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## Model-based Estimation Methods, SS 2016

Solution - Tutorial 10 (July 06, 2016)

# State Estimation

## Solution of Problem 16

a.) The states of the system are  $h_1$  and  $h_2$ .

The inputs of the system is  $F_{in}$ .

The parameters of the system are  $R_1$ ,  $R_2$ ,  $A_1$  and  $A_2$ .

The output of the system is  $h_2$ .

b.) In LTI form the model equations of the two tank system can be rewritten as

$$\begin{bmatrix} \frac{dh_1}{dt} \\ \frac{dh_2}{dt} \end{bmatrix} = \begin{bmatrix} \frac{-1}{R_1 \cdot A_1} & 0 \\ \frac{1}{R_1 \cdot A_1} & \frac{-1}{R_2 \cdot A_2} \end{bmatrix} \cdot \begin{bmatrix} h_1 \\ h_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} \cdot F_{in}$$

$$h_2 = \begin{bmatrix} 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} h_1 \\ h_2 \end{bmatrix}$$

#### Solution of Bonus Problem 10

The error at instant i + 1 can be written as

$$e[i+1] = x[i+1] - \hat{x}[i+1]$$

$$= A_d x[i] + B_d u[i] - (A_d \hat{x}[i] + B_d u[i] + L_d (C_d x[i] - C_d \hat{x}[i]))$$

$$= (A_d - L_d C_d)(x[i] - \hat{x}[i])$$

$$= (A_d - L_d C_d)e[i]$$

In order for the error to decay with time the eigenvalues of the matrix  $(A_d - L_dC_d)$  must lie within the unit circle.

#### Solution of Bonus Problem 11

1.  $P^+[k]$  is defined as

$$P^{+}[k] = E[(x[k] - \hat{x}^{+}[k])(x[k] - \hat{x}^{+}[k])']$$

Substituting 
$$\hat{x}^{+}[k] = \hat{x}^{-}[k] + K[k](\tilde{y} - C_d\hat{x}^{-}[k] - D_du[k])$$

$$P^{+}[k] = E[(x[k] - (\hat{x}^{-}[k] + K[k](\tilde{y} - C_d\hat{x}^{-}[k] - D_du[k])))$$
$$(x[k] - (\hat{x}^{-}[k] + K[k](\tilde{y} - C_d\hat{x}^{-}[k] - D_du[k])))']$$

Since 
$$\tilde{y}[k] = C_d x[k] + D_d u[k] + v[k]$$

$$P^{+}[k] = E[(x[k] - (\hat{x}^{-}[k] + K[k](C_{d}x[k] - C_{d}\hat{x}^{-}[k] + v[k])))$$

$$(x[k] - (\hat{x}^{-}[k] + K[k](C_{d}x[k] - C_{d}\hat{x}^{-}[k] + v[k])))']$$

$$= E[((I - K[k]C_{d})(x[k] - \hat{x}^{-}[k]) - K[k]v[k])$$

$$((I - K[k]C_{d})(x[k] - \hat{x}^{-}[k]) - K[k]v[k])']$$

$$= E[(I - K[k]C_{d})(x[k] - \hat{x}^{-}[k])(x[k] - \hat{x}^{-}[k])'(I - K[k]C_{d})']$$

$$-E[(I - K[k]C_{d})(x[k] - \hat{x}^{-}[k])v'[k]K'[k]] - E[K[k]v[k](x[k] - \hat{x}^{-}[k])'(I - K[k]C_{d})']$$

$$+E[K[k]v[k]v'[k]K'[k]]$$

Since K[k] and  $C_d$  are constants and using the property

$$E[kx] = kE[x] \tag{1}$$

Where

$$x \equiv \text{Stochastic variable}$$

$$k \equiv \text{Constant}$$

We now have,

$$P^{+}[k] = (I - K[k]C_{d})E[(x[k] - \hat{x}^{-}[k])(x[k] - \hat{x}^{-}[k])'](I - K[k]C_{d})'$$

$$-(I - K[k]C_{d})E[(x[k] - \hat{x}^{-}[k])v'[k]]K'[k] - K[k]E[v[k](x[k] - \hat{x}^{-}[k])'](I - K[k]C_{d})'$$

$$+K[k]E[v[k]v'[k]]K'[k]$$

We know,

$$E[(x[k] - \hat{x}^-[k])(x[k] - \hat{x}^-[k])'] = P^-[k]$$

$$E[(x[k] - \hat{x}^-[k])v'[k]] = 0 \qquad \text{(Since } x[k] \text{ and } v[k] \text{ are uncorrelated)}$$

$$E[v[k]v'[k]] = R$$

Therefore,

$$P^{+}[k] = (I - K[k]C_d)P^{-}[k](I - K[k]C_d)' + KRK'$$

2. According to the first order necessary condition for optimality, we have

$$\frac{d(trace(P^+[k]))}{dK[k]} = 0$$

Therefore,

$$0 = trace((I - K[k]C_d)P^{-}[k](-C_d)' + (-C_d)P^{-}[k](I - K[k]C_d)' + KR + RK')$$

Since the trace of a matrix product AB is independent of the order of A and B, we can write the equation above as

$$0 = trace(2(I - K[k]C_d)P^{-}[k](-C_d)' + 2KR)$$
  
$$0 = trace(-P^{-}[k]C_d' + K[k]C_dP^{-}[k]C_d' + KR)$$

$$0 = trace(-P^{-}[k]C'_d + K[k](C_dP^{-}[k]C'_d + R))$$

Therefore, the Kalman gain matrix K[k] which minimizes the  $trace(P^+[k])$  is given by

$$K[k] = P^{-}[k]C'_{d}(C_{d}P^{-}[k]C'_{d} + R)^{-1}$$

3.

$$P^{-}[k+1] = E[(x[k+1] - \hat{x}^{-}[k+1])(x[k+1] - \hat{x}^{-}[k+1])']$$

Substituting  $x[k+1] = A_d x[k] + B_d u[k] + w[k]$  and  $\hat{x}^-[k+1] = A_d \hat{x}^+[k] + B_d u[k]$ , we now have

$$P^{-}[k+1] = E[(A_dx[k] + w[k] - A_d\hat{x}^{+}[k])(A_dx[k] + w[k] - A_d\hat{x}^{+}[k])']$$
$$= E[(A_d(x[k] - \hat{x}^{+}[k]) + w[k])(A_d(x[k] - \hat{x}^{+}[k]) + w[k])']$$

Making use of the property in Eq.1, we have

$$P^{-}[k+1] = A_{d}E[(x[k] - \hat{x}^{+}[k])(x[k] - \hat{x}^{+}[k])']A'_{d}$$
$$+A_{d}E[(x[k] - \hat{x}^{+}[k])w'[k]] + E[w[k](x[k] - \hat{x}^{+}[k])']A'_{d}$$
$$+E[w[k]w'[k]]$$

We know,

$$E[(x[k] - \hat{x}^+[k])(x[k] - \hat{x}^+[k])'] = P^+[k]$$

$$E[(x[k] - \hat{x}^-[k])w'[k]] = 0 \qquad \text{(Since } x[k] \text{ and } w[k] \text{ are uncorrelated)}$$

$$E[w[k]w'[k]] = Q$$

Therefore,

$$P^{-}[k+1] = A_d P^{+}[k] A_d' + Q$$