Solution to GATE ST 2023.40

Devansh Jain - EE22BTECH11018

Question: Let $X_1, X_2, ..., X_{10}$ be a random sample of size 10 from a population having $\mathcal{N}\left(0, \theta^2\right)$ distribution, where $\theta > 0$ is an unknown parameter. Let $T = \frac{1}{10} \sum_{i=1}^{10} X_i^2$. If the mean square error of cT (c > 0) as an estimator of θ^2 , is minimized at $c = c_0$, then the value of c_0 equals

- (i) $\frac{5}{6}$ (ii) $\frac{2}{3}$ (iii) $\frac{3}{5}$
- (iv) $\frac{1}{2}$

(GATE ST 2023)

Solution: The mean of T is given by,

$$E(T) = E\left(\frac{1}{10} \sum_{i=1}^{10} X_i^2\right)$$
 (1)

$$= \frac{1}{10} \sum_{i=1}^{10} E\left(X_i^2\right) \tag{2}$$

Since

$$E\left(X^{2}\right) = V\left(X\right) + \left(E\left(X\right)\right)^{2} \tag{3}$$

Using (3)

$$E(T) = \frac{1}{10} \sum_{i=1}^{10} V(X_i) + E(X_i)^2$$
(4)

$$=\frac{1}{10}\left(10\theta^2\right)\tag{5}$$

$$=\theta^2\tag{6}$$

Similarly, using (3), the variance of T is given by,

$$V(T) = E\left(T^2\right) - E(T)^2\tag{7}$$

$$= \frac{1}{100} \left(E\left(\left(\sum_{i=1}^{10} X_i^2 \right)^2 \right) - \left(E\left(\sum_{i=1}^{10} X_i^2 \right) \right)^2 \right)$$
 (8)

But

$$E\left(\left(\sum_{i=1}^{10} X_i^2\right)^2\right) = E\left(\sum_{i=1}^{10} \sum_{j=1}^{10} X_i^2 X_j^2\right)$$
(9)

$$= \sum_{i=1}^{n} \sum_{j=1}^{n} E(X_i^2 X_j^2)$$
 (10)

and

$$\left(\mathbb{E} \left(\sum_{i=1}^{10} X_i \right) \right)^2 = \left(\sum_{i=1}^{10} \mathbb{E} \left(X_i \right) \right)^2 \tag{11}$$

$$= \sum_{i=1}^{10} \sum_{j=1}^{10} E(X_i) E(X_j)$$
 (12)

Using (10), (12), and the definition of covariance,

$$V(T) = \frac{1}{100} \left(\sum_{i=1}^{10} \sum_{j=1}^{10} \left(E\left[X_i^2 X_j^2 \right] - E\left(X_i^2 \right) E\left(X_j^2 \right) \right) \right)$$
 (13)

$$= \frac{1}{100} \left(\sum_{i=1}^{10} \sum_{j=1}^{10} \text{cov}\left(X_i^2, X_j^2\right) \right)$$
 (14)

As all the variables are i.i.d's and are thus uncorrelated,

$$\operatorname{cov}\left(X_{i}^{2}, X_{j}^{2}\right) = \begin{cases} 0 & \text{if } i \neq j \\ V\left(X_{i}^{2}\right) & \text{if } i = j \end{cases}$$

$$\tag{15}$$

Now,

$$V(T) = \frac{1}{100} \sum_{i=1}^{10} \text{cov}\left(X_i^2, X_i^2\right)$$
 (16)

$$= \frac{1}{100} \sum_{i=1}^{10} V(X_i^2) \tag{17}$$

Using (3) to find $V(X_i^2)$,

$$V\left(X_{i}^{2}\right) = E\left(X_{i}^{4}\right) - E\left(X_{i}^{2}\right)^{2} \tag{18}$$

Using moment generating function to find $E(X_i^4)$ and $E(X_i^2)$,

$$M_X(t) = E\left(e^{tX}\right) \tag{19}$$

$$=e^{\frac{1}{2}\theta^2t^2} \tag{20}$$

Differentiating it with respect to t,

$$\frac{dM_X(t)}{dt} = E\left(Xe^{tX}\right) \tag{21}$$

$$\frac{d^{n}M_{X}\left(t\right)}{dt^{n}}=E\left(X^{n}e^{tX}\right)\tag{22}$$

$$\frac{dM_X(t)}{dt} = \theta^2 t e^{\frac{1}{2}\theta^2 t^2} \tag{23}$$

$$\frac{d^2 M_X(t)}{dt^2} = \theta^2 e^{\frac{1}{2}\theta^2 t^2} + \theta^4 t^2 e^{\frac{1}{2}\theta^2 t^2}$$
 (24)

$$\frac{d^3 M_X(t)}{dt^3} = \theta^4 t e^{\frac{1}{2}\theta^2 t^2} + 2\theta^4 t e^{\frac{1}{2}\theta^2 t^2} + \theta^6 t^3 e^{\frac{1}{2}\theta^2 t^2}$$
(25)

$$=3\theta^4 t e^{\frac{1}{2}\theta^2 t^2} + \theta^6 t^3 e^{\frac{1}{2}\theta^2 t^2} \tag{26}$$

$$\frac{d^4 M_X(t)}{dt^4} = 3\theta^4 e^{\frac{1}{2}\theta^2 t^2} + 3\theta^6 t^2 e^{\frac{1}{2}\theta^2 t^2} + 3\theta^6 t^2 e^{\frac{1}{2}\theta^2 t^2} + \theta^8 t^4 e^{\frac{1}{2}\theta^2 t^2}$$
(27)

$$=3\theta^4 e^{\frac{1}{2}\theta^2 t^2} + 6\theta^6 t^2 e^{\frac{1}{2}\theta^2 t^2} + \theta^8 t^4 e^{\frac{1}{2}\theta^2 t^2}$$
(28)

Now,

$$E(X^4) = \frac{d^4 M_X(t)}{dt^4}|_{t=0}$$
 (29)

$$=3\theta^4\tag{30}$$

$$E(X^{2}) = \frac{d^{2}M_{X}(t)}{dt^{2}}|_{t=0}$$

$$= \theta^{2}$$
(31)

$$=\theta^2\tag{32}$$

$$V(X_i^2) = 3\theta^4 - (\theta^2)^2$$

$$= 2\theta^4$$
(33)

$$=2\theta^4\tag{34}$$

$$V(T) = \frac{1}{100} \sum_{i=1}^{10} 2\theta^4 \tag{35}$$

$$=\frac{1}{5}\theta^4\tag{36}$$

The mean square error of cT as an estimator of θ^2 is given by

$$f = E\left(\left(cT - \theta^2\right)^2\right) \tag{37}$$

Since

$$E(X^2) = V(X) + (E(X))^2$$
 (38)

Now,

$$f = V(cT - \theta^2) + \left(E(cT - \theta^2)^2\right)$$
(39)

$$=V(cT) + \left(E(cT) - \theta^2\right)^2 \tag{40}$$

$$=c^{2}V\left(T\right) +\left(cE\left(T\right) -\theta^{2}\right) ^{2} \tag{41}$$

$$=\frac{c^2}{5}\theta^4 + \theta^4(c-1)^2\tag{42}$$

Minimizing the mean square error,

$$\frac{df}{dc} = 0\tag{43}$$

$$\frac{df}{dc} = \frac{2c}{5}\theta^4 + 2\theta^4 (c - 1) \tag{44}$$

$$\frac{c}{5} + c - 1 = 0 \tag{45}$$

$$6c = 5 \tag{46}$$

$$c = \frac{5}{6} \tag{47}$$

To verify that cT is a good estimate for θ , let

$$\theta = 0.5 \tag{48}$$

$$E\left(\left(cT - \theta^2\right)^2\right) = \frac{1}{5} \left(\frac{25}{36}\right) (0.5)^4 + \left(\frac{-1}{6}\right)^2 (0.5)^4 \tag{49}$$

$$=\frac{1}{96}\tag{50}$$

$$= 0.0104$$
 (51)

The mean square error is small relative to the value of θ , hence cT is a good estimate.

