



In this exercise we consider the second-order system

$$\ddot{x} + k_1\dot{x} + k_2x = k_3u \quad (1)$$

In state-space form, with  $x_1 = x$  and  $x_2 = \dot{x}$ , we get

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -k_2 & -k_1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ k_3 \end{bmatrix} u \quad (2)$$

Discretizing the system using the explicit Euler scheme with sampling time  $T$  gives

$$\frac{x_{t+1} - x_t}{T} = \begin{bmatrix} 0 & 1 \\ -k_2 & -k_1 \end{bmatrix} x_t + \begin{bmatrix} 0 \\ k_3 \end{bmatrix} u_t \quad (3)$$

and hence

$$x_{t+1} = \underbrace{\begin{bmatrix} 1 & T \\ -k_2T & 1 - k_1T \end{bmatrix}}_A x_t + \underbrace{\begin{bmatrix} 0 \\ k_3T \end{bmatrix}}_B u_t \quad (4)$$

Let  $k_1 = k_2 = k_3 = 1$  and  $T = 0.1$ . The initial condition is  $x_0 = [5 \ 1]^\top$ ; the initial state estimate is  $\hat{x}_0 = [6 \ 0]^\top$  when an observer is used.

### Problem 1 (20 %) The Riccati Equation

The algebraic or stationary Riccati equation is stated as

$$P = Q + A^\top P(I + BR^{-1}B^\top P)^{-1}A \quad (5)$$

in the MPC note. Another common form of this equation is

$$A^\top PA - P - A^\top PB(R + B^\top PB)^{-1}B^\top PA + Q = 0 \quad (6)$$

(see, e.g., the MATLAB documentation for the `dlqr` function). Use the matrix inversion lemma (also known as the Sherman-Morrison-Woodbury formula)

$$(S + UTV)^{-1} = S^{-1} - S^{-1}U(T^{-1} + VS^{-1}U)^{-1}VS^{-1} \quad (7)$$

to derive (6) from (5).

## Problem 2 (30 %) LQR and State Estimation

We will in this problem assume that only  $x_1$  is measured; that is,

$$y_t = Cx_t = \begin{bmatrix} 1 & 0 \end{bmatrix} x_t \quad (8)$$

We use LQR and an observer to control the output.

**a** We want to minimize the infinite-horizon objective function

$$f^\infty(z) = \frac{1}{2} \sum_{t=0}^{\infty} \{ \hat{x}_{t+1}^\top Q \hat{x}_{t+1} + u_t^\top R u_t \} \quad (9a)$$

with

$$Q = \begin{bmatrix} 4 & 0 \\ 0 & 4 \end{bmatrix} \quad \text{and} \quad R = 1 \quad (9b)$$

Note that the objective function is formulated in  $\hat{x}_{t+1}$  (the state estimate) as opposed to  $x_{t+1}$  (the actual state). Use the MATLAB function `dlqr` to find the optimal feedback gain  $K$ , assuming that the full state is available for feedback. What is  $K$  and the resulting closed-loop poles (the eigenvalues of  $A - BK$ )?

**b** The state observer needs to be faster than the controller, meaning that the poles of  $A - K_F C$  are faster than the poles of  $A - BK$ . Use the MATLAB function `place` to place the observer poles. You can chose the poles yourself or use the poles  $p_{1,2} = 0.5 \pm 0.03j$  (in the  $z$ -plane); these poles correspond to a time constant of approximately  $1/5$  of the fastest control time constant. Simulate the system for 50 time steps with feedback from the estimator and plot both  $x_t$  and  $\hat{x}_t$ . Comment on the performance and tune the controller and/or the observer if you wish to improve the performance.

**c** The control and estimation equations can be written

$$\xi_{t+1} = \begin{bmatrix} x_{t+1} \\ \tilde{x}_{t+1} \end{bmatrix} = \underbrace{\begin{bmatrix} A - BK & BK \\ 0 & A - K_F C \end{bmatrix}}_{\Phi} \xi_t \quad (10a)$$

$$\tilde{x}_t = x_t - \hat{x}_t \quad (10b)$$

State the full matrix  $\Phi$  with your numerical values and verify that the eigenvalues of  $\Phi$  are the poles of  $A - BK$  and  $A - K_F C$ .

## Problem 3 (30 %) MPC and State Estimation

We now add the input constraint

$$-4 \leq u_t \leq 4 \quad t = 1, \dots, N-1 \quad (11)$$

and use MPC with  $Q$  and  $R$  as given in (9b).

- a** Modify your code and use output-feedback MPC and the observer you designed in Problem 2 to control the system. The output  $y_t$  is the same as in the previous problem. Let the MPC minimize the open-loop objective function

$$f(z) = \frac{1}{2} \sum_{t=0}^{N-1} \{\hat{x}_{t+1}^\top Q \hat{x}_{t+1} + u_t^\top R u_t\} \quad (12)$$

at every time instant with  $N = 10$ . Tune the controller if necessary. Simulate the closed-loop system for 50 time steps.

- b** We now assume that both states are available for feedback; that is,  $C = I$ . Repeat problem 1) with state feedback (do not use the observer). This means the open-loop objective function is

$$f(z) = \frac{1}{2} \sum_{t=0}^{N-1} \{x_{t+1}^\top Q x_{t+1} + u_t^\top R u_t\} \quad (13)$$

Compare the closed-loop response to what you obtained in 1) and comment.

#### Problem 4 (20 %) Infinite-Horizon MPC

- a** Calculate the Riccati matrix  $P$  using the MATLAB function `dlqr`.
- b** Modify your code from Problem 3 2) and minimize the open-loop objective function

$$f(z) = \frac{1}{2} \sum_{t=0}^{N-1} \{x_{t+1}^\top Q x_{t+1} + u_t^\top R u_t\} + x_N^\top P x_N \quad (14)$$

This can be done by modifying  $G$  in the formulation

$$f(z) = \frac{1}{2} z^\top G z \quad (15)$$

Specifically, the last  $Q$  on the diagonal of  $G$  must be replaced by  $P$ . Use  $N = 10$  and compare the closed-loop response with what you obtained in Problem 3 2) and comment. Change  $N$  and look at the open-loop solutions. Are the input constraints always inactive toward the end of the horizon? When does  $N$  become important for performance?