

## FRTN10 Exercise 10. LQG, Preparations for Lab 3

**10.1**  Consider the system

$$\begin{aligned}\dot{x} &= \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} x + \begin{pmatrix} 1 & 6 \\ 0 & 4 \end{pmatrix} u + w_1 \\ y &= (1 \quad 1) x + w_2\end{aligned}$$

where  $w_1$  and  $w_2$  are independent white noise processes. You should design an LQG controller that minimizes the following cost function:

$$J = \mathbf{E} \left( (x_1 + x_2)^2 + u^T u \right)$$

- a.** Design the LQG controller in Matlab, initially assuming that the process and measurement noise have unit intensities.

*Useful commands:* lqr, kalman, lqgreg

- b.** Using the states  $x$  and  $\hat{x}$ , write the closed-loop system in state-space form using symbols. Use  $L$  for the state-feedback gain and  $K$  for the Kalman filter gain.

- c.** Simulate the system without noise from the initial state  $x = (1 \quad -1)^T$ . Plot both process states and estimated states. The Kalman filter begins with its estimates in 0. Try some different values of  $R_2$  (much smaller or much larger than the initial value) and repeat the design and simulation. What conclusions can you draw?

*Useful commands:* lqgreg, feedback, initial

**10.2** Do the three preparatory exercises for Laboratory Session 3. The lab manual is found on the course homepage.

**10.3(\*)** Consider the problem of controlling a double integrator

$$\dot{x} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} x + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u + v_1$$

where the white noise  $v_1$  has intensity  $I$ . We can only measure  $x_1$ , unfortunately with added white noise also of intensity 1. We want to minimize the cost function

$$J = \mathbf{E} (x_1^2 + x_2^2 + u^2)$$

Solve the control design problem by hand and give the LQG controller in state-space form.

## Solutions to Exercise 10. LQG, Preparations for Lab 3

10.1 a. See the Matlab code in **c** below.

- b. Using the state vector  $x_e = (x^T \hat{x}^T)^T$  and the obvious notation  $A, B, C$ , we get the system

$$\begin{aligned} \dot{x}_e &= \begin{pmatrix} A & -BL \\ KC & A - BL - KC \end{pmatrix} x_e + \begin{pmatrix} I \\ 0 \end{pmatrix} w_1 + \begin{pmatrix} 0 \\ K \end{pmatrix} w_2 \\ z &= (C \ 0) x_e \end{aligned}$$

- c. With smaller  $R_2$  the estimated states converge faster to the actual states, and the controlled output  $z$  converges faster to zero. The opposite holds for larger  $R_2$ . See Figures 10.1-10.2 and Matlab code below.

As shown in Exercise 9.1, only the relation between process noise and measurement noise matters. Increasing  $R_1$  by some factor or decreasing  $R_2$  by the same factor will thus have the same effect.

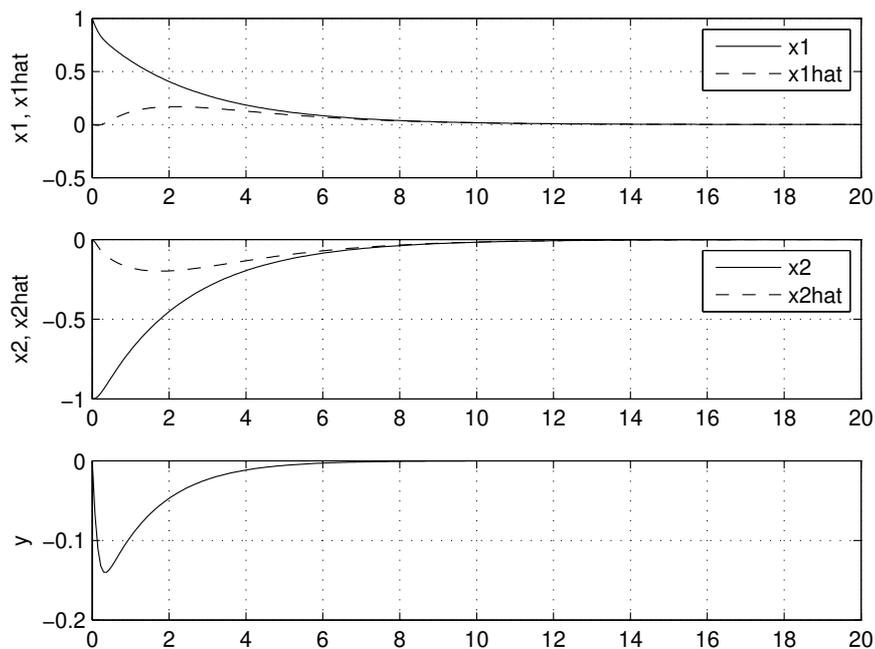
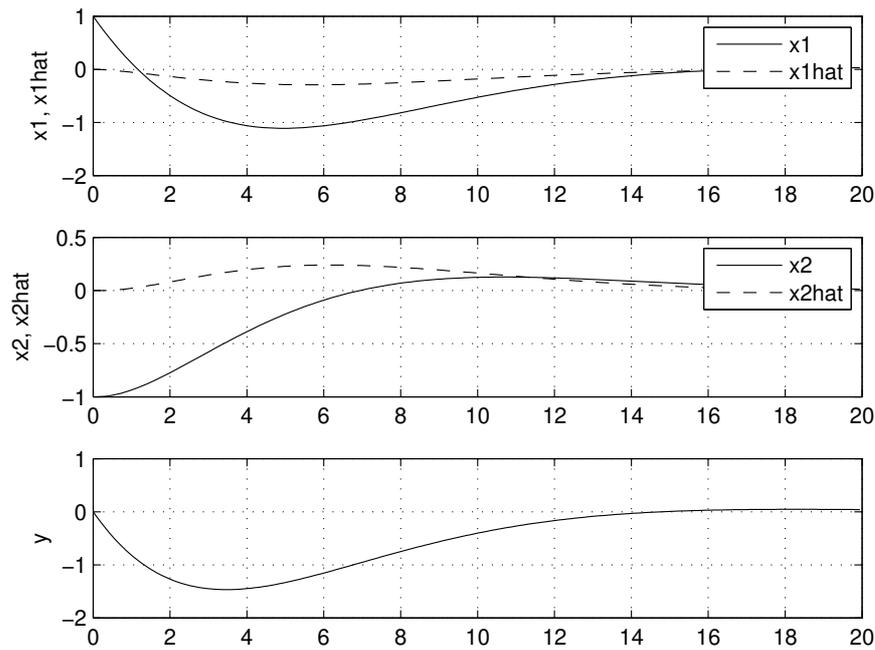


Figure 10.1 Initial response if  $R_2$  10 times smaller.

```
A = [0 1; 0 0];
B = [1 6; 0 4];
C = [1 1];

% LQ design
process = ss(A,B,C,0);
Q1 = C'*C;
Q2 = eye(2);
```



**Figure 10.2** Initial response if  $R_2$  100 times larger).

```

L = lqr(process,Q1,Q2);

% Kalman filter design
G = eye(2);
sysk = ss(A,[B G],C,0);
R1 = eye(2);
R2 = 1;
Kest = kalman(sysk,R1,R2);

% Construct regulator and form closed loop
reg = lqgreg(Kest,L);
closed_loop = feedback(process,-reg);

% Plot response
[Y,T,X] = initial(closed_loop,[1 -1 0 0],0:0.01:20);
subplot(311)
plot(T, X(:,1)); hold on; plot(T, X(:,3),'--'); grid
legend('x1','x1hat'); ylabel('x1, x1hat')
subplot(312)
plot(T, X(:,2)); hold on; plot(T, X(:,4),'--'); grid
legend('x2','x2hat'); ylabel('x2, x2hat')
subplot(313)
plot(T,Y); grid; ylabel('y');

```

**10.2** No solutions provided.

**10.3** To solve the problem we need to design an LQ state feedback and a Kalman

filter. The problem parameters are given by

$$A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \quad C = (1 \ 0),$$

$$Q_1 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \quad Q_2 = 1, \quad R_1 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \quad R_2 = 1$$

For the LQ state feedback gain, we have to solve the Riccati equation

$$A^T S + SA + Q_1 - SBQ_2^{-1}B^T S = 0$$

with

$$S = \begin{pmatrix} s_1 & s_2 \\ s_2 & s_3 \end{pmatrix}$$

This gives the following equations,

$$\begin{aligned} 1 - s_2^2 &= 0 \\ s_1 - s_2 s_3 &= 0 \\ 2s_2 + 1 - s_3^2 &= 0 \end{aligned}$$

with the positive definite solution  $s_1 = s_3 = \sqrt{3}$ ,  $s_2 = 1$ . This gives the state feedback vector  $L = B^T S = (1 \ \sqrt{3})$ .

For the Kalman filter we must solve the Riccati equation

$$AP + PA^T + R_1 - PC^T C P = 0$$

with

$$P = \begin{pmatrix} p_1 & p_2 \\ p_2 & p_3 \end{pmatrix}$$

yielding the equations

$$\begin{aligned} 2p_2 + 1 - p_1^2 &= 0 \\ p_3 - p_1 p_2 &= 0 \\ 1 - p_2^2 &= 0 \end{aligned}$$

Reusing the solution for  $S$  we have that  $p_1 = p_3 = \sqrt{3}$  and  $p_2 = 1$  and  $K = PC^T = \begin{pmatrix} \sqrt{3} \\ 1 \end{pmatrix}$

The LQG controller is given by

$$\begin{aligned} \dot{\hat{x}} &= (A - BL - KC)\hat{x} + Ky \\ u &= -L\hat{x} \end{aligned}$$

and we have that

$$A - BL - KC = \begin{pmatrix} -\sqrt{3} & 1 \\ -2 & -\sqrt{3} \end{pmatrix}$$