

Module 4F2, May 2009 – Robust Multivariable Control – Solutions

ENGINEERING TRIPPOS PART IIB

Wednesday 29 April 2009 2.30 to 4

Module 4F2

ROBUST MULTIVARIABLE CONTROL

Answer not more than two questions.

All questions carry the same number of marks.

The approximate percentage of marks allocated to each part of a question is indicated in the right margin.

There are no attachments.

STATIONERY REQUIREMENTS

Single-sided script paper

SPECIAL REQUIREMENTS

Engineering Data Book

CUED approved calculator allowed

**You may not start to read the questions
printed on the subsequent pages of this
question paper until instructed that you
may do so by the Invigilator**

①

[4+2 - 2009]

Guy-Bart Stan

1 ② Definition : $\|\hat{G}(s)\|_2 = \sqrt{\int_{-\infty}^{+\infty} \text{trace}((\hat{G}(jw))^* \hat{G}(jw)) dw}$ ②

Interpretation in terms of $\|w\|_2$ and $\|z\|_\infty$: $\|z\|_\infty = \sup_t \sqrt{z^T(t) z(t)} \leq \frac{1}{\sqrt{2\pi}} \|\hat{G}(s)\|_2 \|w\|_2$
with $\|w\|_2 = \sqrt{\int_{-\infty}^{+\infty} w^T(t) w(t) dt}$

ii) $\|\hat{G}(s)\|_2 = \sqrt{2\pi} \sqrt{\text{trace}(B^T L B)}$ where L is the solution to the Lyapunov equation

$$A^T L + L A + C^T C = 0$$

The Lyapunov equation admits a solution $L \geq 0$ if the system is stable, i.e. if A has all its eigenvalues in the left hand plane. The solution is positive definite, i.e. $L > 0$ if in addition the system is observable.

b) i) The closed loop system is

$$\begin{cases} \dot{x} = (A - B_2 K)x + B_2 w \\ z = C_2 - D_{22} K x \end{cases}$$

$F_L(s, K)$ is the transfer function from w to z for this closed loop system. Therefore (using ② ii)

$$\|F_L(s, K)\|_2^2 = 2\pi \text{trace}(B_2^T L B_2)$$

where L solves the Lyapunov equation

$$(A - B_2 K)^T L + L(A - B_2 K) + (C_2 - D_{22} K)^T (C_2 - D_{22} K) = 0$$

$$\Leftrightarrow (A - B_2 K)^T L + L(A - B_2 K) + C_2^T C_2 - \cancel{C_2^T D_{22} K} - \cancel{K^T D_{22}^T C_2} + K^T D_{22}^T D_{22} K = 0$$

We know that $D_{22}^T C_2 = 0$ and $D_{22}^T D_{22} = I$
 $(\Leftrightarrow C_2^T D_{22} = 0)$

Therefore, we have that L solves the Lyapunov equation

$$(A - B_2 K)^T L + L(A - B_2 K) + C_2^T C_2 + K^T K = 0$$

(2)

(2)

(ii) we have to show that

$$(A - B_2 K)^T (L - X) + (L - X)(A - B_2 K) + (K - B_2^T X)^T (K - B_2^T X) = 0 \quad (*)$$

If we expand this equation, we obtain

$$\begin{aligned} & \underbrace{(A - B_2 K)^T L + L(A - B_2 K) + K^T K}_{= -C_2^T C_2 \text{ (by (1) (i))}} \\ & \underbrace{-A^T X - XA + X^T B_2 B_2^T X + K^T B_2 X + X B_2 K - X B_2 K - K B_2^T X}_{= C_2^T C_2} \\ & = -C_2^T C_2 + C_2^T C_2 = 0 \quad O.K. \end{aligned}$$

By (2) (ii), the Lyapunov equation (*) has a positive semi-definit solution $(L - X) \geq 0$ if and only if $(A - B_2 K)$ is stable

$(A - B_2 K)$ is stable since, by assumption, K is a stabilizing controller

(iii) $\min_{K \text{ stabilizing}} \| \mathcal{J}_L(L, K) \|_2 = \min_{K \text{ stabilizing}} \sqrt{2\pi} \sqrt{\text{trace}(B_2^T L B_2)}$

by (2) (ii) if $L = L^T \geq 0$ is the solution to $(A - B_2 K)^T L + L(A - B_2 K) + C_2^T C_2 + K^T K = 0$
 then $L - X \geq 0$

Now, $L - X \geq 0 \Rightarrow B_2^T (L - X) B_2 \geq 0$
 (using the hint) $\text{trace}(B_2^T (L - X) B_2) \geq 0$
 $\Rightarrow \boxed{\text{trace}(B_2^T L B_2) \geq \text{trace}(B_2^T X B_2)}$

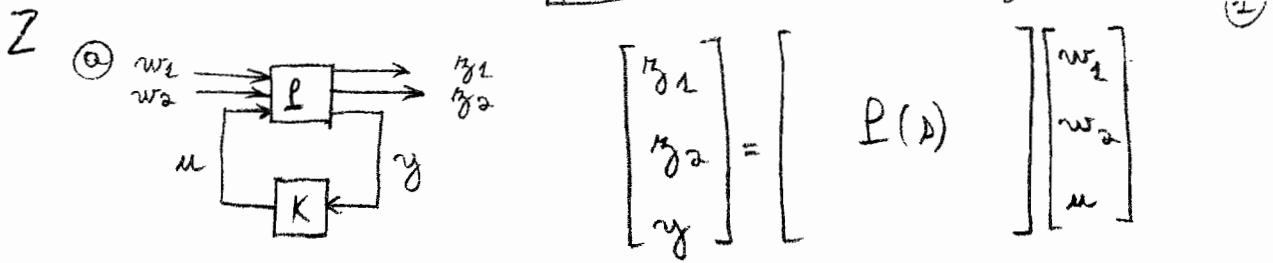
Therefore, the minimum is achieved when $L = X$

$$\Rightarrow \min_{K \text{ stabilizing}} \| \mathcal{J}_L(L, K) \|_2 = \sqrt{2\pi} \sqrt{\text{trace}(B_2^T X B_2)}$$

and the controller that achieves the minimum is

$$K = B_2^T X \text{ from } (*) \text{ with } L = X$$

(3)



From inspection of the block diagram, we have

$$\begin{cases} y = w_2 + y_2 \\ y_2 = \tilde{G} u + w_1 \\ y_2 = u = Ky \end{cases} \Rightarrow L(\Delta) = \begin{bmatrix} I & 0 & \tilde{G} \\ 0 & 0 & I \\ I & I & \tilde{G} \end{bmatrix}$$

$$\begin{aligned} \mathcal{F}_e(L, K) &= T \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} \rightarrow \begin{bmatrix} y_1 \\ y_2 \\ y \end{bmatrix} \Rightarrow y = w_2 + \tilde{G} K y + w_1 \\ &\Leftrightarrow (I - \tilde{G} K) y = w_1 + w_2 \\ &\Rightarrow y = (I - \tilde{G} K)^{-1} (w_1 + w_2) \end{aligned}$$

$$\Rightarrow y_2 = K (I - \tilde{G} K)^{-1} (w_1 + w_2)$$

$$y_2 = \tilde{G} K (I - \tilde{G} K)^{-1} (w_1 + w_2) + w_1$$

$$\Leftrightarrow y_2 = (I + \tilde{G} K (I - \tilde{G} K)^{-1}) w_1 + \tilde{G} K (I - \tilde{G} K)^{-1} w_2$$

$$\mathcal{F}_e(L, K) = \begin{bmatrix} I + \tilde{G} K (I - \tilde{G} K)^{-1} & \tilde{G} K (I - \tilde{G} K)^{-1} \\ K (I - \tilde{G} K)^{-1} & K (I - \tilde{G} K)^{-1} \end{bmatrix}$$

$$\tilde{G} = \frac{k}{\alpha - \omega} \quad (\text{unstable pole at } \alpha > 0)$$

(b) $\left\{ \begin{array}{l} \dot{x} = \alpha x + k u + w_1, x \in \mathbb{R} \\ x_{\infty} = x, \alpha > 0, k > 0 \end{array} \right. \quad \begin{array}{l} y_2 = u \\ y = x + w_2 \end{array} \quad (w_1 \text{ is a process noise})$

State-space realisation for $L(\Delta)$

$$\begin{bmatrix} \dot{x} \\ y_1 \\ y_2 \\ y \end{bmatrix} = \begin{bmatrix} a & & & \\ +1 = c_1 & 1 & 0 & k \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ w_1 \\ w_2 \\ u \end{bmatrix}$$

The problem is of H_2 optim. control with output feedforward $\begin{cases} B_1 = 1, B_2 = k, C_1 = 1, C_2 = 1 \\ A = \alpha \end{cases}$

(4)

$$\text{min}_{K(\Delta) \text{ stabilising}} \| \mathcal{J}_\ell(P(\Delta), K(\Delta)) \|_2^2 = k\pi \left[\text{trace}(B_2^T X B_2) + \text{trace}(F Y F^T) \right]$$

$$\text{where } F = B_2^T X$$

X is the stabilising solution of CARE:

$$\textcircled{1} \quad X A + A^T X + C_2^T C_2 - X B_2 B_2^T X = 0$$

Y is the stabilising solution of FARE:

$$\textcircled{2} \quad Y A^T + A Y + B_2 B_2^T - Y C_2 C_2^T Y = 0$$

$$\text{CARE } \textcircled{1} \quad \alpha x^* + 1 - k^2 x^{*2} = 0 \Leftrightarrow k^2 x^{*2} - \alpha x^* - 1 = 0$$

$$\Rightarrow x^* = \frac{\alpha \pm \sqrt{\alpha^2 + k^2}}{k^2}$$

(the stabilising solution
is the > 0 one $\forall \alpha, k > 0$)

$$\text{FARE } \textcircled{2} \quad + y^* - \alpha x^* y^* - 1 = 0 \Rightarrow y^* = \frac{\alpha + \sqrt{\alpha^2 + 1}}{x^*}$$

$$F = B_2^T X = k x^*$$

$$\min_{K(\Delta) \text{ stab}} \| \mathcal{J}_\ell(P(\Delta), K(\Delta)) \|_2^2 = k\pi \left[\text{trace}(B_2^T X B_2) + \text{trace}(F Y F^T) \right]$$

$$\boxed{B_2^T X B_2 = x^*}$$

$$\boxed{F Y F^T = k^2 x^{*2} y^*}$$

$$\begin{aligned} \min_{K(\Delta) \text{ stab}} \| \mathcal{J}_\ell(P(\Delta), K(\Delta)) \|_2^2 &= \sqrt{2\pi} \sqrt{x^* + k^2 x^{*2} y^*} \\ &= \sqrt{2\pi} \sqrt{x^* (1 + k^2 x^{*2} y^*)} \\ &= \sqrt{2\pi} \sqrt{\frac{(\alpha + \sqrt{\alpha^2 + k^2}) (1 + k^2 \frac{(\alpha + \sqrt{\alpha^2 + k^2})(\alpha + \sqrt{\alpha^2 + 1})}{k^2})}{k^2}} \\ &= \frac{\sqrt{2\pi}}{k} \sqrt{\alpha + \sqrt{\alpha^2 + k^2}} \sqrt{1 + (\alpha + \sqrt{\alpha^2 + k^2})(\alpha + \sqrt{\alpha^2 + 1})} \end{aligned}$$

(5)

- d) The optimal controller K has the following state space realization

$$\begin{bmatrix} \dot{x}_K \\ u \end{bmatrix} = \left[\begin{array}{c|c} A - B_2 F - H C_2 & -H \\ \hline F & 0 \end{array} \right] \begin{bmatrix} x_K \\ y \end{bmatrix}$$

$$\text{with } F = B_2^T X = k x^*, \quad x^* = \frac{\alpha + \sqrt{\alpha^2 + k^2}}{k^2}$$

$$H = Y C_2^T = y^*, \quad y^* = \alpha + \sqrt{\alpha^2 + 1}$$

$$B_2 = k$$

$$C_2 = 1$$

$$\Rightarrow \begin{bmatrix} \dot{x}_K \\ u \end{bmatrix} = \left[\begin{array}{c|c} \alpha - k^2 x^* - y^* & -y^* \\ \hline k x^* & 0 \end{array} \right] \begin{bmatrix} x_K \\ y \end{bmatrix}$$

- e) The poles of the closed loop system are the eigenvalues of $\underbrace{A - B_2 F}_{= \alpha - k^2 x^*}$ and those of $\underbrace{A - H C_2}_{= \alpha - y^*}$

$$= \alpha - k^2 x^*$$

$$= \alpha - y^*$$

$$= \alpha - \frac{k^2}{k^2} (\alpha + \sqrt{\alpha^2 + k^2})$$

$$= \alpha - \alpha - \sqrt{\alpha^2 + 1}$$

$$= -\sqrt{\alpha^2 + k^2} < 0 \quad \forall \alpha, k$$

$$= -\sqrt{\alpha^2 + 1} < 0 \quad \forall \alpha, k$$

\downarrow
stable

\downarrow
stable

Closed-loop system

$$\begin{aligned} \dot{x} &= \alpha x + k u + w_1 \\ \dot{x}_K &= -y^* y + (\alpha - k^2 x^* - y^*) x_K \\ y &= x + w_2 \\ u &= k x^* x_K \end{aligned} \Rightarrow \begin{cases} \dot{x} = \alpha x + k^2 x^* x_K + w_1 \\ \dot{x}_K = -y^* x + (\alpha - k^2 x^* - y^*) x_K - y^* w_2 \end{cases} \rightarrow A_{\text{closed-loop}} = \begin{pmatrix} \alpha & k^2 x^* \\ -y^* & \alpha - k^2 x^* - y^* \end{pmatrix}$$

$$\left(\begin{array}{l} \text{eigenvalues of } A: \lambda_1 \text{ & } \lambda_2 \\ \text{Trace}(A) = \lambda_1 + \lambda_2 = \alpha - k^2 x^* - y^* \\ \text{Det}(A) = \lambda_1 \lambda_2 = \alpha(\alpha - k^2 x^* - y^*) + k^2 x^* y^* \\ = (\alpha - k^2 x^*)(\alpha - y^*) \\ \Rightarrow \lambda_1 = \alpha - k^2 x^* < 0 \text{ & } \lambda_2 = \alpha - y^* < 0 \end{array} \right)$$

- 3 (a) $P(s)$ is in \mathcal{H}_∞ since all poles lie in the open left half plane.

$$\begin{aligned}\|P(s)\|_\infty &= \sup_{\omega} |P(j\omega)| \\ &= \sup_{\omega} \left| \frac{j\omega + 1}{(j\omega + 2)^2} \right| \\ &= \sup_{\omega} \frac{\sqrt{1 + \omega^2}}{4 + \omega^2}\end{aligned}$$

Differentiating with respect to ω^2 and equating to zero gives

$$\frac{(4 + \omega_0^2)^{1/2}(1 + \omega_0^2)^{-1/2} - \sqrt{1 + \omega_0^2}}{(4 + \omega_0^2)^2} = 0$$

or $4 + \omega_0^2 = 2(1 + \omega_0^2)$, which gives $\omega_0^2 = 2$. Thus,

$$\|P(s)\|_\infty = \frac{\sqrt{3}}{6}$$

[10%]

- (b) The small gain theorem for the robust stability of a system states that the feedback uncertain system is stable for all $\|\Delta(s)\|_\infty < \varepsilon$ if and only if $\|T(s)\|_\infty \leq 1/\varepsilon$. [10%]

- (c) Let $X = G_0U$, Then $Y = (I - W\Delta)^{-1}X$, or $Y = W\Delta Y + X$. Let also $Z = \Delta Y$. We need to find the transfer function from Z to Y . From above, $Y = WZ - G_0Y$ or $Y = (I + G_0)^{-1}WZ$. Thus, let $T = (I + G_0)^{-1}W$ and the result follows from the small gain theorem as stated in part (b). [30%]

- (d) (i) The feedback system is stable if and only if $(1 + G(s))^{-1}$ is stable.

In this case

$$(1 + G(s))^{-1} = \frac{(s+1)(s+2+\alpha)}{s^2 + (4+\alpha)s + 4 + \alpha}$$

which is stable if and only if $\alpha > -4$, so the maximum value of A such that the feedback system is guaranteed to be stable for all $|\alpha| < A$ is $A = 4$. [10%]

- (ii) We have that $G(s) = (1 - W\Delta)^{-1}G_0$. Rearranging gives $(1 - W\Delta)G = G_0$ or

$$W\Delta = \frac{G - G_0}{G} = 1 - \frac{G_0}{G} = 1 - \frac{s+2+\alpha}{s+2} = -\frac{\alpha}{s+2}$$

(cont.)

Let $W = A/(s+2)$ and $\Delta = -\alpha/A$ (which results in $\|\Delta\|_\infty < 1$). Stability condition is

$$\left\| \frac{W}{1+G_0} \right\|_\infty \leq 1$$

or

$$\left\| \frac{A}{s+2} \cdot \frac{s+1}{s+2} \right\|_\infty = A \frac{\sqrt{3}}{6} \leq 1$$

by part (a). Thus, $A \leq 2\sqrt{3} = 3.4641$.

[20%]

(iii) We have $G_0 = 1/(s+1) = N/M$ with $|N|^2 + |M|^2 = 1$. Let

$$M = \frac{s+1}{s+\sqrt{2}} \quad \text{and} \quad N = \frac{1}{s+\sqrt{2}}$$

Now, $G = G_0/(1 - W\Delta)$ and $W\Delta = -\alpha/(s+2)$. So,

$$G = \frac{N}{M \left(1 + \frac{\alpha}{s+2} \right)} = \frac{N}{M + \frac{\alpha(s+1)}{(s+2)(s+\sqrt{2})}}$$

Thus, just let

$$\Delta_N = 0, \quad \Delta_M = \frac{\alpha(s+1)}{(s+2)(s+\sqrt{2})}$$

Perturbations measured in the gap metric are equivalent to perturbations to the numerator and denominator in a normalised coprime factorisation:

$$G = (N + \Delta_N)(M + \Delta_M)^{-1}$$

with $\delta_g(G, G_0) = \min \left\| \begin{bmatrix} \Delta_N \\ \Delta_M \end{bmatrix} \right\|_\infty$ over all Δ_N, Δ_M in \mathcal{H}_∞ such that $G = (N + \Delta_N)(M + \Delta_M)^{-1}$. Hence, in the above $\delta_g(G, G_0) \leq \|\Delta_M\|_\infty = \left\| \frac{\alpha(s+1)}{(s+2)(s+\sqrt{2})} \right\|_\infty$.

[20%]

END OF PAPER