

$K_k = 1 \times 2$

3.1 Show $\frac{\partial^2 J}{\partial \hat{x}^2}$ is positive definite

$\frac{\partial J}{\partial \hat{x}} = -2y^T H + 2\hat{x}^T H^T H$

$\frac{\partial^2 J}{\partial \hat{x}^2} = 0 + 2H^T H$

and $H^T H$ is always positive!

3.2 Prove P_k is positive definite given P_{k-1} & R_k are positive definite

$P_k = (I - K_k H_k) P_{k-1} (I - K_k H_k)^T + K_k R_k K_k^T$

Given P_{k-1} is positive definite, then $x^T P_{k-1} x > 0$ for all x ($n \times 1$ vector), including

$x' = x^T (I - K_k H_k)^T$. Same logic w/ R_k . If $(I - K_k H_k) P_{k-1} (I - K_k H_k)^T$ and $K_k R_k K_k^T$ is positive definite, then so does its sum.

3.3 $P_0 = \infty I$, $H_k = 1$
 $K_1 = P_0 H_k^T (H_k P_0 H_k + R_k)^{-1}$
 $= \infty 1 (\infty + R_k)^{-1} = 1$

$P_1 = (I - K_1 H_k) P_0 = (1 - 1 \cdot 1) \infty = 0$

3.4 $\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} x + \begin{bmatrix} 1 \\ 4 \end{bmatrix}$

~~$\hat{x} = (H^T H)^{-1} H^T y$~~

a) $\hat{x} = (H^T R^{-1} H)^{-1} H^T R^{-1} y$
 $= ([1 \ 1] \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}^{-1} [1 \ 1]^T)^{-1} [1 \ 1] \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}^{-1} y$
 $= (1 + \frac{1}{4})^{-1} (\frac{y_1}{1} + \frac{y_2}{4}) = \frac{4}{5} (y_1 + \frac{y_2}{4})$
 $= \frac{4}{5} y_1 + \frac{1}{5} y_2$

b) $\sigma^2 = \frac{4}{5}(1) + \frac{1}{5}(4) = 1.6$
 ~~$\sigma^2 = 1.6$~~

c) $\sigma = \frac{1}{2}(1) + \frac{1}{2}(4) = 2.5$

↑ this might be correct...

b) $P_0 = \infty$
 $K_k = \infty 1 (\infty + 1)^{-1} = 1$
 $P_1 = 0 + 1 = 1$
 $P_2 = 1 \times (1 + 1)^{-1} = \frac{1}{2}$
 $P_3 = (\frac{1}{2})^2 + (\frac{1}{2})^2 = \frac{2}{4} = \frac{1}{2}$
 $P_4 = \frac{4}{5}$

c) $E(\frac{y_1 + y_2}{2} + \hat{x})$
 $= E(\frac{y_1}{2}) + E(\frac{y_2}{2}) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$

3.5 $P_0 = 1$, $R = \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}$
 $K_1 = \frac{1}{H_1} = \frac{1}{1} = 1$
 $P_1 = \frac{1}{2} + \frac{1}{2} = 1$
 $P_2 = \frac{1}{2}$
 $P_3 = \frac{4}{9}$
 $P_4 = 0.2857$

3.6 a) $E(\hat{x}) = \bar{x}$?
 $E(\hat{x}) = E(\frac{1}{n} \sum_{i=1}^n x_i) = \frac{1}{n} n \bar{x} = \bar{x}$
 b) Find $E(x_i x_j)$ in terms of \bar{x} and σ^2
 If $i \neq j$, $E[(x_i - \bar{x})(x_j - \bar{x})] = 0 = E[x_i x_j - \bar{x} x_i - x_j \bar{x} + \bar{x}^2]$
 $\Rightarrow E[x_i x_j] - \bar{x}^2 = 0 \Rightarrow E[x_i x_j] = \bar{x}^2$
 If $i = j$, $\sigma^2 = E[x^2] - \bar{x}^2 \Rightarrow E[x_i x_j] = \bar{x}^2 + \sigma^2$

c) $\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \hat{x})^2 \Rightarrow E(\hat{\sigma}^2) = E(\frac{1}{n} \sum_{i=1}^n (x_i - \hat{x})^2)$
 $= E(\frac{1}{n} \sum_{i=1}^n (x_i^2 - 2x_i \hat{x} + \hat{x}^2)) = \frac{1}{n} [\sum_{i=1}^n E(x_i^2) - 2 \sum_{i=1}^n E(x_i \hat{x}) + \sum_{i=1}^n E(\hat{x}^2)]$
 $= \frac{1}{n} [\sum_{i=1}^n E(x_i^2) - \frac{2}{n} \sum_{i=1}^n \sum_{j=1}^n E(x_i x_j) + \frac{1}{n} \sum_{i,j,k=1}^n E(x_i x_j x_k)]$
 $= \frac{1}{n} [\sum_{i=1}^n E(x_i^2) - \frac{2}{n} \sum_{i,j=1}^n E(x_i x_j)] = \frac{1}{n} [n(\bar{x}^2 + \sigma^2) - \frac{2}{n} (n^2 \bar{x}^2 + n \sigma^2)]$

3.7 actual	1	2	3	RMSE	Absolute Error	
est 1	3	4	1	2	2	1.88
est 2	1	2	6	1.73	1	1.41 <small>best rms</small>
est 3	5	6	7	4	4	0 <small>best std</small>

est 2 best intuitive est 3 worst intuitive

3.8 $f(x) = \begin{cases} \frac{x}{0.75} & x \in [0, 0.75] \\ \frac{2}{1.25} - \frac{x}{1.25} & x \in [0.75, 2] \end{cases}$

$$\int_0^{0.75} \frac{x}{0.75} dx = \frac{1}{0.75} \frac{x^2}{2} \Big|_0^{0.75} = 0.375$$

$$\int_{0.75}^2 \left(\frac{2}{1.25} - \frac{x}{1.25} \right) dx = \left. \frac{2}{1.25}x - \frac{1}{1.25} \frac{x^2}{2} \right|_{0.75}^2 = 0.625$$

$$\int_0^\infty \left(\frac{2}{1.25} - \frac{x}{1.25} \right) dx = 0.882$$

$$\frac{2}{1.25}x \Big|_x - \frac{1}{1.25} \frac{x^2}{2} \Big|_x = 0.882 = \frac{2}{1.25}x - \frac{1}{1.25} \frac{x^2}{2}$$

$$\Rightarrow \hat{x} = \frac{0.882}{1.25} = 0.7056 \approx 0.71$$

b) mode $\hat{x} = 0.75$

c) mean = $\int_0^{0.75} \frac{x^2}{0.75} dx + \int_{0.75}^2 x \left(\frac{2}{1.25} - \frac{x}{1.25} \right) dx$

$$= \frac{x^3}{3 \cdot 0.75} \Big|_0^{0.75} + \left. \frac{2}{1.25} \frac{x^2}{2} - \frac{1}{1.25} \frac{x^3}{3} \right|_{0.75}^2 = \frac{11}{12}$$

d) $\hat{x} = 1$ or 2 (not really sure why)

3.9 $\sigma_1^2 = 10$

a) combine $\sigma_1^2 = 10$ w/ $\sigma_2^2 = 0$

b) combine $\sigma_1^2 = \sigma_2^2 = \sigma_3^2 = 10$

a) $P_0 = \infty$
 $K_1 = \frac{P_0}{P_0 + R_0} = \frac{\infty}{\infty + 10} = 1$

$$P_1 = (1 - K_1)^2 P_0 + K_1^2 R_0 = 10$$

$$K_2 = \frac{P_1}{P_1 + R_1} = \frac{5}{8}$$

$$P_2 = (1 - K_2)^2 P_1 + K_2^2 R_1 = 3.75$$

b) $P_0 = \infty$

$$K_1 = 1$$

$$P_1 = 10$$

$$K_2 = \frac{1}{2}$$

$$P_2 = 5$$

$$K_3 = \frac{P_2}{P_2 + R_2} = \frac{1}{3}$$

$$P_3 = (1 - K_3)^2 P_2 + K_3^2 R_2 = 3.33$$

\therefore b is better!

3.10 $\dot{x} + 3x = u$ where $u(t)$ is an impulse

$$sX(s) + 3X(s) = 1 \quad \mathcal{L}\{x(t)\} = X(s)$$

$$X(s) = \frac{1}{s+3} \Rightarrow x(t) = e^{-3t}$$

$$X(s) = \frac{1}{s+3} \Rightarrow x(t) = e^{-3t}$$

not sure how to get the other one...

causal & stable: $e^{-3t} u(t)$

anticausal & unstable: $-e^{-3t} u(-t)$

3.11 $S_x(s) = \frac{1-s^2}{s^4-5s^2+4}$ $S_v(s) = 1$

a) optimal noncausal

$$G(\omega) = \frac{S_x(\omega)}{S_x(\omega) + S_v(\omega)} = \frac{1-s^2}{s^4-5s^2+4} = \frac{1-s^2}{s^4-6s^2+5}$$

$$= \frac{1-s^2}{(s^2-5)(s^2+1)} = \frac{1}{5-s^2} \quad \text{or} \quad \frac{1}{5-(j\omega)^2} = \frac{1}{5+\omega^2}$$

i think...

b) optimal causal

$$S_{xv}(\omega) = \frac{s^4-6s^2+5}{s^4-5s^2+4} = \frac{s^2-5}{s^2-4} = \frac{(j\omega)^2-5}{(j\omega)^2-4} = \frac{\omega^2+5}{\omega^2+4}$$

$$G(\omega) = \frac{(s+5)(s-5)}{(s+2)(s-2)} = \frac{(s+5)(s-5)}{(s+2)(s-2)}$$

$$G(s) = \frac{s+5}{s+2} \left[\begin{array}{l} \text{causal} \\ \text{or} \end{array} \right] \frac{-1+s^2}{(s-5)(s-2)}$$

$$= \frac{s+2}{s+5} \left[\frac{s-2}{s-5} \frac{1}{(2-s)(2+s)} \right]$$

$$= \frac{1}{s+5}$$

3.12 $G(s) = \frac{1}{s-3}$

similar to 3.10

$$\Rightarrow e^{-3t} u(t)$$

$$\Rightarrow -e^{-3t} u(-t)$$