

Exercise 2:

Given, $X \sim N(\mu, \sigma^2)$ and $\epsilon \sim N(0, \tau^2)$.

So, $E[X] = \mu$ and $\text{Var}[X] = \sigma^2$

$$\begin{aligned} \text{and } E[X^2] &= \text{Var}[X] + (E[X])^2 \\ &= \sigma^2 + \mu^2 \end{aligned}$$

and. $E[\epsilon] = 0$ and $\text{Var}[\epsilon] = \tau^2$

$$\begin{aligned} \text{and } E[\epsilon^2] &= \text{Var}[\epsilon] + (E[\epsilon])^2 \\ &= \tau^2 \end{aligned}$$

and . X and ϵ are independent.

so, $E[\epsilon X] = E[\epsilon] \cdot E[X] = 0$

Given, $\hat{x} = cY + d$ and $Y = gX + \epsilon$

$$\begin{aligned} E[(x - \hat{x})^2] &= E[(x - cY - d)^2] \quad [\text{Replace with the value of } \hat{x}] \\ &= E[(x - cgx - c\epsilon - d)^2] \quad [\text{Replace with value of } Y] \\ &= E[((1 - cg)x - (c\epsilon + d))^2] \\ &= E[(1 - cg)^2 x^2 - 2(1 - cg)x(c\epsilon + d) \\ &\quad + (c\epsilon + d)^2] \\ &= (1 - cg)^2 E[x^2] - 2(1 - cg)(E[c\epsilon x] + E[xd]) \\ &\quad + E[(c\epsilon + d)^2] \\ &= (1 - cg)^2 (\sigma^2 + \mu^2) - 2(1 - cg)[c \cdot E[\epsilon x] + d E[x]] \\ &\quad + E[c^2 \epsilon^2 + 2c \cdot \epsilon \cdot d + d^2] \\ &= (1 - cg)^2 (\sigma^2 + \mu^2) - 2(1 - cg) \cdot d \mu + c^2 \tau^2 + d^2 \\ &\quad \boxed{[\because E[\epsilon x] = 0 \text{ and } E[\epsilon] = 0]} \end{aligned}$$

To find the minimize of $E[(x - \hat{x})^2]$, we will derivate $E[(x - \hat{x})^2]$ with respect to c and d and set the derivative result zero.

Derivate $E[(x - \hat{x})^2]$ with respect to d.

$$\frac{\partial}{\partial d} E[(x - \hat{x})^2] = 0 - 2(1 - cg)M + 2d$$

$$\text{Now, } 2d = 2(1 - cg) \cdot M$$

$$\therefore d = (1 - cg) \cdot M \quad \dots \text{(i)}$$

Derivate $E[(x - \hat{x})^2]$ with respect to c.

$$\frac{\partial}{\partial c} E[(x - \hat{x})^2] = -2g(1 - cg)[\sigma^2 + M^2] + 2gdM + 2c\pi^2$$

$$\text{Now, } -2g(1 - cg)[\sigma^2 + M^2] + 2gdM + 2c\pi^2 = 0$$

$$\Rightarrow -g(1 - cg)[\sigma^2 + M^2] + g(1 - cg)M^2 + 2c\pi^2 = 0$$

[values of d in (i)]

$$\Rightarrow -g\sigma^2 - gM^2 + cg^2\sigma^2 + cg^2M^2 + gM^2 - cg^2M^2 + c\pi^2 = 0$$

$$\Rightarrow -g\sigma^2 + cg^2\sigma^2 + c\pi^2 = 0$$

$$\therefore c = \frac{g\sigma^2}{g^2\sigma^2 + \pi^2} \quad \dots \text{(ii)}$$

put the value of (ii) in (i)

$$d = \left(1 - \frac{g^2\sigma^2}{g^2\sigma^2 + \pi^2}\right) \cdot M = \frac{Mg^2\sigma^2 + M\pi^2 - Mg^2\pi^2}{g^2\sigma^2 + \pi^2}$$

$$\therefore d = \frac{M\pi^2}{g^2\sigma^2 + \pi^2}$$

If the second order derivative is positive definite (>0) then we can say it a global minimum.

$$\frac{\partial}{\partial d^r} [E(x - \hat{x})] = 2 > 0$$

$$\text{and } \frac{\partial^2}{\partial d^r} [E(x - \hat{x})^2] = 2(\sigma^2 + \mu^2)g^r + 2\tau^2 > 0$$

[because $\sigma^2 \neq 0$ and τ^2 can't be negative]

So, both of them are positive definite. and we can say that c and d are also global minimum.