

Lecture 6

Control Synthesis Using Linear Matrix Inequalities.

Text: Duernd/Paganini, chapter 7.

- o H_2 Optimal State Feedback Using LMIs
- o The Kalman - Yakubovich - Lemma
- o H_∞ Optimal State Feedback
- o The Matrix Elimination Lemma
- o H_∞ Optimal Output Feedback

H_2 Optimal State Feedback Using LMIs

Problem: Given $\dot{x} = Ax + B_1 w + B_2 u$ find a stabilizing control law $u = Kx$ that minimizes $E(|x|^2 + |u|^2)$.

Solution: The closed loop system is $\dot{x} = (A + B_2 K)x + B_1 w$, so $\Sigma := Exx^T$ satisfies $(A + B_2 K)\Sigma + \Sigma(A + B_2 K)^T + B_1 B_1^T = 0$. This is a linear constraint on $(\Sigma, K\Sigma)$! ($Y = K\Sigma$)

Minimize $\text{trace } \Sigma + \text{trace}(Y \Sigma^{-1} Y)$

with $\Sigma > 0$, $(A\Sigma + B_2 Y) + (A\Sigma + B_2 Y)^T + B_1 B_1^T = 0$

Then the optimal control law is $K = Y \Sigma^{-1}$!

Stationary Stochastic Processes

If A is Hurwitz and w is white noise with intensity I , then the stationary solution to

$\dot{x} = Ax + Bw$ has covariance $Exx^T = \Sigma$ satisfying

$$A\Sigma + \Sigma A^T + B B^T = 0$$

The KYP Lemma

Given A, B and $M = M^T$, with $\det(i\omega I - A) \neq 0$ for $\omega \in \mathbb{R}$, the following are equivalent:

(i)
$$\begin{bmatrix} (i\omega I - A)^{-1} B \\ I \end{bmatrix}^* M \begin{bmatrix} (i\omega I - A)^{-1} B \\ I \end{bmatrix} < 0 \quad \text{for all } \omega.$$

(ii) There exists $P = P^T$ such that

$$M + \begin{bmatrix} A^T P + P A & P B \\ B^T P & 0 \end{bmatrix} < 0$$

The KYP is a classical result in systems theory connecting frequency domain to time domain.

Proof Sketch for KYP Lemma

Suppose (ii) holds and (x, w) is a ^{nonzero} square integrable solution to $\dot{x} = Ax + Bw$.

$$0 > \int_{-\infty}^{\infty} \begin{bmatrix} x \\ w \end{bmatrix}^T \begin{bmatrix} M + PA & PB \\ B^T P & 0 \end{bmatrix} \begin{bmatrix} x \\ w \end{bmatrix} dt = \int_{-\infty}^{\infty} \begin{bmatrix} x \\ w \end{bmatrix}^T M \begin{bmatrix} x \\ w \end{bmatrix} + \dot{x}^T P \dot{x} dt$$

$$= \int_{-\infty}^{\infty} \begin{bmatrix} x \\ w \end{bmatrix}^T M \begin{bmatrix} x \\ w \end{bmatrix} dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{w}(i\omega)^* \begin{bmatrix} (i\omega - A)^T B^* \\ I \end{bmatrix} M \begin{bmatrix} \hat{x}(i\omega) \\ \hat{w}(i\omega) \end{bmatrix} d\omega$$

Hence (i) follows.

The implication (i) is more difficult, but can be done by a separating hyperplane argument.

More equivalent conditions:

$$(3) \begin{bmatrix} (i\omega - A_K)^T B_1^* \\ I \end{bmatrix} M \begin{bmatrix} (i\omega - A_K)^T B_1 \\ I \end{bmatrix} \leq 0 \text{ when } M = \begin{bmatrix} I + K^T K & 0 \\ 0 & -\gamma^2 I \end{bmatrix}$$

$$(4) \exists P = P^T: \begin{bmatrix} A_K^T P + P A_K + I + K^T K & P B_1 \\ B_1^T P & -\gamma^2 I \end{bmatrix} \leq 0$$

$$(5) \exists P = P^T: \begin{bmatrix} (A_1^T + B_2 K P^T)^T + A_1 P^{-1} + B_2 K P^{-1} + P^{-1} (I + K^T K) P^{-1} & B_1^T \\ B_1^T & -\gamma^2 I \end{bmatrix} \leq 0$$

Solve by convex optimization over $(P^{-1}, K P^{-1}) \forall$.

H_∞ Optimal State Feedback

Problem: Given $\dot{x} = Ax + B_1 u + B_2 w$, $x(0) = 0$, find control law $u = Kx$ such that the closed loop satisfies

$$\int_0^{\infty} |x|^2 + |u|^2 dt \leq \gamma^2 \int_0^{\infty} |w|^2 dt$$

Equivalent conditions (with notation $A_K = A + B_2 K$)

$$(1) \int_{-\infty}^{\infty} |\hat{x}|^2 + |\hat{u}|^2 d\omega \leq \gamma^2 \int_{-\infty}^{\infty} |\hat{w}|^2 d\omega \text{ when } \begin{cases} \hat{x} = (i\omega - A_K)^{-1} B_1 \hat{w} \\ \hat{u} = K \hat{x} \end{cases}$$

$$(2) B_1^T (i\omega - A_K)^{-*} (i\omega - A_K)^{-1} B_1 + B_1^T (i\omega - A_K)^{-*} K^T K (i\omega - A_K)^{-1} B_1 \leq \gamma^2 I$$

Comparison to Riccati Approach

o More expensive to solve LMIs than ARES.

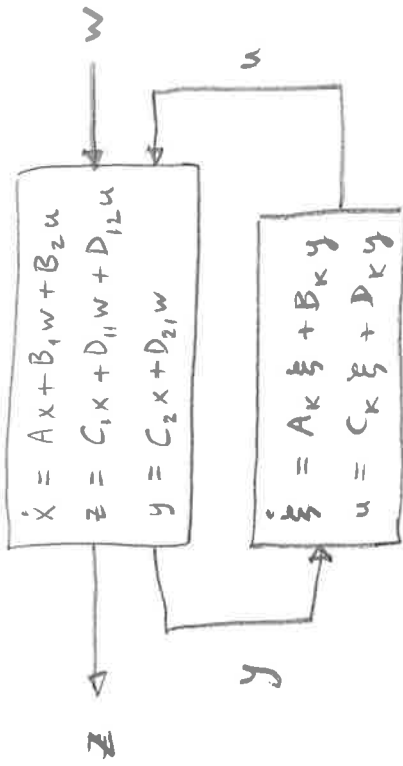
o Fewer assumptions

o Extensions possible at the expense of conservatism

- Diagonal P^{-1} and sparse $K P^{-1}$ gives sparse controller K .

- H_2 and H_{∞} specifications can be merged.

Output Feedback Synthesis



Determine $J := \begin{bmatrix} A_K & B_K \\ C_K & D_K \end{bmatrix}$ to get $\int_0^\infty |z|^2 dt \leq \gamma^2 \int_0^\infty |w|^2 dt$.

Notation

$$\bar{A} = \begin{bmatrix} A & 0 \\ 0 & 0 \end{bmatrix} \quad \bar{B} = \begin{bmatrix} B_1 \\ 0 \end{bmatrix} \quad \bar{C} = [C_1 \ 0] \quad \bar{D} = \begin{bmatrix} 0 & I \\ C_2 & 0 \end{bmatrix}$$

$$\underline{B} = \begin{bmatrix} 0 & B_2 \\ I & 0 \end{bmatrix} \quad \underline{D}_{21} = \begin{bmatrix} 0 \\ D_{21} \end{bmatrix} \quad \underline{D}_{12} = [0 \ D_{12}]$$

Then

$$\begin{bmatrix} A_{cl} & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix} = \begin{bmatrix} \bar{A} & \bar{B} \\ \bar{C} & \bar{D} \end{bmatrix} + \begin{bmatrix} B \\ D_{12} \end{bmatrix} J \begin{bmatrix} C \\ D_{21} \end{bmatrix}$$

$$\hat{M}_{cl}(s) = C_{cl}(sI - A_{cl})^{-1} B_{cl} + D_{cl}$$

Closed Loop System

$$\begin{bmatrix} \dot{\bar{x}} \\ \dot{\bar{\xi}} \end{bmatrix} = \underbrace{\begin{bmatrix} A + B_2 D_K C_2 & B_2 C_K \\ B_K C_2 & A_K \end{bmatrix}}_{A_{cl}} \begin{bmatrix} x \\ \xi \end{bmatrix} + \underbrace{\begin{bmatrix} B_1 + B_2 D_K D_{21} \\ B_K D_{21} \end{bmatrix}}_{B_{cl}} w$$

$$z = \underbrace{[C_1 + D_{12} D_K C_2 \quad D_{12} C_K]}_{C_{cl}} \begin{bmatrix} x \\ \xi \end{bmatrix} + \underbrace{(D_{11} + D_{12} D_K D_{21})}_{D_{cl}} w$$

LMI condition on the closed loop

It follows from the KYP Lemma that the following two conditions are equivalent:

(i) A_{cl} is Hurwitz and $\|\hat{M}_{cl}\|_\infty < 1$

(ii) There exists a matrix $\bar{X}_{cl} = \bar{X}_{cl}^T$ such that

$$\begin{bmatrix} A_{cl}^T \bar{X}_{cl} + \bar{X}_{cl} A_{cl} & \bar{X}_{cl} B_{cl} & C_{cl}^T \\ B_{cl}^T \bar{X}_{cl} & -I & D_{cl}^T \\ C_{cl} & D_{cl} & -I \end{bmatrix} < 0 \quad (*)$$

Notice: The inequality is affine in \bar{X}_{cl} and J individually, but not jointly.

The Matrix Elimination Lemma (Dullerud/Paganini 7.2)

Given matrices $P, Q, H = H^T$ let N_P and N_Q be full rank matrices with $\text{Im } N_P = \text{Ker } P, \text{Im } N_Q = \text{Ker } Q$.

Then the following are equivalent:

- (i) There exists J with $H + P^T J^T Q + Q^T J P^T < 0$
- (ii) The inequalities $N_P^T H N_P < 0, N_Q^T H N_Q < 0$ hold.

The inequality $N_P^T H N_P < 0$ can equivalently be written

$$N_P^T T_{\Sigma_{cl}} N_P < 0$$

where

$$T_{\Sigma_{cl}} = \begin{bmatrix} \bar{A} \Sigma_{cl}^{-1} + \Sigma_{cl}^{-1} \bar{A}^T & \bar{B} & \Sigma_{cl}^{-1} \bar{C}^T \\ \bar{B}^T & -I & D_{11}^T \\ \bar{C} \Sigma_{cl}^{-1} & D_{11} & -I \end{bmatrix}$$

Notice: $\Sigma_{cl} = \begin{bmatrix} \Sigma & \Sigma_2 \\ \Sigma_2^T & \Sigma_3 \end{bmatrix}$ and $H_{\Sigma_{cl}}$ depends only on Σ .
 $\Sigma_{cl}^{-1} = \begin{bmatrix} Y & Y_2 \\ Y_2^T & Y_3 \end{bmatrix}$ and $T_{\Sigma_{cl}}$ depends only on Y .

Notice that (*) can be written as

$$H_{\Sigma_{cl}} + Q^T J^T P_{\Sigma} + P_{\Sigma}^T J Q < 0$$

where

$$H_{\Sigma_{cl}} = \begin{bmatrix} \bar{A}^T \Sigma_{cl} + \Sigma_{cl} \bar{A} & \Sigma_{cl} \bar{B} & \bar{C}^T \\ \bar{B}^T \Sigma_{cl} & -I & D_{11}^T \\ \bar{C} & D_{11} & -I \end{bmatrix} \quad P_{\Sigma} = \begin{bmatrix} \bar{B}^T \Sigma_{cl} & 0 & D_{12}^T \\ 0 & 0 & 0 \end{bmatrix}$$

$$Q = \begin{bmatrix} \bar{C} & D_{21} & 0 \end{bmatrix}$$

By the Matrix Elimination Lemma, J exists iff

$$N_{P_{\Sigma}}^T H_{\Sigma_{cl}} N_{P_{\Sigma}} < 0 \quad \text{and} \quad N_Q^T H_{\Sigma_{cl}} N_Q < 0$$

Lemma (Dullerud/Paganini 7.9)

Given $\Sigma > 0$ and $Y > 0$, the following are equivalent:

- (i) There exist $\Sigma_2, Y_2, \Sigma_3 = \Sigma_3^T, Y_3 = Y_3^T \in \mathbb{R}^{n \times n}$ such that $0 < \begin{bmatrix} \Sigma & \Sigma_2 \\ \Sigma_2^T & \Sigma_3 \end{bmatrix} = \begin{bmatrix} Y & Y_2 \\ Y_2^T & Y_3 \end{bmatrix}^{-1}$

- (ii) $\begin{bmatrix} \Sigma & I \\ I & Y \end{bmatrix} > 0$ and $\text{rank} \begin{bmatrix} \Sigma & I \\ I & Y \end{bmatrix} \leq n + n_k$

The rank condition goes away if $n_k = n$.

H_∞ Output Feedback Synthesis

A controller that gives $\|M_d\|_{\infty} < 1$ exists iff there exist symmetric matrices $X > 0, Y > 0$ such that

$$\begin{bmatrix} N_0 & 0 \\ 0 & I \end{bmatrix}^T \begin{bmatrix} A^T X + XA & XB & C^T \\ B^T X & -I & D_1^T \\ C & D_1 & -I \end{bmatrix} \begin{bmatrix} N_0 & 0 \\ 0 & I \end{bmatrix} < 0$$

$$\begin{bmatrix} X & 0 \\ 0 & I \end{bmatrix} > 0, \quad \begin{bmatrix} N_c & 0 \\ 0 & I \end{bmatrix}^T \begin{bmatrix} AY + AY^T & YC^T & B_1^T \\ C_1 Y & -I & D_2^T \\ B_1^T & D_2 & -I \end{bmatrix} \begin{bmatrix} N_c & 0 \\ 0 & I \end{bmatrix} < 0$$

where N_0, N_c are full rank matrices with $\text{Im } N_0 = \text{Ker } [C_2 \ D_2]$, $\text{Im } N_c = \text{Ker } [B_2^T \ D_2^T]$.

Summary

- Both H_2 and H_{∞} synthesis problems can be stated as convex optimization in terms of linear matrix inequalities.
- Convexity requires $n_k = n$.
- The H_{∞} optimization involves coupling between estimation and control.