

# Linear System Theory

## Solution to Homework 8

1. (a) The closed-loop system is  $\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{b}_2 \mathbf{k}_2) \mathbf{x}$ . Let  $\mathbf{k}_2 = [k_{21}, k_{22}, k_{23}]$ .

Then

$$\det \begin{bmatrix} s+1 & 0 & 0 \\ -2 & s+2 & 3 \\ 2+k_{21} & k_{22} & s-1+k_{23} \end{bmatrix} = (s+1)((s+2)(s-1+k_{23})-3k_{22}) = (s+1)^3$$

Hence

$$s^2 + (1+k_{23})s - 2(1-k_{23}) - 3k_{22} = s^2 + 2s + 1 \Rightarrow k_{23} = 1, k_{22} = \frac{-1}{3}$$

$k_{21}$  can be arbitrary. We choose  $k_2 = [0, -\frac{1}{3}, 1]$ .

- (b) By PBH test,  $(\mathbf{A} - \mathbf{b}_1 \mathbf{k}_1, \mathbf{b}_2)$  is uncontrollable if and only if  $\exists \eta \in \mathbb{C}^3$  and  $\exists \lambda \in \mathbb{C}$  such that  $\eta^T (\mathbf{A} - \mathbf{b}_1 \mathbf{k}_1) = \lambda \eta^T$  and  $\eta^T \mathbf{b}_2 = 0$ . Because all the vectors that are orthogonal to  $\mathbf{b}_2$  can be expressed as a linear combination of  $\eta_1 = [1, 0, 0]^T$  and  $\eta_2 = [0, 1, 0]^T$ , controllability of  $(\mathbf{A} - \mathbf{b}_1 \mathbf{k}_1, \mathbf{b}_2)$  is assured if  $\alpha \eta_1 + \beta \eta_2$  is not a left eigenvector of  $\mathbf{A} - \mathbf{b}_1 \mathbf{k}_1$  for any  $\alpha, \beta \in \mathbb{R}$ .

Hence

$$\begin{aligned} & (\alpha \eta_1 + \beta \eta_2)^T (\mathbf{A} - \mathbf{b}_1 \mathbf{k}_1) \\ &= \alpha \begin{bmatrix} -1 & 0 & 0 \end{bmatrix} - \mathbf{k}_1 + \beta \begin{bmatrix} 2 & -2 & -3 \end{bmatrix} \\ &= \begin{bmatrix} -\alpha + 2\beta & -2\beta & -3\beta \end{bmatrix} - \mathbf{k}_1 \\ &\neq \lambda (\alpha \eta_1 + \beta \eta_2)^T \\ &= \lambda \begin{bmatrix} \alpha & \beta & 0 \end{bmatrix} \end{aligned}$$

We can choose  $\mathbf{k}_1 = [0, 1, 0]$ , since  $-2\beta - 1 = \lambda\beta$  and  $-3\beta = 0$  cannot hold simultaneously. With such a  $\mathbf{k}_1$ ,  $(\mathbf{A} - \mathbf{b}_1 \mathbf{k}_1, \mathbf{b}_2)$  is controllable.

2. (a) Suppose that  $(\mathbf{A}, \mathbf{B})$  is uncontrollable. Then by PBH test, there exist a nonzero vector  $\eta \in \mathbb{C}^{2n}$  and  $\lambda \in \mathbb{C}$  such that  $\eta^T \mathbf{A} = \lambda \eta^T$ , and  $\eta^T \mathbf{B} = \mathbf{0}$ . Let  $\eta = [\eta_1^T, \eta_2^T]^T$ , where  $\eta_1, \eta_2 \in \mathbb{C}^n$ . Then

$$\begin{aligned} \begin{bmatrix} \eta_1^T & \eta_2^T \end{bmatrix} \begin{bmatrix} \mathbf{0} \\ \mathbf{M}^{-1} \end{bmatrix} &= \eta_2^T \mathbf{M}^{-1} = 0 \Rightarrow \eta_2 = \mathbf{0} \\ \begin{bmatrix} \eta_1^T & \eta_2^T \end{bmatrix} \begin{bmatrix} \mathbf{0} & \mathbf{I} \\ -\mathbf{M}^{-1} \mathbf{K} & -\mathbf{M}^{-1} \mathbf{D} \end{bmatrix} &= \begin{bmatrix} \mathbf{0} & \eta_1^T \end{bmatrix} = \lambda \begin{bmatrix} \eta_1^T & \mathbf{0} \end{bmatrix} \Rightarrow \eta_1 = \mathbf{0} \end{aligned}$$

However, this is a contradiction because  $\eta_1$  and  $\eta_2$  cannot vanish simultaneously. Hence  $(\mathbf{A}, \mathbf{B})$  is controllable.

(b) Define  $\mathbf{x}_3 = \ddot{\mathbf{q}}$  and  $\mathbf{v} = \dot{\mathbf{u}}$ . Expand the model as follows:

$$\frac{d}{dt} \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \mathbf{x}_3 \end{bmatrix} = \begin{bmatrix} \mathbf{0} & \mathbf{I} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \\ \mathbf{0} & -\mathbf{M}^{-1}\mathbf{K} & -\mathbf{M}^{-1}\mathbf{D} \end{bmatrix} \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \mathbf{x}_3 \end{bmatrix} + \begin{bmatrix} \mathbf{0} \\ \mathbf{0} \\ \mathbf{M}^{-1} \end{bmatrix} \mathbf{v}$$

Let

$$\bar{\mathbf{A}} = \begin{bmatrix} \mathbf{0} & \mathbf{I} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \\ \mathbf{0} & -\mathbf{M}^{-1}\mathbf{K} & -\mathbf{M}^{-1}\mathbf{D} \end{bmatrix}, \quad \bar{\mathbf{B}} = \begin{bmatrix} \mathbf{0} \\ \mathbf{0} \\ \mathbf{M}^{-1} \end{bmatrix}, \quad \bar{\mathbf{x}} = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \mathbf{x}_3 \end{bmatrix}$$

Define the cost function as

$$J = \int_0^\infty (\mathbf{q}^T \mathbf{Q} \mathbf{q} + \mathbf{v}^T \mathbf{R} \mathbf{v}) dt = \int_0^\infty (\bar{\mathbf{x}}^T \bar{\mathbf{Q}} \bar{\mathbf{x}} + \mathbf{v}^T \mathbf{R} \mathbf{v}) dt, \quad \bar{\mathbf{Q}} = \begin{bmatrix} \mathbf{Q} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \end{bmatrix}$$

where  $\mathbf{Q}, \mathbf{R} \in \mathbb{R}^{n \times n}$  are symmetric positive definite. Then the optimal state feedback control law that minimizes  $J$  is

$$\mathbf{v} = -\mathbf{R}^{-1} \bar{\mathbf{B}}^T \bar{\mathbf{P}} \mathbf{x}$$

where  $\bar{\mathbf{P}}$  is the solution to the following algebraic Riccati equation (ARE)

$$\bar{\mathbf{A}}^T \bar{\mathbf{P}} + \bar{\mathbf{P}} \bar{\mathbf{A}} - \bar{\mathbf{P}} \bar{\mathbf{B}} \mathbf{R}^{-1} \bar{\mathbf{B}}^T \bar{\mathbf{P}} + \bar{\mathbf{Q}} = \mathbf{0}$$

Let  $\mathbf{R}^{-1} \bar{\mathbf{B}}^T \bar{\mathbf{P}} = \bar{\mathbf{K}} = [\mathbf{K}_1, \mathbf{K}_2, \mathbf{K}_3]$ , where  $\mathbf{K}_1, \mathbf{K}_2, \mathbf{K}_3 \in \mathbb{R}^{n \times n}$ . Then

$$\mathbf{v} = -\bar{\mathbf{K}} \bar{\mathbf{x}} = -(\mathbf{K}_1 \mathbf{x}_1 + \mathbf{K}_2 \mathbf{x}_2 + \mathbf{K}_3 \mathbf{x}_3) = -(\mathbf{K}_1 \mathbf{q} + \mathbf{K}_2 \dot{\mathbf{q}} + \mathbf{K}_3 \ddot{\mathbf{q}})$$

Thus the control law is

$$\mathbf{u}(t) = \int_0^t \mathbf{v}(\tau) d\tau = -\left( \mathbf{K}_1 \int_0^t \mathbf{q}(\tau) d\tau + \mathbf{K}_2 \mathbf{q}(t) + \mathbf{K}_3 \dot{\mathbf{q}}(t) \right)$$

Clearly, this is a PID controller with  $\mathbf{K}_P = \mathbf{K}_2$ ,  $\mathbf{K}_D = \mathbf{K}_3$ , and  $\mathbf{K}_I = \mathbf{K}_1$ .

3. Let  $\mathbf{P} = \begin{bmatrix} p_1 & p_3 \\ p_3 & p_2 \end{bmatrix}$ , where  $p_1 > 0$  and  $p_1 p_2 - p_3^2 > 0$ . The ARE  $\mathbf{A}^T \mathbf{P} + \mathbf{P} \mathbf{A} - \mathbf{P} \mathbf{B} \mathbf{R}^{-1} \mathbf{B}^T \mathbf{P} + \mathbf{Q} = \mathbf{0}$  yields

$$\begin{aligned} -p_3^2 + q &= 0 \\ p_1 - 10p_3 - p_2 p_3 &= 0 \\ p_2^2 + 20p_2 - 2p_3 &= 0 \end{aligned}$$

Hence  $p_3 = \sqrt{q}$ ,  $p_2 = -10 + \sqrt{100 + 2\sqrt{q}}$ ,  $p_1 = \sqrt{q} \sqrt{100 + 2\sqrt{q}}$ .

*Note: The solution  $p_3 = -\sqrt{q}$  leads to  $p_2 < 0$ , which violates the requirement that  $\mathbf{P} > 0$ .*

Then LQR control law

$$\mathbf{u} = -\mathbf{R}^{-1} \mathbf{B}^T \mathbf{P} \mathbf{x} = -(p_3 x_1 + p_2 x_2) = -\sqrt{q} x_1 - \left( \sqrt{100 + 2\sqrt{q}} - 10 \right) x_2$$

The closed-loop system is  $\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{B} \mathbf{K}) \mathbf{x} = \begin{bmatrix} 0 & 1 \\ -\sqrt{q} & -\sqrt{100 + 2\sqrt{q}} \end{bmatrix} \mathbf{x}$ .

Hence the closed-loop characteristic equation is

$$s^2 + \sqrt{100 + 2\sqrt{q}} s + \sqrt{q} = 0$$

4. The optimal feedback gain is given by  $\mathbf{K} = \mathbf{R}^{-1}\mathbf{B}^T\mathbf{P}$ , where  $\mathbf{P} > 0$  satisfies ARE:

$$\mathbf{A}^T\mathbf{P} + \mathbf{P}\mathbf{A} - \mathbf{P}\mathbf{B}\mathbf{R}^{-1}\mathbf{B}^T\mathbf{P} + \mathbf{Q} = 0$$

$\mathbf{Q} > 0$  and  $\mathbf{R} > 0$ . To show  $\mathbf{A} - \alpha\mathbf{B}\mathbf{K}$  is stable, it suffices to show that

$$\mathbf{Q}_\alpha = (\mathbf{A} - \alpha\mathbf{B}\mathbf{R}^{-1}\mathbf{B}^T\mathbf{P})^T\mathbf{P} + \mathbf{P}(\mathbf{A} - \alpha\mathbf{B}\mathbf{R}^{-1}\mathbf{B}^T\mathbf{P}) < 0$$

By ARE we have

$$-\mathbf{Q}_\alpha = \mathbf{Q} + (2\alpha - 1)\mathbf{P}\mathbf{B}\mathbf{R}^{-1}\mathbf{B}^T\mathbf{P}$$

Since  $\mathbf{Q} > 0$  and  $\mathbf{P}\mathbf{B}\mathbf{R}^{-1}\mathbf{B}^T\mathbf{P} \geq 0$ , we have  $-\mathbf{Q}_\alpha > 0$  for  $\alpha \geq \frac{1}{2}$ .

5. (a) Let  $x_1 = \theta$  and  $x_2 = \dot{\theta}$ . Then the state equation is

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$

- (b) Use the Matlab command `lqr` to find the optimal feedback gain  $\mathbf{K} = [1, 0.3181]$ . The states and the control input are shown in Figure 1

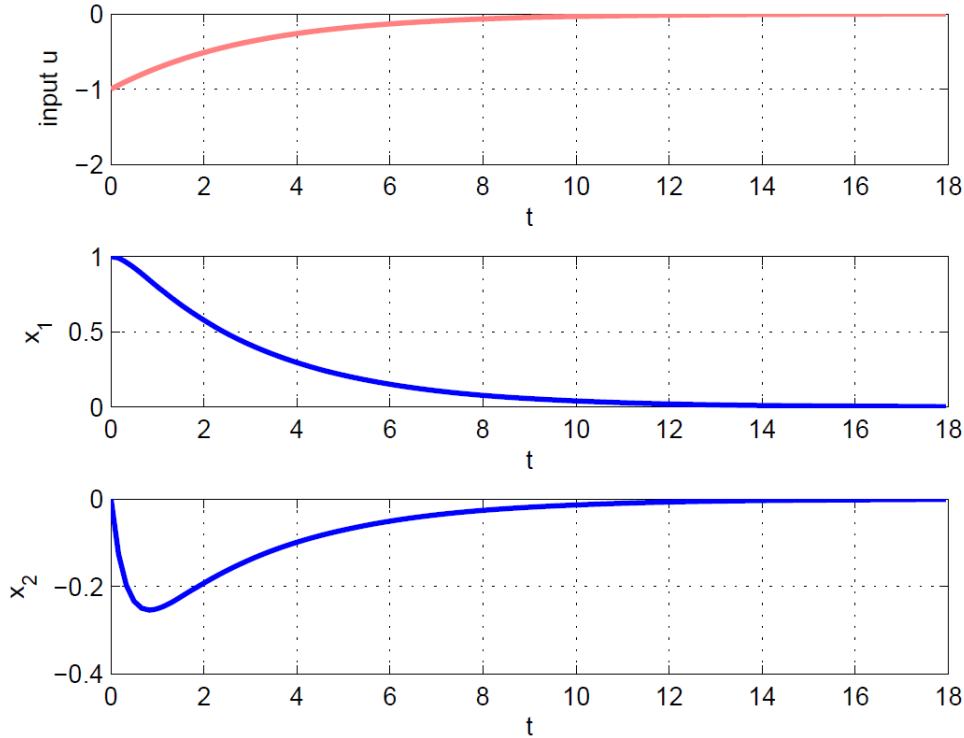


Figure 1: Control Input and States for Problem 5(b)

- (c) The discretized system is  $\mathbf{x}(k+1) = \mathbf{A}_d\mathbf{x}(k) + \mathbf{B}_d u(k)$ . Use Matlab command `c2d` to find  $\mathbf{A}_d = \begin{bmatrix} 1 & 0.009851 \\ 0 & 0.9704 \end{bmatrix}$  and  $\mathbf{B}_d = \begin{bmatrix} 4.95 \times 10^{-5} \\ 0.009851 \end{bmatrix}$ .
- (d) Solve the difference Riccati equation and find the feedback gain  $\mathbf{K}(k)$  for each time step  $k$ . The gain is shown in Figure 2. The states and the control input are shown in Figure 3.

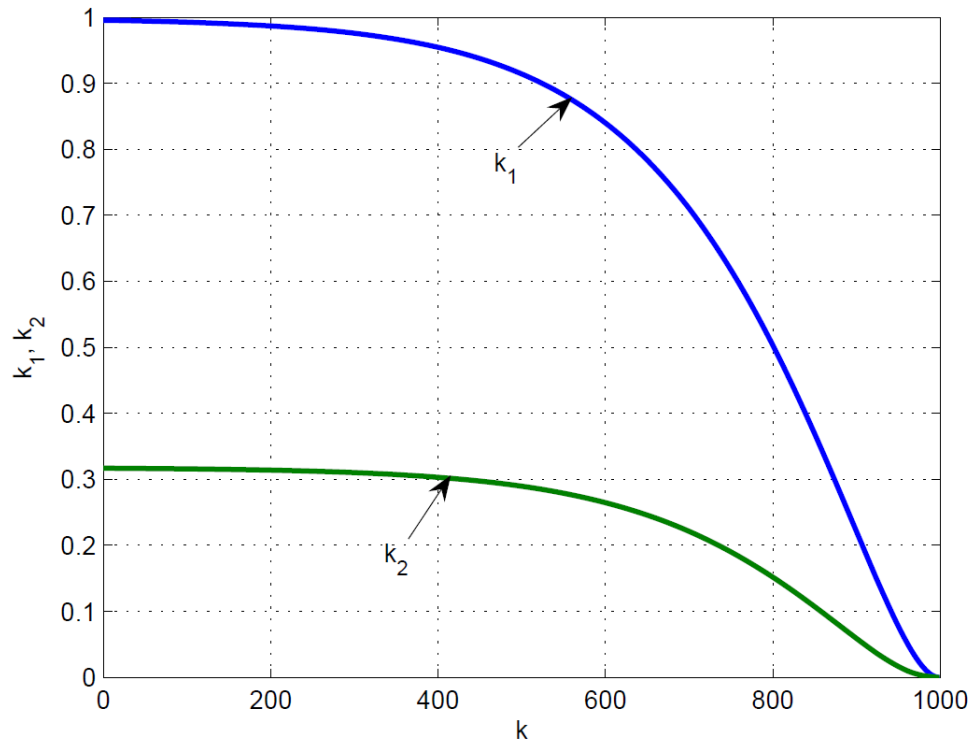


Figure 2: The feedback gain for Problem 5(d)

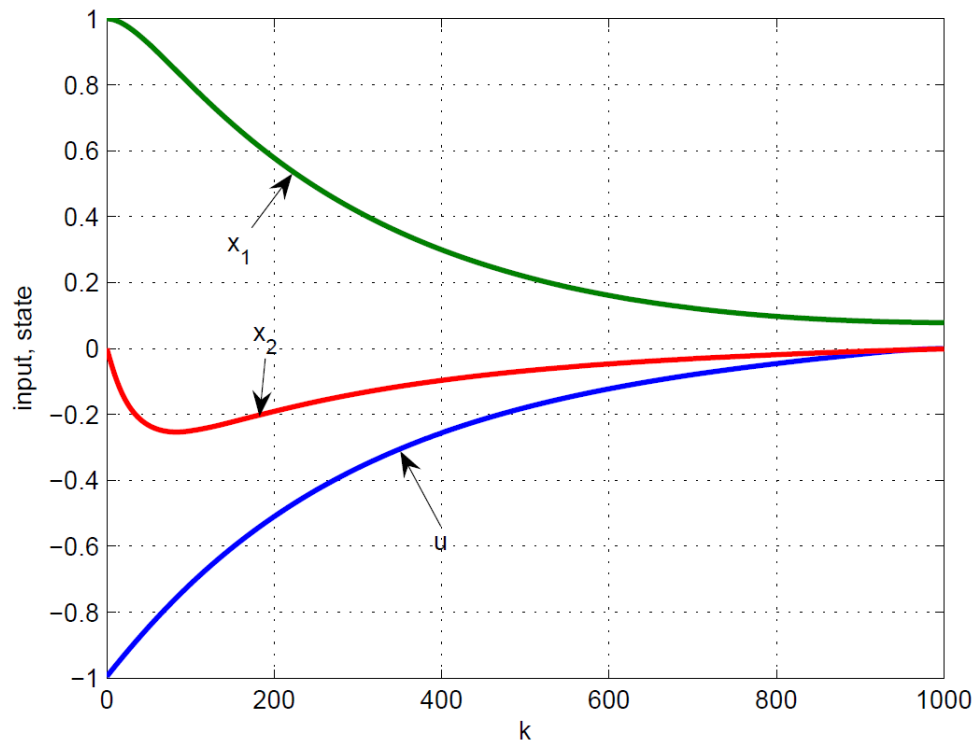


Figure 3: Control Input and States for Problem 5(d)