

EL2520 - Control Theory and Practice - Advanced  
Course  
Solution/Answers – 2018-05-29

1. (a) (i) The determinant is  $\det G = \frac{2(s-1)}{(s+2)^2}$  and the denominator is the LCD for all minors. Hence there are two poles in  $s = -2$  and one zero at  $s = 1$ . The zero direction is defined as  $y_z^H G(z) = 1$ , such that  $|y_z| = 1$  and we get  $y_z^H = \frac{1}{\sqrt{5}}[-2 \ 1]$ . (ii) With two poles we need 2 states in a minimal state space realization.
- (b) The singular values are given by  $\sigma_i(i\omega) = \sqrt{\lambda(G(i\omega)^H G(i\omega))} = \frac{1}{\sqrt{4\omega^2+1}} \sqrt{\lambda(G(0)^H G)}$  (here one can also use the fact that the singular values of a symmetric matrix equals the magnitude of the eigenvalues). This gives  $\bar{\sigma} = \frac{3}{\sqrt{4\omega^2+1}}$  and  $\underline{\sigma} = \frac{1}{\sqrt{4\omega^2+1}}$ . We have  $\|G\|_\infty = \sup_\omega \bar{\sigma}(G) = 3$ .
- (c) The closed-loop eigenvalue is  $2 - K$  and the system is stable for  $-1 < 2 - K < 1$  which gives  $1 < K < 3$ .
2. (a) Since  $\bar{\sigma}(S) > |1 - \bar{\sigma}(T)|$  we can not have  $\bar{\sigma}(S) < 0.5$  when  $\bar{\sigma}(T) < 0.5$  (and vice versa) at any frequency. Hence we can not have the magnitude of both weights  $> 2$  at the same frequency and get  $\|W_S S\|_\infty < 1$  and  $\|W_T T\|_\infty < 1$ . Here both weights are larger than 2 around frequency  $0.3 \text{ rad/s}$  and hence the objective is infeasible.
- (b) For internal stability we require that the sensitivity function  $S$  has a RHP zero at the same position as  $G(s)$  has a RHP pole. In this case we have a pole at  $s = 1$  in  $G(s)$ , but the only zero of  $S(s)$  is at  $s = 0$ . Hence the proposed controller does not give internal stability since there is a pole-zero cancellation in the RHP between the plant and the controller.
- (c) Here  $G(s)$  has a RHP zero at  $s = 1$ . For any RHP zero  $s = z > 0$  we require  $|y_z^H G_d(z)| < 1$  where  $y_z$  is the corresponding zero direction. We get  $y_z^H = \frac{1}{\sqrt{13}}[3 \ -2]$  and  $G_d(z) = \frac{1}{11}[2 \ 3]^T$  which gives  $|y_z^H G_d(z)| = 0 < 1$ . Since there are no other limitations, the objective is feasible.
3. (a) (i) We get  $T = K/(s - 1 + K)$  and  $W_T T = \frac{2K}{3} \frac{s+1}{s-1+K}$ . Closed-loop stability for  $K > 1$ . The peak-value of  $|W_T T(i\omega)|$ , corresponding to  $\|W_T T\|_\infty$ , occurs at  $\omega = 0$  if  $1 < K < 2$  and at  $\omega = \infty$  if  $K > 2$ . The peak values are  $2K/(3(-1 + K))$  and  $2K/3$ , respectively. Hence, the minimum is obtained for  $K = 2$  corresponding to  $\|W_T T\|_\infty = 4/3$ . (ii) Since there is a RHP pole at  $s = p = 1$  and  $|W_T(p)| = \frac{4}{3} > 1$  it is not possible to achieve  $\|W_T T\|_\infty < 1$ .
- (b) (i) We have  $y = G S_I w_u = S G w_u$  and hence we require  $\|W_1 S G\|_\infty < 1$  with  $W_1 = \frac{1}{2} \frac{s+1}{s}$  which enforces  $\bar{\sigma}(S G(i0.5)) < 1$  and  $\bar{\sigma}(S G(0)) =$

0. Robust stability with relative uncertainty at the input is ensured if  $\|W_2 T_I\|_\infty < 1$  where  $W_2 = 0.2$ , corresponding to the relative uncertainty. To get a scalar objective function value we stack the two objectives into one matrix to obtain

$$\min_{F_y} \left\| \begin{pmatrix} W_1 S G \\ W_2 T_I \end{pmatrix} \right\|_\infty$$

(ii) For  $SG$  we choose  $z_{e1} = W_1 y = W_1 S G w_u$ . For  $T_I$  we have that  $u = u_c + w_u$  where  $u_c$  is the output of the controller  $F_y$ , and hence  $u_c = F_y G(I + F_y G)^{-1} w_u = T_I w_u$ . Hence we choose  $z_{e2} = W_2 u_c = W_2 T_I w_u$ . Hence, with  $w_e = w_u$  we get in closed-loop

$$z_e = \begin{pmatrix} W_1 S G \\ W_2 T_I \end{pmatrix} w_e$$

Now the signal minimization problem given corresponds to minimizing the  $\mathcal{H}_\infty$ -norm of the transfer-function above.

4. (a) The system is on state space form

$$\dot{y} = 2u(t)$$

(i) This is an LQ problem with  $A = 0, B = 2, M = 1, Q_1 = 1, Q_2 = 0.1$ . The Riccati equation is then

$$1 - 40P^2 = 0$$

and the solution  $P > 0$  is  $P = 1/(2\sqrt{10})$ . The optimal feedback gain is then  $L = 20P = \sqrt{10}$ . Thus, optimal  $K_P = \sqrt{10}$ . (ii) We have  $A = 0, R_1 = 0, R_2 = 0.1, C = 1$  for the Riccati equation for the Kalman filter, yielding

$$10P^2 = 0$$

and hence  $P = 0$ . Thus, the optimal estimate is

$$\hat{y} = 2u(t)$$

(Note that there is no feedback from the measurement. This will by some not be considered a Kalman filter, as it often is a requirement that a process disturbance, i.e.,  $R_1 \neq 0$ , is included in the model in which case we would get  $P \neq 0$ .) The estimator does not influence the optimal gain in (i) due to the separation principle.

- (b) (i) The state-space model is  $\dot{y} = 2u(t)$  and hence, in discrete time,  $A = e^{0T} = 1$  and  $B = \int_0^T 2e^{0T} dt = 2T = 1$ . Thus, the discrete time state space model is

$$y_{k+1} = y_k + u_k$$

(ii) With  $N = 1$  we get the objective function to be minimized

$$V = Q_y y_k^2 + Q_y (y_k + u_k)^2 + u_k^2 + u_{k+1}^2 = 2Q_y y_k^2 + 2Q_y y_k u_k + (Q_y + 1)u_k^2 + u_{k+1}^2$$

Since we can not influence  $y_k$  (the current value of  $y$ ), we can remove the terms with only  $y_k$ . Also,  $u_{k+1}^2$  can be removed since obviously the optimum value will be  $u_{k+1} = 0$ . Thus,

$$V = (Q_y + 1)u_k^2 + 4Q_y y_k u_k$$

and hence  $H = Q_y + 1$  and  $h = 4Q_y y_k$ . The constraint  $|u_k| < 1$  can be written as  $u_k < 1$  and  $-u_k < 1$ , i.e.,  $L = [1 \ -1]^T$  and  $b = [1 \ 1]^T$ .

5. (a) Because we employ the Small Gain Theorem for analyzing stability of the loop with  $\Delta$  included, and then require all blocks to be stable.
- (b) Writing the closed-loop on  $M$ - $\Delta$ -form we get  $M = -W_{iI}S_I$  and hence from the SGT we get the sufficient robust stability condition

$$\|W_{iI}S_I\|_\infty < 1$$

and nominal stability.

- (c) We get

$$T_I = \frac{K_p}{K_p + s + 1} ; \quad S_I = \frac{s + 1}{K_p + s + 1}$$

and nominal stability for  $K_p > -1$ . For the uncertainty weights we get

$$k = (1 + W_I \Delta_I) \Rightarrow |W_I \Delta_I| < |1 - k| < 0.75$$

and

$$k = (1 + W_{iI} \Delta_{iI})^{-1} \Rightarrow |W_{iI} \Delta_{iI}| < \left| \frac{1}{k} - 1 \right| < 3$$

With these weights we get

$$\|W_I \Delta_I T_I\|_\infty = 0.75 \left| \frac{K_p}{K_p + 1} \right|$$

(peaks at  $\omega = 0$  since only one pole and no zeros) and hence RS for all  $K_p > -4/7$ . For  $-1 < K_p < 0$  we get

$$\|W_{iI} \Delta_{iI} S_I\|_\infty = 3 \frac{1}{K_p + 1} > 3$$

(peaks at  $\omega = 0$  since pole smaller magnitude than zero) while with  $K_p > 0$  we get

$$\|W_{iI} \Delta_{iI} S_I\|_\infty = 3$$

(peaks at  $\omega = \infty$  since zero smaller magnitude than pole) and hence we can not guarantee RS with any value of  $K_p$  with the inverse multiplicative uncertainty.

(d) For the relative input uncertainty we get

$$W_I \Delta_I = \frac{1 - \alpha}{s + \alpha}$$

and the robustness condition becomes

$$\|W_I \Delta_I T_I\|_\infty = \left| \frac{K_p}{K_p + 1} \right| \left| \frac{1 - \alpha}{\alpha} \right| \leq \left| \frac{3K_p}{1 + K_p} \right| < 1$$

which then gives  $-0.25 < K_p < 0.5$  for robust stability.

For the inverse uncertainty we get

$$W_{iI} \Delta_{iI} = \frac{\alpha - 1}{s + 1}$$

and the robustness condition

$$\|W_{iI} \Delta_{iI} S_I\|_\infty = \frac{|\alpha - 1|}{|1 + K_p|} \leq \frac{1}{1 + K_p} < 1$$

which gives that we have robust stability for all  $K_p > 0$ .

Thus, we see that it is important to choose the "correct" uncertainty description for robustness analysis, as we otherwise may get overly conservative results.