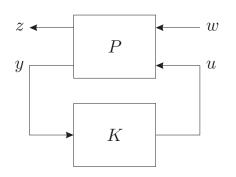
Modern Optimal Control

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Lecture 21: Optimal Output Feedback Control

Optimal Output Feedback

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:

$$u = Ky$$
 where $K = \begin{bmatrix} A_K & B_K \\ \hline C_K & D_K \end{bmatrix}$

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 the Matrix Inversion Lemma:

Lemma 1.

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}^{-1}$$

$$= \begin{bmatrix} A^{-1} + A^{-1}B(D - CA^{-1}B)^{-1}CA^{-1} & -A^{-1}B(D - CA^{-1}B)^{-1} \\ -(D - CA^{-1}B)^{-1}CA^{-1} & (D - CA^{-1}B)^{-1} \end{bmatrix}$$

$$= \begin{bmatrix} (A - BD^{-1}C)^{-1} & -(A - BD^{-1}C)^{-1}BD^{-1} \\ -D^{-1}C(A - BD^{-1}C)^{-1} & D^{-1} + D^{-1}C(A - BD^{-1}C)^{-1}BD^{-1} \end{bmatrix}$$

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

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

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 $\begin{bmatrix} A_{K2} & B_{K2} \\ C_{K2} & D_{K2} \end{bmatrix}$.

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

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

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

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

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

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

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

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 2 (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

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

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

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Proof.

Since

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

by the Schur complement $X_1 > 0$ and $X_1^{-1} - Y_1 > 0$. Since $I - X_1Y_1 = X_1(X_1^{-1} - Y_1)$, we conclude that $I - X_1Y_1$ is invertible.

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

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

• Because Y_2 is non-singular,

$$egin{align*} Y_{cl}^T = egin{bmatrix} Y_1 & Y_2 \ I & 0 \end{bmatrix} ext{ and } X_{cl} = egin{bmatrix} I & 0 \ X_1 & X_2 \end{bmatrix} \end{split}$$

are also non-singular.

Proof.

ullet Now define X and Y as

$$X = Y_{cl}^{-T} X_{cl} \qquad \text{ and } \qquad Y = X_{cl}^{-1} Y_{cl}^T.$$

Then

$$XY = Y_{cl}^{-1} X_{cl} X_{cl}^{-1} Y_{cl} = I$$



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Lemma 3 (Converse Transformation Lemma).

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

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

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 rank, $X_{cl}=\begin{bmatrix}I&0\\X_1&X_2\end{bmatrix}$ also has full column rank. Note that XY=I implies

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

Hence

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

has full column rank. Now, since XY = I implies $X_1Y_1 + X_2Y_2^T = I$, we have

$$X_{cl}Y_{cl} = \begin{bmatrix} I & 0 \\ X_1 & X_2 \end{bmatrix} \begin{bmatrix} Y_1 & I \\ Y_2^T & 0 \end{bmatrix} = \begin{bmatrix} Y_1 & I \\ X_1Y_1 + X_2Y_2^T & X_1 \end{bmatrix} = \begin{bmatrix} Y_1 & I \\ I & X_1 \end{bmatrix}$$

Furthermore, because Y_{cl} has full rank,

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

Theorem 4.

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

$$\begin{bmatrix} AY_1 + Y_1A^T + B_2C_n + C_n^TB_2^T & *^T & *^T & *^T \\ A^T + A_n + [B_2D_nC_2]^T & X_1A + A^TX_1 + B_nC_2 + C_2^TB_n^T & *^T & *^T \\ [B_1 + B_2D_nD_{21}]^T & [XB_1 + B_nD_{21}]^T & -\gamma I \\ C_1Y_1 + D_{12}C_n & C_1 + D_{12}D_nC_2 & D_{11} + D_{12}D_nD_{21} & -\gamma I \end{bmatrix} < 0$$

Moreover,

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

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

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

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

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

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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

$$\begin{bmatrix} Y_{cl}^T & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix} \begin{bmatrix} A_{cl}^TX + XA_{cl} & XB_{cl} & C_{cl}^T \\ B_{cl}^TX & -\gamma I & D_{cl}^T \\ C_{cl} & D_{cl} & -\gamma I \end{bmatrix} \begin{bmatrix} Y_{cl} & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix} = \begin{bmatrix} AY_1 + Y_1A^T + B_2C_n + C_n^TB_2^T & *^T & *^T & *^T \\ A^T + A_n + [B_2D_nC_2]^T & X_1A + A^TX_1 + B_nC_2 + C_2^TB_n^T & *^T & *^T \\ [B_1 + B_2D_nD_{21}]^T & [XB_1 + B_nD_{21}]^T & -\gamma I \\ C_1Y_1 + D_{12}C_n & C_1 + D_{12}D_nC_2 & D_{11} + D_{12}D_nD_{21} & -\gamma I \end{bmatrix} < 0$$

Hence, by the KYP lemma,
$$\underline{S}(P,K)=\begin{bmatrix} A_{cl} & B_{cl} \\ \hline C_{cl} & D_{cl} \end{bmatrix}$$
 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

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

such that

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

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

Define

$$Y = \begin{bmatrix} Y_1 & Y_2 \\ Y_2^T & Y_3 \end{bmatrix} = X^{-1} \qquad \text{ and } \qquad Y_{cl} = \begin{bmatrix} Y_1 & I \\ Y_2^T & 0 \end{bmatrix}$$

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

$$\begin{bmatrix} AY_1 + Y_1A^T + B_2C_n + C_n^TB_2^T & *^T & *^T & *^T \\ A^T + A_n + [B_2D_nC_2]^T & X_1A + A^TX_1 + B_nC_2 + C_2^TB_n^T & *^T & *^T \\ [B_1 + B_2D_nD_{21}]^T & [XB_1 + B_nD_{21}]^T & -\gamma I \\ C_1Y_1 + D_{12}C_n & C_1 + D_{12}D_nC_2 & D_{11} + D_{12}D_nD_{21} & -\gamma I \end{bmatrix}$$

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

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Conclusion

To solve the H_{∞} -optimal state-feedback problem, we solve

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

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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_{∞} -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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Alternative Formulation

Theorem 5.

Let N_o and N_c be full-rank matrices whose images satisfy

$$\begin{array}{ll} \textit{Im } N_o = \textit{Ker } \begin{bmatrix} C_2 & D_{21} \end{bmatrix} \\ \textit{Im } N_c = \textit{Ker } \begin{bmatrix} B_2^T & D_{12}^T \end{bmatrix} \end{array}$$

Then 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 $\|S(K,P)\|_{H_\infty}<1$.
- There exist X_1,Y_1 such that $\begin{bmatrix} X_1 & I \\ I & Y_1 \end{bmatrix} > 0$ and

$$\begin{bmatrix} N_o & 0 \\ 0 & I \end{bmatrix}^T \begin{bmatrix} X_1 A + A^T X_1 & *^T & *^T \\ B_1^T X_1 & -I & *^T \\ C_1 & D_{11} & -I \end{bmatrix} \begin{bmatrix} N_o & 0 \\ 0 & I \end{bmatrix} < 0$$

$$\begin{bmatrix} N_c & 0 \\ 0 & I \end{bmatrix}^T \begin{bmatrix} AY_1 + Y_1 A & *^I & *^I \\ C_1 Y_1 & -I & *^T \\ B_1^T & D_{11}^T & -I \end{bmatrix} \begin{bmatrix} N_c & 0 \\ 0 & I \end{bmatrix} < 0$$