thus we have

$$\begin{bmatrix} \begin{bmatrix} A + B_2 D_K Q C_2 & B_2 (I + D_K Q D_{22}) C_K \\ B_K Q C_2 & A_K + B_K Q D_{22} C_K \end{bmatrix} & B_1 + B_2 D_K Q D_{21} \\ \hline \begin{bmatrix} C_1 + D_{12} D_K Q C_2 & D_{12} (I + D_K Q D_{22}) C_K \end{bmatrix} & D_{11} + D_{12} D_K Q D_{21} \end{bmatrix}$$

where  $Q = (I - D_{22}D_K)^{-1}$ .

- This is nonlinear in  $(A_K, B_K, C_K, D_K)$ .
- Hence we make a change of variables (First of several).

$$A_{K2} = A_K + B_K Q D_{22} C_K$$

$$B_{K2} = B_K Q$$

$$C_{K2} = (I + D_K Q D_{22}) C_K$$

$$D_{K2} = D_K Q$$

This yields the system

$$\begin{bmatrix} \begin{bmatrix} A + B_2 D_{K2} C_2 & B_2 C_{K2} \\ B_{K2} C_2 & A_{K2} \end{bmatrix} & B_1 + B_2 D_{K2} D_{21} \\ B_{K2} D_{21} & B_{K2} D_{21} \end{bmatrix}$$
$$\begin{bmatrix} C_1 + D_{12} D_{K2} C_2 & D_{12} C_{K2} \end{bmatrix} & D_{11} + D_{12} D_{K2} D_{21} \end{bmatrix}$$

Which is affine in  $\begin{bmatrix} A_{K2} & B_{K2} \\ \hline C_{K2} & D_{K2} \end{bmatrix}$ .

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Hence we can optimize over our new variables.

However, the change of variables must be invertible.

If we recall that

$$(I - QM)^{-1} = I + Q(I - MQ)^{-1}M$$

then we get

$$I + D_K Q D_{22} = I + D_K (I - D_{22} D_K)^{-1} D_{22} = (I - D_K D_{22})^{-1}$$

Examine the variable  $C_{K2}$ 

$$C_{K2} = (I + D_K (I - D_{22} D_K)^{-1} D_{22}) C_K$$
$$= (I - D_K D_{22})^{-1} C_K$$

Hence, given  $C_{K2}$ , we can recover  $C_K$  as

$$C_K = (I - D_K D_{22}) C_{K2}$$

Now suppose we have  $D_{K2}$ . Then

$$D_{K2} = D_K Q = D_K (I - D_{22} D_K)^{-1}$$

implies that

$$D_K = D_{K2}(I - D_{22}D_K) = D_{K2} - D_{K2}D_{22}D_K$$

or

$$(I + D_{K2}D_{22})D_K = D_{K2}$$

which can be inverted to get

$$D_K = (I + D_{K2}D_{22})^{-1}D_{K2}$$

Once we have  $C_K$  and  $D_K$ , the other variables are easily recovered as

$$B_K = B_{K2}Q^{-1} = B_{K2}(I - D_{22}D_K)$$
  

$$A_K = A_{K2} - B_K(I - D_{22}D_K)^{-1}D_{22}C_K$$

To summarize, the original variables can be recovered as

$$D_K = (I + D_{K2}D_{22})^{-1}D_{K2}$$

$$B_K = B_{K2}(I - D_{22}D_K)$$

$$C_K = (I - D_KD_{22})C_{K2}$$

$$A_K = A_{K2} - B_K(I - D_{22}D_K)^{-1}D_{22}C_K$$

$$\left[ \begin{array}{c|c} A_{cl} & B_{cl} \\ \hline C_{cl} & D_{cl} \end{array} \right] := \left[ \begin{array}{c|c} \left[ A + B_2 D_{K2} C_2 & B_2 C_{K2} \\ B_{K2} C_2 & A_{K2} \end{array} \right] & \begin{array}{c|c} B_1 + B_2 D_{K2} D_{21} \\ B_{K2} D_{21} \\ \hline \left[ C_1 + D_{12} D_{K2} C_2 & D_{12} C_{K2} \right] & D_{11} + D_{12} D_{K2} D_{21} \end{array} \right]$$

$$\begin{bmatrix} A_{cl} & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix} = \begin{bmatrix} A & 0 & B_1 \\ 0 & 0 & 0 \\ C_1 & 0 & D_{11} \end{bmatrix} + \begin{bmatrix} 0 & B_2 \\ I & 0 \\ 0 & D_{12} \end{bmatrix} \begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix} \begin{bmatrix} 0 & I & 0 \\ C_2 & 0 & D_{21} \end{bmatrix}$$

Or

$$\begin{split} A_{cl} &= \begin{bmatrix} A & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & B_2 \\ I & 0 \end{bmatrix} \begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix} \begin{bmatrix} 0 & I \\ C_2 & 0 \end{bmatrix} \\ B_{cl} &= \begin{bmatrix} B_1 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 & B_2 \\ I & 0 \end{bmatrix} \begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix} \begin{bmatrix} 0 \\ D_{21} \end{bmatrix} \\ C_{cl} &= \begin{bmatrix} C_1 & 0 \end{bmatrix} + \begin{bmatrix} 0 & D_{12} \end{bmatrix} \begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix} \begin{bmatrix} 0 & I \\ C_2 & 0 \end{bmatrix} \\ D_{cl} &= \begin{bmatrix} D_{11} \end{bmatrix} + \begin{bmatrix} 0 & D_{12} \end{bmatrix} \begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix} \begin{bmatrix} 0 \\ D_{21} \end{bmatrix} \end{split}$$

### Lemma 1 (Transformation Lemma).

Suppose that

$$\begin{bmatrix} Y_1 & I \\ I & X_1 \end{bmatrix} > 0$$

Then there exist  $X_2, X_3, Y_2, Y_3$  such that

$$X_{cl} = \begin{bmatrix} X_1 & X_2 \\ X_2^T & X_3 \end{bmatrix} = \begin{bmatrix} Y_1 & Y_2 \\ Y_2^T & Y_3 \end{bmatrix}^{-1} > 0$$

where  $\begin{vmatrix} Y & I \\ Y_2^T & 0 \end{vmatrix}$  has full column rank.

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### Proof.

Since

$$\begin{bmatrix} Y_1 & I \\ I & X_1 \end{bmatrix} > 0,$$

this matrix must be invertible and hence I - XY is invertible.

ullet Choose any two square invertible matrices  $X_2$  and  $Y_2$  such that

$$X_2 Y_2^T = I - X_1 Y_1$$

ullet Because  $Y_2$  is non-singular,

$$Y_{cl} = \begin{bmatrix} Y & I \\ Y_2^T & 0 \end{bmatrix}$$

is also non-singular

Now define

$$X_{cl} = \begin{bmatrix} Y_1 & Y_2 \\ I & 0 \end{bmatrix}^{-1} \begin{bmatrix} I & 0 \\ X_1 & X_2 \end{bmatrix}$$

• Then XY = I

### Lemma 2 (Converse Transformation Lemma).

Given 
$$X_{cl} = \begin{bmatrix} X_1 & X_2 \\ X_2^T & X_3 \end{bmatrix} > 0$$
 where  $X_2$  has full column rank. Let

$$X_{cl}^{-1} = Y = \begin{bmatrix} Y_1 & Y_2 \\ Y_2^T & Y_3 \end{bmatrix}$$

then

$$\begin{bmatrix} Y_1 & I \\ I & X_1 \end{bmatrix} > 0$$

and  $Y_{cl} = \begin{bmatrix} Y_1 & I \\ Y_2^T & 0 \end{bmatrix}$  has full column rank.

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#### Proof.

Since  $X_2$  is full row rank,  $\begin{bmatrix} I & 0 \\ X_1 & X_2 \end{bmatrix}$  also has full row rank. Note that

$$\begin{bmatrix} Y_1 & Y_2 \\ I & 0 \end{bmatrix} \begin{bmatrix} X_1 & X_2 \\ X_2^T & X_3 \end{bmatrix} = \begin{bmatrix} I & 0 \\ X_1 & X_2 \end{bmatrix}.$$

Hence

$$Y_{cl} = \begin{bmatrix} Y_1 & I \\ Y_2^T & 0 \end{bmatrix} = \begin{bmatrix} I & X_1^T \\ 0 & X_2^T \end{bmatrix} Y$$

has full column rank. Furthermore, because  $Y_{cl}$  has full column rank,

$$\begin{bmatrix} Y_1 & I \\ I & X_1 \end{bmatrix} = Y_{cl}^T X_{xl} Y_{cl} > 0$$

### Theorem 3.

The following are equivalent.

• There exists a 
$$\hat{K}=\left[\begin{array}{c|c}A_K & B_K\\\hline C_K & D_K\end{array}\right]$$
 such that  $\|S(K,P)\|_{H_\infty}<\gamma.$ 

• There exist 
$$X_1,Y_1,A_n,B_n,C_n,D_n$$
 such that  $\begin{bmatrix} X_1 & I \\ I & Y_1 \end{bmatrix}>0$ 

$$\begin{bmatrix} AY_1 + Y_1A & *^T & *^T & *^T \\ A^T + A_n + [B_2D_nC_2]^T & X_1A + A^TX_1 & *^T & *^T \\ B_1^T + [B_2D_nD_{21}]^T & B_1^TX_1 + D_{21}B_n & -\gamma I & *^T \\ C_1Y_1 + D_{12}C_n & C_1 + D_{12}D_nC_2 & D_{11} + D_{12}D_nD_{21} & -\gamma I \end{bmatrix} < 0$$

Then

$$\begin{bmatrix} A_{K2} & B_{K2} \\ \hline C_{K2} & D_{K2} \end{bmatrix} = \begin{bmatrix} X_2 & X_1B_2 \\ 0 & I \end{bmatrix}^{-1} \begin{bmatrix} \begin{bmatrix} A_n & B_n \\ C_n & D_n \end{bmatrix} - \begin{bmatrix} X_1AY_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} Y_2^T & 0 \\ C_2Y_1 & I \end{bmatrix}^{-1}$$

for any full-rank  $X_2$  and  $Y_2$  such that

$$\begin{bmatrix} X_1 & X_2 \\ X_2^T & X_3 \end{bmatrix} = \begin{bmatrix} Y_1 & Y_2 \\ Y_2^T & Y_3 \end{bmatrix}^{-1}$$

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#### Proof: If.

Suppose there exist  $X_1, Y_1, A_n, B_n, C_n, D_n$  such that the LMI is feasible. Since

$$\begin{bmatrix} X_1 & I \\ I & Y_1 \end{bmatrix} > 0,$$

by the transformation lemma, there exist  $X_2, X_3, Y_2, Y_3$  such that

$$X_{cl} := \begin{bmatrix} X & X_2 \\ X_2^T & X_3 \end{bmatrix} = \begin{bmatrix} Y & Y_2 \\ Y_2^T & Y_3 \end{bmatrix}^{-1} > 0$$

where 
$$Y_{cl} = \begin{bmatrix} Y & I \\ Y_2^T & 0 \end{bmatrix}$$
 has full row rank. Let  $K = \begin{bmatrix} A_K & B_K \\ \hline C_K & D_K \end{bmatrix}$  where

$$D_K = (I + D_{K2}D_{22})^{-1}D_{K2}$$

$$B_K = B_{K2}(I - D_{22}D_K)$$

$$C_K = (I - D_KD_{22})C_{K2}$$

$$A_K = A_{K2} - B_K(I - D_{22}D_K)^{-1}D_{22}C_K.$$

### Proof: If.

and where

$$\begin{bmatrix} A_{K2} & B_{K2} \\ \hline C_{K2} & D_{K2} \end{bmatrix} = \begin{bmatrix} X_2 & X_1 B_2 \\ 0 & I \end{bmatrix}^{-1} \begin{bmatrix} \begin{bmatrix} A_n & B_n \\ C_n & D_n \end{bmatrix} - \begin{bmatrix} X_1 A Y_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} Y_2^T & 0 \\ C_2 Y_1 & I \end{bmatrix}^{-1}.$$

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#### Proof: If.

As discussed previously, this means the closed-loop system is

$$\begin{split} \begin{bmatrix} A_{cl} & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix} &= \begin{bmatrix} A & 0 & B_1 \\ 0 & 0 & 0 \\ C_1 & 0 & D_{11} \end{bmatrix} + \begin{bmatrix} 0 & B_2 \\ I & 0 \\ 0 & D_{12} \end{bmatrix} \begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix} \begin{bmatrix} 0 & I & 0 \\ C_2 & 0 & D_{21} \end{bmatrix} \\ &= \begin{bmatrix} A & 0 & B_1 \\ 0 & 0 & 0 \\ C_1 & 0 & D_{11} \end{bmatrix} + \begin{bmatrix} 0 & B_2 \\ I & 0 \\ 0 & D_{12} \end{bmatrix} \\ \begin{bmatrix} X_2 & X_1 B_2 \\ 0 & I \end{bmatrix}^{-1} \begin{bmatrix} \begin{bmatrix} A_n & B_n \\ C_n & D_n \end{bmatrix} - \begin{bmatrix} X_1 A Y_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} Y_2^T & 0 \\ C_2 Y_1 & I \end{bmatrix}^{-1} \begin{bmatrix} 0 & I & 0 \\ C_2 & 0 & D_{21} \end{bmatrix} \end{split}$$

We now apply the KYP lemma to this system

#### Proof: If.

Expanding out, we obtain

$$\begin{bmatrix} Y_{cl}^T & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix} \begin{bmatrix} A_{cl}^T X_{cl} + X_{cl} A_{cl} & X_{cl} & C_{cl}^T \\ B_{cl}^T X_{cl} & -\gamma I & D_{cl}^T \\ C_{cl} & D_{cl} & -\gamma I \end{bmatrix} \begin{bmatrix} Y_{cl} & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix}$$
 
$$= \begin{bmatrix} AY_1 + Y_1 A & *^T & *^T & *^T \\ A^T + A_n + [B_2 D_n C_2]^T & X_1 A + A^T X_1 & *^T & *^T \\ B_1^T + [B_2 D_n D_{21}]^T & B_1^T X_1 + D_{21} B_n & -\gamma I \\ C_1 Y_1 + D_{12} C_n & C_1 + D_{12} D_n C_2 & D_{11} + D_{12} D_n D_{21} & -\gamma I \end{bmatrix} < 0$$

Hence, by the KYP lemma, 
$$\underline{S}(P,K) = \begin{bmatrix} A_{cl} & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix}$$
 satisfies  $\|\underline{S}(P,K)\|_{H_{\infty}} < \gamma$ .

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### Proof: Only If.

Now suppose that  $\|\underline{\mathbf{S}}(P,K)\|_{H_\infty} < \gamma$  for some  $K = \left\lfloor \frac{A_K \mid B_K}{C_K \mid D_K} \right\rfloor$ . Since

 $\|\underline{\mathbf{S}}(P,K)\|_{H_{\infty}}<\gamma,$  by the KYP lemma, there exists a

$$X_{cl} = \begin{bmatrix} X_1 & X_2 \\ X_2^T & X_3 \end{bmatrix} > 0$$

such that

$$\begin{bmatrix} A_{cl}^{T} X_{cl} + X_{cl} A_{cl} & X_{cl} & C_{cl}^{T} \\ B_{cl}^{T} X_{cl} & -\gamma I & D_{cl}^{T} \\ C_{cl} & D_{cl} & -\gamma I \end{bmatrix} < 0$$

Because the inequalities are strict, we can assume that  $X_2$  has full row rank.

Define

$$Y = egin{bmatrix} Y_1 & Y_2 \ Y_2^T & Y_3 \end{bmatrix} = X_{cl}^{-1} \qquad ext{ and } \qquad Y_{cl} = egin{bmatrix} Y_1 & I \ Y_2^T & 0 \end{bmatrix}$$

Then, according to the converse transformation lemma,  $Y_{cl}$  has full row rank and  $\begin{bmatrix} X_1 & I \\ I & Y_1 \end{bmatrix} > 0.$ 

### **Proof: Only If.**

Now, using the given  $A_K, B_K, C_K, D_K$ , define the variables

$$\begin{bmatrix} A_n & B_n \\ C_n & D_n \end{bmatrix} = \begin{bmatrix} X_2 & X_1B_2 \\ 0 & I \end{bmatrix} \begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix} \begin{bmatrix} Y_2^T & 0 \\ C_2Y_1 & I \end{bmatrix} + \begin{bmatrix} X_1AY_1 & 0 \\ 0 & 0 \end{bmatrix}.$$

where

$$A_{K2} = A_K + B_K (I - D_{22}D_K)^{-1} D_{22}C_K$$
  $B_{K2} = B_K (I - D_{22}D_K)^{-1}$   
 $C_{K2} = (I + D_K (I - D_{22}D_K)^{-1} D_{22})C_K$   $D_{K2} = D_K (I - D_{22}D_K)^{-1}$ 

Then as before

$$\begin{bmatrix} A_{cl} & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix} = \begin{bmatrix} A & 0 & B_1 \\ 0 & 0 & 0 \\ C_1 & 0 & D_{11} \end{bmatrix} + \begin{bmatrix} 0 & B_2 \\ I & 0 \\ 0 & D_{12} \end{bmatrix}$$

$$\begin{bmatrix} X_2 & X_1B_2 \\ 0 & I \end{bmatrix}^{-1} \begin{bmatrix} \begin{bmatrix} A_n & B_n \\ C_n & D_n \end{bmatrix} - \begin{bmatrix} X_1AY_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} Y_2^T & 0 \\ C_2Y_1 & I \end{bmatrix}^{-1} \begin{bmatrix} 0 & I & 0 \\ C_2 & 0 & D_{21} \end{bmatrix}$$

### Proof: Only If.

Expanding out the LMI, we find

$$\begin{bmatrix} AY_1 + Y_1A & *^T & *^T & *^T \\ A^T + A_n + [B_2D_nC_2]^T & X_1A + A^TX_1 & *^T & *^T \\ B_1^T + [B_2D_nD_{21}]^T & B_1^TX_1 + D_{21}B_n & -\gamma I \\ C_1Y_1 + D_{12}C_n & C_1 + D_{12}D_nC_2 & D_{11} + D_{12}D_nD_{21} & -\gamma I \end{bmatrix}$$

$$= \begin{bmatrix} Y_{cl}^T & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix} \begin{bmatrix} A_{cl}^TX_{cl} + X_{cl}A_{cl} & X_{cl} & C_{cl}^T \\ B_{cl}^TX_{cl} & -\gamma I & D_{cl}^T \\ C_{cl} & D_{cl} & -\gamma I \end{bmatrix} \begin{bmatrix} Y_{cl} & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix} < 0$$

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### Conclusion

To solve the  $H_{\infty}$ -optimal state-feedback problem, we solve

$$\begin{split} & \min_{\gamma, X_1, Y_1, A_n, B_n, C_n, D_n} \gamma & \text{ such that } \\ & \begin{bmatrix} X_1 & I \\ I & Y_1 \end{bmatrix} > 0 \\ & \begin{bmatrix} AY_1 + Y_1 A & *^T & *^T & *^T \\ A^T + A_n + [B_2 D_n C_2]^T & X_1 A + A^T X_1 & *^T & *^T \\ B_1^T + [B_2 D_n D_{21}]^T & B_1^T X_1 + D_{21} B_n & -\gamma I & *^T \\ C_1 Y_1 + D_{12} C_n & C_1 + D_{12} D_n C_2 & D_{11} + D_{12} D_n D_{21} & -\gamma I \end{bmatrix} < 0 \end{split}$$

### Conclusion

Then, we construct our controller using

$$\begin{split} D_K &= (I + D_{K2}D_{22})^{-1}D_{K2} \\ B_K &= B_{K2}(I - D_{22}D_K) \\ C_K &= (I - D_KD_{22})C_{K2} \\ A_K &= A_{K2} - B_K(I - D_{22}D_K)^{-1}D_{22}C_K. \end{split}$$

where

$$\begin{bmatrix} A_{K2} & B_{K2} \\ \hline C_{K2} & D_{K2} \end{bmatrix} = \begin{bmatrix} X_2 & X_1 B_2 \\ 0 & I \end{bmatrix}^{-1} \begin{bmatrix} \begin{bmatrix} A_n & B_n \\ C_n & D_n \end{bmatrix} - \begin{bmatrix} X_1 A Y_1 & 0 \\ 0 & 0 \end{bmatrix} \end{bmatrix} \begin{bmatrix} Y_2^T & 0 \\ C_2 Y_1 & I \end{bmatrix}^{-1}.$$

and where  $X_2$  and  $Y_2$  are any matrices which satisfy  $X_2Y_2 = I - X_1Y_1$ .

- e.g. Let  $Y_2 = I$  and  $X_2 = I X_1 Y_1$ .
- The optimal controller is NOT uniquely defined.
- Don't forget to check invertibility of  $I-D_{22}D_K$

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### Conclusion

The  $H_{\infty}$ -optimal controller is a dynamic system.

• Transfer Function  $\hat{K}(s) = \begin{bmatrix} A_K & B_K \\ \hline C_K & D_K \end{bmatrix}$ 

Minimizes the effect of external input (w) on external output (z).

$$\|z\|_{L_2} \leq \|\underline{\mathsf{S}}(P,K)\|_{H_\infty} \|w\|_{L_2}$$

Minimum Energy Gain

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### Extensions

#### $H_2$ -optimal control

 $H_2$ -optimal control minimizes the  $H_2$ -norm of the transfer function.

• The  $H_2$ -norm has no direct interpretation.

$$\|G\|_{H_2}^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} \mathrm{Trace}(\hat{G}^*(\imath \omega) \hat{G}(\imath \omega)) d\omega$$

Motivation: Assume external input is Gaussian noise with signal variance (Power)

$$E[w(t)^2] = \frac{1}{2\pi} \int_{-}^{\infty} \infty^{\infty} \operatorname{Trace}(S(\imath \omega)) d\omega$$

Then the output has signal variance (Power)

$$\begin{split} E[z(t)^2] &= \frac{1}{2\pi} \int_{-}^{\infty} \infty^{\infty} \mathrm{Trace}(\hat{G}^*(\imath \omega) S(\imath \omega) \hat{G}(\imath \omega)) d\omega \\ &= \|S\|_{H_{\infty}} \|G\|_{H_{2}}^{2} \end{split}$$

If the input signal is white noise, then  $S(\imath\omega)=I$  and

$$E[z(t)^2] = ||G||_{H_2}^2$$

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### Extensions

#### $H_2$ -optimal control

Thus minimizing the  $H_2$ -norm minimizes the effect of white noise on the power of the output noise.

• This is why  $H_2$  control is often called Least-Quadratic-Gaussian (LQG).

### LQR:

- Full-State Feedback
- Choose K to minimize the cost function

$$\int_0^\infty x(t)^T Qx(t) + u(t)^T Ru(t) dt$$

subject to dynamic constraints

$$\dot{x}(t) = Ax(t) + Bu(t)$$
$$u(t) = Kx(t)$$

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