

Normal example

MTU6136p13a.pdf

Now as $\hat{\phi} = \frac{1}{n} \sum_{i=1}^n Y_i^2$

$$\begin{aligned} \text{Var}[\hat{\phi}] &= \frac{1}{n^2} \sum_{i=1}^n \text{Var}[Y_i^2] \\ &= \frac{1}{n} \text{Var}[Y^2] \end{aligned}$$

(as Y_i^2 $i=1, \dots, n$ are independent and also IID)
 (independent and identically distributed $N(\mu, \sigma^2)$)

and $\text{Var}[Y^2] = E[Y^4] - (E[Y^2])^2$

Now $E[Y^2] = \mu^2 + \sigma^2$

+ We know for the Normal that

$$E[(Y-\mu)^3] = 0$$

$$E[(Y-\mu)^4] = 3\sigma^4$$

so (Prob. 2 or differentiate mgf of $Y-\mu$ is $e^{-\frac{1}{2}\sigma^2 t^2}$ 3 and 4 times w.r.t. t = put $t=0$)

$$(Y-\mu)^3 = Y^3 - 3Y^2\mu + 3Y\mu^2 - \mu^3$$

$$\Rightarrow 0 = E[Y^3] - 3\mu(\mu^2 + \sigma^2) + 3\mu^3 - \mu^3$$

$$\rightarrow E[Y^3] = \underline{3\mu\sigma^2 + \mu^3}$$

and

$$\begin{aligned} (Y-\mu)^4 &= (Y^2 - 2Y\mu + \mu^2)^2 \\ &= (Y^4 - 4Y^3\mu + 6Y^2\mu^2 - 4Y\mu^3 + \mu^4) \end{aligned}$$

$$\rightarrow 3\sigma^4 = E[Y^4] - 4(3\mu\sigma^2 + \mu^3)\mu + 6\mu^2(\mu^2 + \sigma^2) - 4\mu^4 + \mu^4$$

$$\rightarrow E[Y^4] = \underline{3\sigma^4 + 6\mu^2\sigma^2 + \mu^4}$$

Hence

$$\begin{aligned} \text{Var}[\hat{\phi}] &= \frac{1}{n} \left[3\sigma^4 + 6\mu^2\sigma^2 + \mu^4 - (\mu^2 + \sigma^2)^2 \right] \\ &= \frac{1}{n} \left[2\sigma^4 + 4\mu^2\sigma^2 \right] = \frac{2\sigma^2}{n} (2\mu^2 + \sigma^2) \end{aligned}$$

which is the CRLB

Hence $\hat{\phi}$ is the MVUE

Is there an easier way to prove this?

Yes(!) by finding sufficient and complete statistics (see later) or using the condition of attaining the CRLB: see lecture