

15.10.2019

ELL 333

$$\frac{x(t+h) - x(t)}{h} = Ax(t) + Bu(t) + v(t)$$

blue A ≠ grad A

Discrete-Time Kalman Filter

System

$$x_{k+1} = Ax_k + Bu_k + v_k$$

$$y_k = Cx_k + w_k$$

$$\dot{x} = Ax + Bu + v$$

$$y = Cx + w$$

$$t \in \mathbb{R}$$

$$k \in \mathbb{Z}$$

v_k, w_k are noise terms

$$E\{v_k\} = 0$$

$$E\{v_k w_k\} = 0$$

$$E\{w_k\} = 0$$

$$E\{v_k v_j'\} = \begin{cases} 0, & k \neq j \\ R_v, & k = j \end{cases}$$

$$E\{w_k w_j'\} = \begin{cases} 0, & k \neq j \\ R_w, & k = j \end{cases}$$

Observer

$$\hat{x}_{k+1} = A\hat{x}_k + Bu_k + L_k(y_k - C\hat{x}_k)$$

this can be time-varying

Error $e_k = x_k - \hat{x}_k$

Error dynamics

$$e_{k+1} = x_{k+1} - \hat{x}_{k+1}$$

$$= Ax_k + Bu_k + v_k - A\hat{x}_k - Bu_k - L_k (Cx_k + w_k - C\hat{x}_k)$$

$$= A(x_k - \hat{x}_k) + v_k - L_k w_k - L_k (x_k - \hat{x}_k)$$

$$e_{k+1} = (A - L_k C) e_k + v_k - L_k w_k$$

$$\Rightarrow E\{e_{k+1}\} = (A - L_k C) E\{e_k\}$$

What are our objectives with regard to error?

- error $\rightarrow 0$ (same as before)

- additionally, we want error covariance $E\{e_k e_k'\}$ to be as small as possible.

\hookrightarrow so, we choose L_k to minimize $\left. \begin{matrix} \{ \\ \downarrow \\ P_k \end{matrix} \right\}$ at each timestep.

$$P_k = E\{e_k e_k'\} = E\{(x_k - \hat{x}_k)(x_k - \hat{x}_k)'\}$$

\hookrightarrow How does this propagate?

Equation for propagation of variance.

$$P_{k+1} = E \{ e_{k+1} e_{k+1}' \}$$

$$= E \{ (A - L_k C) e_k + v_k - L_k w_k \} \\ \{ (A - L_k C) e_k + v_k - L_k w_k \}' \}$$

$$= E \{ (A - L_k C) e_k + v_k - L_k w_k \} \\ \{ e_k' (A - L_k C)' + v_k' - w_k' L_k' \}$$

$$= E \{ (A - L_k C) e_k e_k' (A - L_k C)' \\ + v_k v_k' \\ + L_k w_k w_k' L_k' \\ + () e_k v_k' + () v_k e_k' \\ + () e_k w_k' + () w_k e_k' \\ + () \cancel{v_k w_k'} + () \cancel{v_k' w_k} \}$$

$E \{ e_k v_k' \} = 0$
 $\hookrightarrow v_{k-1}, v_{k-2}, \dots, v_1$
 $+ e_0$ assumed independent of v_k

$\swarrow = 0$ $\swarrow = 0$

$$= (A - L_k C) E \{ e_k e_k' \} (A - L_k C)' \\ + E \{ v_k v_k' \} + L_k E \{ w_k w_k' \} L_k'$$

$\hookrightarrow P_k$
 $\hookrightarrow R_v$ $\hookrightarrow R_w$

$$= (A - L_k C) P_k (A - L_k C)' + R_v + L_k R_w L_k'$$

$$\therefore P_{k+1} = (A - L_k C) P_k (A - L_k C)' + R_v + L_k R_w L_k'$$

To minimize RHS by choosing L_k

scalar version

$$f(l) = (a - lc)^2 P + R_v + l^2 R_w$$

What l minimizes f ?

$$= Pa^2 - 2acPl + Pc^2 l^2 + R_v + l^2 R_w$$

$$= l^2 (R_w + Pc^2) - 2acPl + R_v + Pa^2$$

$$= (R_w + Pc^2) \left(l^2 - 2 \frac{acP}{(R_w + Pc^2)} l \right) + R_v + Pa^2$$

$$= (R_w + Pc^2) \left(l^2 - 2 \cdot \frac{acP}{(R_w + Pc^2)} l + \left(\frac{acP}{(R_w + Pc^2)} \right)^2 - \left(\frac{acP}{(R_w + Pc^2)} \right)^2 \right) + R_v + Pa^2$$

$$+ R_v + Pa^2$$

$$= (R_w + Pc^2) \left(l - \frac{acP}{R_w + Pc^2} \right)^2$$

$$+ R_v + Pa^2 - \frac{a^2 c^2 P^2}{R_w + Pc^2}$$

$$\therefore l = \frac{acP}{R_w + Pc^2} \text{ minimizes } f(l)$$