

1.

$$\min_{u_0} \{ J \} = \min_{u_0} [y_d(N) - y(N)]^T \bar{Q}_f [y_d(N) - y(N)] + \sum_{k=0}^{N-1} [y_d(k) - y(k)]^T \bar{Q} [y_d(k) - y(k)] + u^T(k) R u(k) \}$$

where $u_k = \{ u(k), u(k+1), \dots, u(N-1) \}$

when $k \in [0, N-1]$ we can define

$$J_k^0[x(k)] = \min_{u_k} [y_d(N) - y(N)]^T \bar{Q}_f [y_d(N) - y(N)] + \sum_{i=k}^{N-1} [y_d(i) - y(i)]^T \bar{Q} [y_d(i) - y(i)] + u^T(i) R u(i) \}$$

especially, for $k=N$

$$J_N^0[x(N)] = [y_d(N) - y(N)]^T \bar{Q}_f [y_d(N) - y(N)]$$

since $y_d(k)$ is specified for all k , and $y(k) = Cx(k)$

we have

$$\begin{aligned} J_N^0[x(N)] &= [y_d(N) - Cx(N)]^T \bar{Q}_f [y_d(N) - Cx(N)] \\ &= x^T(N) C^T \bar{Q}_f C x(N) + y_d^T(N) \bar{Q}_f y_d(N) \\ &\quad - y_d^T(N) \bar{Q}_f C x(N) - x^T(N) C^T \bar{Q}_f y_d(N) \end{aligned}$$

$$\text{since } y_d^T(N) \bar{Q}_f C x(N) = x^T(N) C^T \bar{Q}_f y_d(N)$$

$$\text{then } J_N^0[x(N)] = x^T(N) C^T \bar{Q}_f C x(N) + y_d^T(N) \bar{Q}_f y_d(N) - 2 x^T(N) C^T \bar{Q}_f y_d(N)$$

we can define $P(N) = C^T \bar{Q}_f C$, $b(N) = -2 C^T \bar{Q}_f y_d(N)$, $c(N) = y_d^T(N) \bar{Q}_f y_d(N)$

so that $J_N^0[x(N)]$ can be simplified as below

$$J_N^0[x(N)] = x^T(N) P(N) x(N) + x^T(N) b(N) + c(N)$$

With Bellman's principle, we can obtain $J_{k-1}^0[x(k-1)]$ from $J_k^0[x(k)]$

Assume $J_k^0[x(k)]$ has the form that

$$J_k^0[x(k)] = x^T(k) P(k) x(k) + x^T(k) b(k) + c(k)$$

since $x(k) = Ax(k-1) + Bu(k-1)$

$$\begin{aligned} J^0[x(k)] &= [Ax(k-1) + Bu(k-1)]^T P(k) [Ax(k-1) + Bu(k-1)] \\ &\quad + [Ax(k-1) + Bu(k-1)]^T b(k) + c(k) \\ &= x^T(k-1) A^T P(k) A x(k-1) + u^T(k-1) B^T P(k) B u(k-1) \\ &\quad + 2u^T(k-1) B^T P(k) A x(k-1) + x^T(k-1) A^T b(k) + u^T(k-1) B^T b(k) + c(k) \end{aligned}$$

then

$$\begin{aligned} J_{k-1}^0[x(k-1)] &= \min_{u(k-1)} \left\{ [y_d(k-1) - y(k-1)]^T \bar{Q} [y_d(k-1) - y(k-1)] \right. \\ &\quad \left. + u^T(k-1) R u(k-1) + J_k^0[x(k)] \right\} \\ &= \min_{u(k-1)} \left\{ x^T(k-1) C^T \bar{Q} C x(k-1) + y_d^T(k-1) \bar{Q} y_d(k-1) \right. \\ &\quad \left. - y_d^T(k-1) \bar{Q} C x(k-1) - x^T(k-1) C^T \bar{Q} y_d(k-1) + u^T(k-1) R u(k-1) \right. \\ &\quad \left. + x^T(k-1) A^T P(k) A x(k-1) + u^T(k-1) B^T P(k) B u(k-1) \right. \\ &\quad \left. + 2u^T(k-1) B^T P(k) A x(k-1) + x^T(k-1) A^T b(k) + u^T(k-1) B^T b(k) + c(k) \right\} \\ &= \min_{u(k-1)} \left\{ x^T(k-1) [C^T \bar{Q} C + A^T P(k) A] x(k-1) \right. \\ &\quad \left. + x^T(k-1) [A^T b(k) - 2C^T \bar{Q} y_d(k-1)] \right. \\ &\quad \left. + u^T(k-1) [R + B^T P(k) B] u(k-1) \right. \\ &\quad \left. + u^T(k-1) B^T [2P(k) A x(k-1) + b(k)] \right. \\ &\quad \left. + y_d^T(k-1) \bar{Q} y_d(k-1) + c(k) \right\} \end{aligned}$$

Then, we can take the partial derivative of the term, and set it to 0.

i.e. $[R + B^T P(k) B] u^0(k-1) + \frac{1}{2} B^T [2P(k) A x(k-1) + b(k)] = 0$

then $u^0(k-1) = -[R + B^T P(k) B]^{-1} B^T [P(k) A x(k-1) + \frac{1}{2} b(k)]$

bring the $u^0(k-1)$ back to the $J_{k-1}^0[x(k-1)]$, we have

$$\begin{aligned} J_{k-1}^0[x(k-1)] &= x^T(k-1) \left\{ C^T \bar{Q} C + A^T P(k) A - A^T P(k) B [R + B^T P(k) B]^{-1} B^T P(k) A \right\} x(k-1) \\ &\quad + x^T(k-1) \left\{ A^T b(k) - 2C^T \bar{Q} y_d(k-1) - A^T P(k) B [R + B^T P(k) B]^{-1} B^T b(k) \right\} \\ &\quad + \left\{ y_d^T(k-1) \bar{Q} y_d(k-1) + c(k) - \frac{1}{4} b^T(k) B [R + B^T P(k) B]^{-1} B^T b(k) \right\} \end{aligned}$$

so, we have $P(k-1) = C^T Q C + A^T P(k) A - A^T P(k) B [R + B^T P(k) B]^{-1} B^T P(k) A$

$$b(k-1) = A^T b(k) - 2C^T \bar{Q} y_d(k-1) - A^T P(k) B [R + B^T P(k) B]^{-1} B^T b(k)$$

$$c(k-1) = y_d^T(k-1) \bar{Q} y_d(k-1) + c(k) - \frac{1}{2} b^T(k) B [R + B^T P(k) B]^{-1} B^T b(k)$$

and $J_{k-1}^0[x(k-1)] = x^T(k-1) P(k-1) x(k-1) + x^T(k-1) b(k-1) + c(k-1)$

2.

define $\bar{x}(k) = [x(k) \quad x_w(k)]^T$ $\bar{A} = \begin{bmatrix} A & B_w C_w \\ 0 & A_w \end{bmatrix}$

$$\bar{B} = [B \quad 0]^T \quad \bar{B}_\eta = [0 \quad B_\eta]^T$$

then the augmented state equation can be written as

$$\bar{x}(k+1) = \bar{A} \bar{x}(k) + \bar{B} u(k) + \bar{B}_\eta \eta(k)$$

then we have

$$E\{\bar{x}(0)\} = \bar{x}_0 = [x_0 \quad x_{w0}]^T$$

$$E\{(\bar{x}(0) - \bar{x}_0)(\bar{x}(0) - \bar{x}_0)^T\} = \bar{X}_0 = \begin{bmatrix} X_0 & 0 \\ 0 & X_{w0} \end{bmatrix}$$

define $\bar{Q} = \begin{bmatrix} Q & 0 \\ 0 & 0 \end{bmatrix}$, $\bar{Q}_f = \begin{bmatrix} Q_f & 0 \\ 0 & 0 \end{bmatrix}$

then $J = E\left\{\bar{x}^T(N) \bar{Q}_f \bar{x}(N) + \sum_{k=0}^{N-1} [\bar{x}^T(k) \bar{Q} \bar{x}(k) + u^T(k) R u(k)]\right\}$

(a) if $u(k)$ is allowed to be a function of both $x(0), \dots, x(k)$, $x_w(0), \dots, x_w(k)$

then $u(k)$ can be a function of $\bar{x}(k)$

so from previous study, we have $u(k) = -k(k+1) \bar{x}(k)$

$$k(k) = (\bar{B}^T P(k) \bar{B} + R)^{-1} \bar{B}^T P(k) \bar{A}$$

$$P(N) = \bar{Q}_f$$

$$P(k-1) = \bar{A}^T P(k) \bar{A} + \bar{Q} - \bar{A}^T P(k) \bar{B} (\bar{B}^T P(k) \bar{B} + R)^{-1} \bar{B}^T P(k) \bar{A}$$

(b) if we can not measure the state, and only have access to the output

$$\text{since } y(k) = Cx(k) + v(k)$$

$$\text{we can let } \bar{C} = [C \ 0]$$

$$\text{then } y(k) = \bar{C} \bar{x}(k) + v(k)$$

then we can have the kalman filter

$$\hat{\bar{x}}^o(k+1) = \bar{A} \hat{\bar{x}}(k) + \bar{B} u(k)$$

$$\hat{\bar{x}}(k) = \hat{\bar{x}}^o(k) + F(k) [y(k) - \bar{C} \hat{\bar{x}}^o(k)]$$

$$\text{where, } F(k) = M(k) \bar{C}^T [\bar{C} M(k) \bar{C}^T + V]^{-1}$$

$$M(k+1) = \bar{A} Z(k) \bar{A}^T + \bar{B}_\eta \bar{\Gamma} \bar{B}_\eta^T$$

$$Z(k) = M(k) - M(k) \bar{C}^T [\bar{C} M(k) \bar{C}^T + V]^{-1} \bar{C} M(k)$$

$$M(0) = \bar{X}_0$$

then we use the state estimation to have $u(k)$

$$u(k) = -K(k+1) \hat{\bar{x}}(k)$$

$$K(k) = (\bar{B}^T P(k) \bar{B} + R)^{-1} \bar{B}^T P(k) \bar{A}$$

$$P(N) = \bar{Q}_f$$

$$P(k-1) = \bar{A}^T P(k) \bar{A} + \bar{Q} - \bar{A}^T P(k) \bar{B} (\bar{B}^T P(k) \bar{B} + R)^{-1} \bar{B}^T P(k) \bar{A}$$