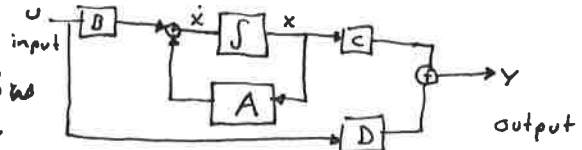


Kalman Filter

Estimate the states of a system given known inputs and measurements

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

$$y = Cx + Du + \underbrace{v}_{\text{noise!}}$$



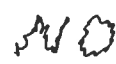

Given y and u , can we determine x ? optimally?

• Try simple algebra.

$$\text{If } y = Cx + Du \Rightarrow Cx = y - Du \Rightarrow x = C^{-1}y - C^{-1}Du$$

What is fragile about this estimate?

• Both the measured output y and the input u could be noisy signals or could just be noisy actual values

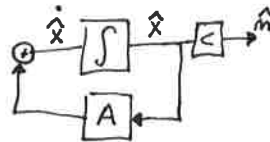
• How can we estimate x in the presence of ~~noise~~  ~~noise~~  ?

The same way you read the letters, 1) Smoothing 2) Fitting to expected shapes.

• Kalman Filter

• Operate a state machine that seeks to optimally track outputs through an error correcting update.

• Without inputs, we could mimic the actual system with ^{state} estimate \hat{x}

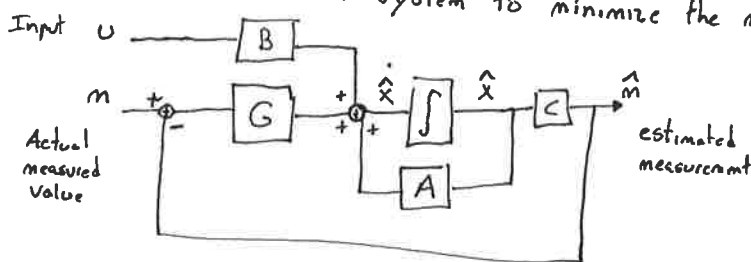


Need to know the system! $[A]$

$$\dot{\hat{x}} = A\hat{x}$$

$$\hat{m} = C\hat{x}$$

• Add a control system to minimize the measurement error $(m - \hat{m})$



This is a Kalman Filter

$$\dot{\hat{x}} = A\hat{x} + Bu + G(m - C\hat{x})$$

Error:

$$e = x - \hat{x} \equiv \text{actual value} - \text{estimated value of states}$$

$$\dot{e} = \dot{x} - \dot{\hat{x}}$$

$$= Ax + Bu \overset{+Bw}{\leftarrow} - A\hat{x} - Bu - G(m - C\hat{x})$$

$$= Ax + Bu \overset{+Bw}{\leftarrow} - A\hat{x} - Bu - G(Cx + Du + v) - C\hat{x}$$

~~noise~~ why? noisy signals

$$= A(x - \hat{x}) - GC(x - \hat{x}) + Bw - Gv$$

$$= Ae - GCe + \begin{bmatrix} B & G \end{bmatrix} \begin{pmatrix} w \\ v \end{pmatrix}$$

$$\dot{e} = (A - GC)e + \begin{bmatrix} B & G \end{bmatrix} \begin{pmatrix} w \\ v \end{pmatrix}$$

- Given that w and v are Gaussian noise with correlation S_w and S_v

$$E(w(t)w(t+\tau)) = S_w \delta(\tau)$$

what does this mean?

- Result is the Riccati Equation (since $\frac{d}{dt}(V_1 V_2) = V_1 \frac{dV_2}{dt} + \frac{dV_1}{dt} V_2$)

$$\dot{\Sigma}_e(t) = \Sigma_e A^T + A \Sigma_e + B_w S_w B_w^T - \Sigma_e C^T S_v^{-1} C \Sigma_e$$

such that

$$G(t) = \Sigma_e(t) C^T S_v^{-1}$$

and

$$\dot{\hat{x}}(t) = A \hat{x} + Bu + G(t)(m(t) - C \hat{x})$$

- Comments

- As the measurement noise v increases (i.e. S_v), the Gain G decreases.

- As input noise increases (i.e. S_w), the gain increases! Why?

Steady State Kalman Filter (i.e. $G \neq G(t)$)

$$\dot{\Sigma}_e(t) = 0$$

- Solve steady Riccati Equation

(Matlab)

$$\Sigma_e A^T + A \Sigma_e + B_w S_w B_w^T - \Sigma_e C^T S_v^{-1} C \Sigma_e = 0$$

- Solve Hamiltonian Matrix (eigenvectors)

($Av = \lambda v$)

$$\begin{bmatrix} -A^T & C^T S_v^{-1} C \\ B_w S_w B_w^T & A \end{bmatrix} \Psi = \Psi \Lambda$$

$$\text{Then partition } \Psi = \begin{bmatrix} \Psi_{11} & \Psi_{12} \\ \Psi_{21} & \Psi_{22} \end{bmatrix}$$

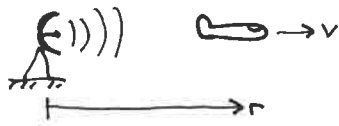
$$\Rightarrow \Sigma_e = \Psi_{22} (\Psi_{12}^{-1})$$

$$\hat{=} G = \Psi_{22} \Psi_{12}^{-1} C^T S_v^{-1}$$

More information

Linear Optimal Control, Jeffrey Burt

Ex: Radar Tracking



We measure distance from time of flight.
Can we estimate velocity too?

$$\begin{pmatrix} \dot{r} \\ \ddot{r} \end{pmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{pmatrix} r \\ \dot{r} \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} w$$

$$\Rightarrow A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad B = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$m = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{pmatrix} r \\ \dot{r} \end{pmatrix} + v$$

$$C = \begin{bmatrix} 1 & 0 \end{bmatrix} \quad D = \begin{bmatrix} 1 \end{bmatrix}$$

If the std-dev of range error is 100 ft, then $S_v = \sigma^2 \Delta t$

Say, the scanning time is 1 second

$$S_v = (100)^2 \cdot 1 = 10000 \text{ ft}^2 \text{ s}$$

Also assume that the air is glassy smooth, the std-dev ~~variance~~ of flight speed is ~~2000~~! ^{acceleration} $S_w = \frac{1 \text{ ft}}{5^2} \frac{1}{\text{Hz}}$

The Hamiltonian is

$$\begin{bmatrix} 0 & 0 \\ -1 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 10000 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

vectors
Eigenvectors are

$$\begin{matrix} -0.0007 & -0.0007 \\ 0 & 0 \\ 0.99499 & 0.99499 \\ -0.07036 & -0.07036 \end{matrix}$$

Yuck!
~~Not~~ Not numerically stable!

Iterate Riccati eqn in time (until $\dot{\hat{z}}_e = 0$)

$$\Sigma = \begin{bmatrix} 1414 & 100 \\ 100 & 14.14 \end{bmatrix} \Rightarrow G = \begin{bmatrix} 1414 & 100 \\ 100 & 14.14 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \frac{1}{10000} = \begin{pmatrix} 0.1414 \\ 0.01 \end{pmatrix}$$

$$\dot{\hat{x}} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \hat{x} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u + \begin{pmatrix} 0.1414 \\ 0.01 \end{pmatrix} (m - \begin{bmatrix} 1 & 0 \end{bmatrix} \hat{x})$$