SNU MAE

Multivariable Control, Fall 08

## Homework Assignment 2

Issued: Oct 1 Due: Oct 15 (Wed)

1. Consider the linear, time-invariant system

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

with  $x(t) \in \mathbb{R}^n$ , and  $u(t) \in \mathbb{R}_{n_u}$ . Suppose the  $Q \in \mathbb{R}^{n \times n}$ , with  $Q = Q^T \ge 0$ . Assume that (A, B) is stabilizable, and  $(A, Q^{\frac{1}{2}})$  is detectable. Using a "completion of squares" approach, determine the value of the function  $J : \mathbb{R}^n \to \mathbb{R}$ 

$$J(x_0) := \min_{u \in \mathcal{L}_2} \int_0^\infty [x^T(t)Qx(t) + u^T(t)u(t)]dt$$

subject to

$$\begin{aligned} x(0) &= x_0 \\ \dot{x}(t) &= Ax(t) + Bu(t) \end{aligned}$$

Also determine the input u which achieves the optimum.

2. Two uncertain parameters.

Consider a plant and a compensator shown in Fig. 1, with the following transfer functions.

$$P(s) = \frac{g}{s^2(1+s\theta)}, \quad C(s) = \frac{k+T_ds}{1+T_0s}.$$

This model includes scaling factors  $\alpha_1$  and  $\beta_1$  for the uncertainty  $\Delta_g$  and scaling factors  $\alpha_2$  and  $\beta_2$  for the uncertainty  $\Delta_{\theta}$ .  $\alpha_1\beta_1 = \varepsilon_1$  is the largest possible uncertainty g, while  $\alpha_2\beta_2 = \varepsilon_2$  is the largest possible uncertainty in  $\theta$ .

(a) Show that the transfer matrix from  $p_1, p_2$  to  $q_1, q_2$  is in the following form:

$$H(s) = \frac{\frac{1}{1+s\theta_0}}{1+L_0(s)} \begin{bmatrix} \beta_1 C(s) \\ -\beta_2 \end{bmatrix} \begin{bmatrix} -\alpha_1/s^2 & \alpha_2 s \end{bmatrix}.$$

(b) Show that the largest eigenvalue of  $H^T(-j\omega)H(j\omega)$  is

$$\begin{split} \bar{\sigma}^2(\omega) &= \frac{\frac{1}{1+\omega^2\theta_0^2}}{|1+L_0(j\omega)|^2} (\beta_1^2 |C(j\omega)|^2 + \beta_2^2) (\frac{\alpha_1^2}{\omega^4} + \alpha_2^2 \omega^2), \quad \omega \in \mathbb{R}. \\ &= \frac{(\beta_1^2 (k^2 + \omega^2 T_d^2) + \beta_2^2 (1 + \omega^2 T_0^2)) (\alpha_1^2 + \alpha_2^2 \omega^6)}{|\chi(j\omega)|^2} \end{split}$$

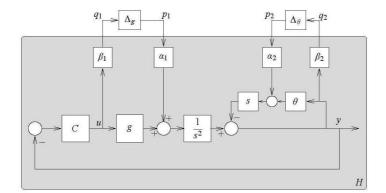


Figure 1: (Prob 2) Two uncertain parameters

with  $\chi$  the closed-loop characteristic polynomial

$$\chi(s) = \theta_0 T_0 s^4 + (\theta_0 + T_0) s^3 + s^2 + g_0 T_d s + g_0 k.$$

with the nominal values  $g_0, \theta_0$ .

(c) For k = 1,  $T_d = \sqrt{2}$ ,  $\theta_0 = 0.1$ ,  $T_0 = 1/10$ ,  $.5 \le g \le 5$ ,  $g_0 = 2.75$ , and  $\varepsilon_1 = 2.25$ ,  $\varepsilon_2 = 0.1$ ,

$$\alpha_1 = \beta_1 = \sqrt{\varepsilon_1}, \quad \alpha_2 = \beta_2 = \sqrt{\varepsilon_2}.$$

plot  $\bar{\sigma}(\omega)$ .

(d) Let  $\beta_1 = \varepsilon_1/\alpha_1$ ,  $\beta_2 = \varepsilon_2/\alpha_2$  and  $\rho = \alpha_1^2/\alpha_2^2$ . Show that, for fixed  $\omega$  the quantity  $\bar{\sigma}^2(\omega)$  is minimized for

$$\rho = \omega^3 \frac{\varepsilon_1}{\varepsilon_2} \sqrt{\frac{k^2 + \omega^2 T_d^2}{1 + \omega^2 T_0^2}},$$

and that for this value of  $\rho$ 

$$\bar{\sigma}(\omega) = \frac{\varepsilon_1 \sqrt{K^2 + \omega^2 T_d^2} + \varepsilon_2 \omega^3 \sqrt{1 + \omega^2 T_0^2}}{|\chi(j\omega)|}, \quad \omega \ge 0.$$

plot  $\bar{\sigma}(\omega)$  for the same case as (c). Compare the result with (c).

3. (a) Show that the structured singular value of the  $2 \times 2$  dyadic matrix

$$M = \begin{bmatrix} a_1b_1 & a_1b_2 \\ a_2b_1 & a_2b_2 \end{bmatrix} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} \begin{bmatrix} b_1 & b_2 \end{bmatrix}$$

with  $a_1, a_2, b_1$ , and  $b_2$  complex numbers, with respect to the perturbation structure

$$\Delta = \begin{bmatrix} \Delta_1 & 0\\ 0 & \Delta_2 \end{bmatrix}, \quad \Delta_1 \in \mathbb{C}, \ \Delta_2 \in \mathbb{C}$$

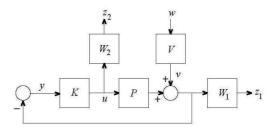


Figure 2: (Prob 4) mixed sensitivity design

is

$$\mu(M) = |a_1b_1| + |a_2b_2|.$$

- (b) Apply this fact to the prob 3, and compute  $\mu(H(j\omega))$ . Compare the result with 2(d).
- 4. We want to design a mixed-sensitivity controller for  $P(s) = 1/s^2$  using

$$\left\| \left[ \begin{array}{c} W_1 SV \\ W_2 UV \end{array} \right] \right\|_{\infty}$$

where  $U = K(I + PK)^{-1}$   $S = (I + PK)^{-1}$  and the weighting functions are:

$$V(s) = \frac{s^2 + s\sqrt{2} + 1}{s^2}$$

 $W_1 = 1$ , and  $W_2(s) = c(1 + rs)$ .

(a) Show that, when r = 0, the plant is

$$\dot{x} = \underbrace{\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}}_{A} x + \underbrace{\begin{bmatrix} 1 \\ \sqrt{2} \end{bmatrix}}_{B_1} w + \underbrace{\begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{B_2} u,$$

$$z = \underbrace{\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}}_{C_1} x + \underbrace{\begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{D_{11}} w + \underbrace{\begin{bmatrix} 0 \\ c \end{bmatrix}}_{D_{12}} u,$$

$$y = \underbrace{\begin{bmatrix} 0 & -1 \end{bmatrix}}_{C_2} x + \underbrace{\begin{bmatrix} -1 \end{bmatrix}}_{D_{21}} w.$$

(b) What do you think we can expect from the given choices of weighting functions (when r = 0 and  $r \neq 0$ )?

- (c) Design  $H_{\infty}$  controller using Matlab for r = 0, c = 0.1. What is the value of the resulting  $H_{\infty}$  norm?
- 5. [EXTRA] State-space calculation of  $\|\cdot\|_{\infty}$  norm : Consider a linear system

$$\dot{x}(t) = Ax(t) + Bu(t) y(t) = Cx(t)$$

In this analysis, it is not necessary to assume that A is stable, but we must assume that A has no imaginary-axis eigenvalues. Introduce the notation  $G^{\sim}(s) = [G(-\bar{s})]^*$ .

First, check that

$$M(s) := [I - G(s)G^{\sim}(s)]^{-1} = \begin{bmatrix} A & BB^* & 0\\ -CC^* & -A^* & -C^*\\ \hline C & 0 & I \end{bmatrix}$$

- (a) Consider a given frequency  $\bar{\omega}$ . Show that  $G(j\bar{\omega})$  has a singular value equal to 1 (some singular value not necessarily the maximum) if and only if  $I G(s)G^{\sim}(s)$  is singular at  $s = j\bar{\omega}$ .
- (b) Consider a given frequecy  $\bar{\omega}$ . Show that  $G(j\bar{\omega})$  has a singular value equal to 1 if and only if M(s) has a pole at  $s = j\bar{\omega}$ .
- (c) Hence, the imaginary axis poles of M(s) are the same as the points on the imaginary axis where G has a singular value equal to 1. In this part, we will show that the imaginary-axis poles of M(s) are exactly equal to the imaginary axis eigenvalues of the "A" matrix for M (it could be possible that some eigenvalues of the "A" matrix are uncontrollable and/or unobservable, so they would not show up in the transfer function - this calculation will rule that possibility out). To do this, show the any imaginary-axis eigenvalues of

$$\left[\begin{array}{cc} A & BB^* \\ -CC^* & -A^* \end{array}\right]$$

are controllable through

$$\left[\begin{array}{c} 0\\ -C^* \end{array}\right],$$

and unobservable through

$$\begin{bmatrix} C & 0 \end{bmatrix}.$$

[Hint] Use the Popov-Bellman-Hautus (Kailath) test for controllability and observability.

(d) Hence, we have proven the statement:  $G(j\omega)$  has a singular value equal to 1 if and only if

$$\left[\begin{array}{cc} A & BB^* \\ -CC^* & -A^* \end{array}\right]$$

has eigenvalue equal to  $j\omega$ . In other words: for all  $\omega \in \mathbb{R}, G(j\omega)$  has no singular values equal to 1 if and only if

$$\left[\begin{array}{cc} A & BB^* \\ -CC^* & -A^* \end{array}\right]$$

has no imaginary axis eigenvalues. Generalize these two statements to the case where  $G(j\omega)$  has a singular value equal to some positive number  $\gamma \neq 1$ .

(e) Prove the following: For  $\gamma > 0$ ,

$$\sup_{\omega \in \mathbb{R}} \bar{\sigma}[G(j\omega)] < \gamma$$

if and only if

$$\left[\begin{array}{cc} A & \frac{1}{\gamma^2}BB^*\\ -CC^* & -A^* \end{array}\right]$$

has no imaginary axis eigenvalues.

Two good reference for this problem are

- Boyd, Balakkrishnan and Kabamba, "A bisection method for computing the  $H_{\infty}$  norm of a transfer matrix and related problems," *Math Control Signals and Systems*, 2(3):207-219, 1989.

- Bruinsma and Steinbuch, "A fast algorithm to compute the  $H_{\infty}$  norm of a transfer function matrix, *Systems and Control Letters*, 14, pp. 287-293, 1990.