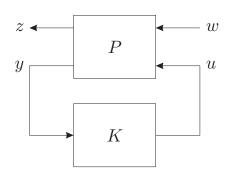
Modern Optimal Control

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Lecture 20: Optimal Full-State 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:

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

M. Peet Lecture 20: 2 / 13

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

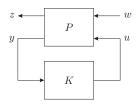
$$\left\| \begin{bmatrix} \begin{bmatrix} A & 0 \\ 0 & A_K \end{bmatrix} + \begin{bmatrix} B_2 & 0 \\ 0 & B_K \end{bmatrix} \begin{bmatrix} I & -D_K \\ -D_{22} & I \end{bmatrix}^{-1} \begin{bmatrix} 0 & C_K \\ C_2 & 0 \end{bmatrix} & B_1 + B_2 D_K Q D_{21} \\ B_K Q D_{21} \end{bmatrix} \right\|_{H^{\frac{1}{2}}}$$

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

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

3 / 13 M. Peet Lecture 20:

Optimal Full-State Feedback Control



For the full-state feedback case, we consider a controller of the form

$$u(t) = Fx(t)$$

Controller:

$$u = Ky$$
 where

$$u = Ky$$
 where $K = \begin{bmatrix} 0 & 0 \\ \hline 0 & F \end{bmatrix}$

Plant:

$$\begin{bmatrix} z \\ y \end{bmatrix} = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \begin{bmatrix} w \\ u \end{bmatrix} \qquad \text{where} \qquad P = \begin{bmatrix} A & B_1 & B_2 \\ \hline C_1 & D_{11} & D_{12} \\ I & 0 & 0 \end{bmatrix}$$

M. Peet 4 / 13 Lecture 20:

Optimal Full-State Feedback Control

Thus the closed-loop state-space representation is

$$\underline{S}(\hat{P}, \hat{K}) = \begin{bmatrix} A + B_2 F & B_1 \\ \hline C_1 + D_{12} F & D_{11} \end{bmatrix}$$

By the KYP lemma, $\|\underline{\mathbf{S}}(\hat{P},\hat{K})\|_{H_\infty}<\gamma$ if and only if there exists some X>0 such that

$$\begin{bmatrix} (A + B_2 F)^T X + X(A + B_2 F) & X B_1 \\ B_1^T X & -\gamma I \end{bmatrix}
+ \frac{1}{\gamma} \begin{bmatrix} (C_1 + D_{12} F)^T \\ D_{11}^T \end{bmatrix} [(C_1 + D_{12} F) & D_{11} \end{bmatrix} < 0$$

This is a matrix inequality, but is nonlinear

- Quadratic (Not Bilinear)
- May NOT apply variable substitution trick.

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Schur Complement

The KYP condition is

$$\begin{bmatrix} A^TX + XA & XB \\ B^TX & -\gamma I \end{bmatrix} + \frac{1}{\gamma} \begin{bmatrix} C^T \\ D^T \end{bmatrix} \begin{bmatrix} C & D \end{bmatrix} < 0$$

Recall the Schur Complement

Theorem 1 (Schur Complement).

For any $S \in \mathbb{S}^n$, $Q \in \mathbb{S}^m$ and $R \in \mathbb{R}^{n \times m}$, the following are equivalent.

$$\begin{array}{ccc}
\mathbf{1}. & \begin{bmatrix} M & R \\ R^T & Q \end{bmatrix} < 0
\end{array}$$

2.
$$Q < 0$$
 and $M - RQ^{-1}R^T < 0$

In this case, let $Q=-\frac{1}{\gamma}I<0$,

$$M = \begin{bmatrix} A^T X + XA & XB \\ B^T X & -\gamma I \end{bmatrix} \qquad R = \begin{bmatrix} C & D \end{bmatrix}^T$$

Note we are making the LMI Larger.

M. Peet Lecture 20: 6 / 13

Schur Complement

The Schur Complement says that

$$\begin{bmatrix} A^TX + XA & XB \\ B^TX & -\gamma I \end{bmatrix} + \frac{1}{\gamma} \begin{bmatrix} C^T \\ D^T \end{bmatrix} \begin{bmatrix} C & D \end{bmatrix} < 0$$

if and only if

$$\begin{bmatrix} A^TX + XA & XB & C^T \\ B^TX & -\gamma I & D^T \\ C & D & -\gamma I \end{bmatrix} < 0$$

This leads to the

Full-State Feedback Condition

$$\begin{bmatrix} (A+B_2F)^TX + X(A+B_2F) & XB_1 & (C_1+D_{12}F)^T \\ B_1^TX & -\gamma I & D_{11}^T \\ (C_1+D_{12}F) & D_{11} & -\gamma I \end{bmatrix} < 0$$

which is now bilinear in X and F.

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Dual KYP Lemma

To apply the variable substitution trick, we must also construct the dual form of this LMI.

Lemma 2 (KYP Dual).

Suppose

$$\hat{G}(s) = \left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right].$$

Then the following are equivalent.

- $||G||_{H_{\infty}} \leq \gamma$.
- There exists a Y > 0 such that

$$\begin{bmatrix} YA^T + AY & B & YC^T \\ B^T & -\gamma I & D^T \\ CY & D & -\gamma I \end{bmatrix} < 0$$

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Dual KYP Lemma

Proof.

Let
$$X=Y^{-1}$$
. Then
$$\begin{bmatrix} YA^T+AY & B & YC^T \\ B^T & -\gamma I & D^T \\ CY & D & -\gamma I \end{bmatrix} < 0 \qquad \text{and} \qquad Y>0$$

if and only if ${\cal X}>0$ and

$$\begin{bmatrix} Y^{-1} & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix} \begin{bmatrix} YA^T + AY & B & YC^T \\ B^T & -\gamma I & D^T \\ CY & D & -\gamma I \end{bmatrix} \begin{bmatrix} Y^{-1} & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & I \end{bmatrix}$$
$$= \begin{bmatrix} A^TX + XA & XB & C^T \\ B^TX & -\gamma I & D^T \\ C & D & -\gamma I \end{bmatrix} < 0.$$

By the Schur complement this is equivalent to

$$\begin{bmatrix} A^TX + XA & XB \\ B^TX & -\gamma I \end{bmatrix} + \frac{1}{\gamma} \begin{bmatrix} C^T \\ D^T \end{bmatrix} \begin{bmatrix} C & D \end{bmatrix} < 0$$

By the KYP lemma, this equivalent to $||G||_{H_{\infty}} \leq \gamma$.

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We can now apply this result to the state-feedback problem.

Theorem 3.

The following are equivalent:

- There exists an F such that $\|\underline{S}(P, K(0, 0, 0, F))\|_{H_{\infty}} \leq \gamma$.
- There exist Y > 0 and Z such that

$$\begin{bmatrix} YA^T + AY + Z^TB_2^T + B_2Z & B_1 & YC_1^T + Z^TD_{12}^T \\ B_1^T & -\gamma I & D_{11}^T \\ C_1Y + D_{12}Z & D_{11} & -\gamma I \end{bmatrix} < 0$$

Then $F = ZY^{-1}$.

M. Peet Lecture 20: 10 / 13

Proof.

Suppose there exists an F such that $\|\underline{S}(P,K(0,0,0,F))\|_{H_\infty} \leq \gamma$. By the Dual KYP lemma, this implies there exists a Y>0 such that

$$\begin{bmatrix} Y(A+B_2F)^T + (A+B_2F)Y & B_1 & Y(C_1+D_{12}F)^T \\ B_1^T & -\gamma I & D_{11}^T \\ (C_1+D_{12}F)Y & D_{11} & -\gamma I \end{bmatrix} < 0$$

Let
$$Z = FY$$
. Then

$$\begin{bmatrix} YA^T + Z^TB_2^T + AY + B_2Z & B_1 & YC_1^T + Z^TD_{12}^T)^T \\ B_1^T & -\gamma I & D_{11}^T \\ C_1Y + D_{12}Z & D_{11} & -\gamma I \end{bmatrix}$$

$$= \begin{bmatrix} YA^T + YF^TB_2^T + AY + B_2FY & B_1 & YC_1^T + YF^TD_{12}^T)^T \\ B_1^T & -\gamma I & D_{11}^T \\ C_1Y + D_{12}FY & D_{11} & -\gamma I \end{bmatrix}$$

$$= \begin{bmatrix} Y(A + B_2F)^T + (A + B_2F)Y & B_1 & Y(C_1 + D_{12}F)^T \\ B_1^T & -\gamma I & D_{11}^T \\ (C_1 + D_{12}F)Y & D_{11} & -\gamma I \end{bmatrix} < 0.$$

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Proof.

Now suppose there exists a Y > 0 and Z such that

$$\begin{bmatrix} YA^T + Z^TB_2^T + AY + B_2Z & B_1 & YC_1^T + Z^TD_{12}^T \\ B_1^T & -\gamma I & D_{11}^T \\ C_1Y + D_{12}Z & D_{11} & -\gamma I \end{bmatrix} < 0$$

Let $F = ZY^{-1}$. Then

$$\begin{bmatrix} Y(A+B_{2}F)^{T} + (A+B_{2}F)Y & B_{1} & Y(C_{1}+D_{12}F)^{T} \\ B_{1}^{T} & -\gamma I & D_{11}^{T} \\ (C_{1}+D_{12}F)Y & D_{11} & -\gamma I \end{bmatrix}$$

$$= \begin{bmatrix} YA^{T} + YF^{T}B_{2}^{T} + AY + B_{2}FY & B_{1} & YC_{1}^{T} + YF^{T}D_{12}^{T} \\ B_{1}^{T} & -\gamma I & D_{11}^{T} \\ C_{1}Y + D_{12}FY & D_{11} & -\gamma I \end{bmatrix}$$

$$= \begin{bmatrix} YA^{T} + Z^{T}B_{2}^{T} + AY + B_{2}Z & B_{1} & YC_{1}^{T} + Z^{T}D_{12}^{T} \\ B_{1}^{T} & -\gamma I & D_{11}^{T} \\ C_{1}Y + D_{12}Z & D_{11} & -\gamma I \end{bmatrix} < 0$$

M. Peet Lecture 20: 12 / 13

Therefore the following optimization problems are equivalent

Form A

$$\min_F \lVert \underline{\mathsf{S}}(P,K(0,0,0,F)) \rVert_{H_\infty}$$

Form B

$$\min_{\gamma,Y,Z} \gamma: \\
\begin{bmatrix}
-Y & 0 & 0 & 0 \\
0 & YA^T + AY + Z^TB_2^T + B_2Z & B_1 & YC_1^T + Z^TD_{12}^T \\
0 & B_1^T & -\gamma I & D_{11}^T \\
0 & C_1Y + D_{12}Z & D_{11} & -\gamma I
\end{bmatrix} < 0$$

The optimal controller is given by $F = ZY^{-1}$.

Next Lecture: Optimal Output Feedback

M. Peet Lecture 20: 13 / 13