

n Sample $N \sim$ mean μ var σ^2

$$(i) \frac{\sum (x_i - \bar{x})^2}{\sigma^2} \sim \frac{(n-1)\sigma^2}{\sigma^2} \sim \chi_{n-1}^2$$

$$(ii) \hat{\sigma}^2 = \frac{1}{n+b} \sum (x_i - \bar{x})^2$$

bias $\hat{\sigma}^2$ (unbiased $b = -1$)



$$\begin{aligned} \hat{\sigma}^2 &= \frac{1}{n-1} \sum (x_i - \bar{x})^2 \\ \frac{(n-1)\sigma^2}{\sigma^2} &= \sum \frac{(x_i - \bar{x})^2}{\sigma^2} \end{aligned}$$

$$b, \text{ as } (\hat{\sigma}^2) = E(\hat{\sigma}^2) - \sigma^2$$

$$E\left[\frac{\sum (x_i - \bar{x})^2}{\sigma^2}\right] = n-1$$

From $\sum (x_i - \bar{x})^2 = (n+b)\hat{\sigma}^2$

$$E\left(\frac{(n+b)\hat{\sigma}^2}{\sigma^2}\right) = n-1 \Rightarrow \underbrace{(n+b)}_{\sigma^2} E(\hat{\sigma}^2) = (n-1)$$

$$E(\sigma^2) = \frac{(n-1)}{n+b} \sigma^2$$

$$bias(\sigma^2) = E(\sigma^2) - \sigma^2 = \left(\frac{n-1}{n+b} \right) \sigma^2 - \sigma^2$$

$$= \frac{n\sigma^2 - \sigma^2 - n\sigma^2 - b\sigma^2}{n+b} = - \frac{(1+b)\sigma^2}{n+b}$$

$$d_{b=1} \Rightarrow - \frac{(1-1)\sigma^2}{n-1} = 0$$

$$MS E (\hat{\sigma}^2) = \frac{2(n-1) + (1+b)^2 \cdot \sigma^4}{(n+b)^2}$$

$$Var \left(\frac{\sum (x_i - \bar{x})^2}{\sigma^2} \right) = 2(n-1) - \cancel{\chi^2} Var$$

$$Var \left(\frac{(n+b)\hat{\sigma}^2}{\sigma^2} \right) = 2(n-1) \Rightarrow \frac{(n+b)^2}{\sigma^4} Var(\hat{\sigma}^2) = \frac{2(n-1)}{2(n-1)}$$

$$\text{Var}(\hat{\sigma}^2) = \frac{2(n-1)}{(n+b)^2} \sigma^4$$

$$b \cdot \sigma^2 = \frac{(1+b)^2}{(n+b)} \sigma^2$$

MSE =

$n \rightarrow \infty$

MSE = ?

0

consistent

PQ4 - Unit 1 Apr 2008 Q 11

(i) no. of samples n , K policies expire &
Exp is stopped $0 < K < n$

$$\text{Exp}(\lambda) = \text{dist}^n$$

$$P(T > t_K) = 1 - P(T \leq t_K) = 1 - F_T(t_K)$$

$\stackrel{\text{experiment}}{=}$
stop time

from pg 11 of tables we have $F_T(t_K) = 1 - e^{-\lambda t_K}$
 $\Rightarrow 1 - (1 - e^{-\lambda t_K}) = e^{-\lambda t_K}$ n.p

$$b) L(\lambda) = \frac{\lambda^K}{\prod_{i=1}^K f(t_i)} \prod_{j=K+1}^n P(t > t_K) \xrightarrow{n-K} \text{polices still active}$$

$$= \frac{\lambda^K}{\prod_{i=1}^K} (\lambda e^{-\lambda t_i}) \prod_{j=K+1}^n e^{-\lambda t_K} \dots \text{from part (a)}$$

$$L(\lambda) = \lambda^K e^{-\lambda \sum t_i} e^{-(n-K)\lambda t_K} \rightarrow \text{information ceases}$$

$$\log L(\lambda) = K \log \lambda - \lambda \sum_{i=1}^K t_i - (n-K)\lambda t_K$$

taking derivative w.r.t λ

$$L'(\lambda) = \frac{K}{\lambda} - \sum t_i - (n-K)t_K \rightarrow 0$$

Equating $\frac{dL(\lambda)}{d\lambda} = 0$ we get

$$0 = \frac{k}{\lambda} - \sum t_i - (n-k)t_k$$

$$\frac{k}{\lambda} = \sum t_i + (n-k)t_k \Rightarrow$$

$$\hat{\lambda} = \frac{k}{\sum t_i + (n-k)t_k}$$

differentiating eq (1) w.r.t λ again

$$\frac{d^2 L(\lambda)}{d\lambda^2} = -\frac{k}{\lambda^2} < 0$$

$$c) n = 20 \quad k = 5 \quad t_k = 21.54 \text{ (lost expiry)}$$

$$\sum_{i=1}^{k=5} t_i = 54.82$$

$$\lambda = \frac{k}{\sum t_i + (n-k) t_k} = \frac{5}{54.82 + 15 \times 21.54} = 0.0132$$

(ii)a) n policies which are independent
and each of these have prob p of
expiring at time t_0

$\text{Bin}(n, p)$ so we have to find P

$$P = P(\tau \leq t_0) = 1 - e^{-\lambda t_0}$$

and policies will remain active with
prob $1 - p$ $K \sim \text{bin}(n, 1 - e^{-\lambda t_0})$

PA 8 Sept 09 Q5

$$(i) E(X) = \int_0^\theta x f(x) dx = \int_0^\theta 2x^2 \theta^{-2} dx$$

$$E(X) = \frac{2}{\theta^2} \int_0^\theta x^2 dx = \frac{2\theta^3}{3\cancel{\theta^2}} = \frac{2\theta}{3} \rightarrow \text{biased estimator}$$

$$\boxed{S = \frac{3X}{2}} \rightarrow \text{Estimator}$$

$$E(S) = \frac{3}{2} E(X) = \frac{\cancel{3}}{2} \times \frac{2}{\cancel{3}} \times \theta \Rightarrow \text{unbiased estimator}$$

PQ 5 Sept 08 Q10

(i) Solved Verbally

(ii) $x_1 + x_2 + x_3 = n$ Claims

$x_1 = n_j \in 1st$

$x_2 = n_j \in 2nd$

$x_3 = p_{ss} \in 2nd$

$$L(\theta) = (1-\theta)^{x_1} [\theta(1-\theta)]^{x_2} (\theta^2)^{x_3}$$

$$= (1-\theta)^{x_1+x_2} \theta^{x_2+2x_3}$$

$$\log L(\theta) = (x_1 + x_2) \log(1-\theta) + (x_2 + 2x_3) (\log \theta)$$

$$\frac{d \log L(\theta)}{d\theta} = -\left(\frac{x_1 + x_2}{(1-\theta)}\right) + \frac{x_2 + 2x_3}{\theta}$$

equating to zero

$$\frac{x_2 + 2x_3}{\theta} = \frac{x_1 + x_2}{1-\theta} \Rightarrow \theta(x_1 + x_2) = (1-\theta)(x_2 + 2x_3)$$

$$= \theta(x_1 + x_2 + x_2 + 2x_3) = x_2 + 2x_3$$

$$\Rightarrow \hat{\theta} = \frac{x_2 + 2x_3}{x_1 + 2x_2 + 2x_3}$$

$$\frac{d^2 \log L(\theta)}{d\theta^2} = \frac{d}{d\theta} \left[\frac{(x_2 + 2x_3)}{\theta} - \frac{(x_1 + x_2)}{1-\theta} \right]$$

$$= -\frac{(x_2 + 2x_3)}{\theta^2} - \frac{1(x_1 + x_2)}{(1-\theta)^2}$$

$$-E \left[\frac{1}{\frac{d^2 \log L(\theta)}{d\theta^2}} \right] = \text{CRLB}$$

$$E \left[\frac{(x_1 + x_2)}{(1-\theta)^2} - \frac{(x_2 + 2x_3)}{\theta^2} \right]$$

$$= \frac{n(1-\theta) - n\theta(1-\theta)}{(1-\theta)^2} - \frac{n\theta(1-\theta) - 2n\theta^2}{\theta^2}$$

$$= \frac{n - n\theta}{1-\theta} - \frac{n(1-\theta) - 2n\theta}{\theta}$$

$$= \frac{n(1+\theta)}{1-\theta} - \frac{n[1-\theta+2\theta]}{\theta} = -\frac{n(1+\theta)}{1-\theta} - \frac{n(1+\theta)}{\theta}$$

$$-n(1+\theta) \left[\frac{\cancel{\theta} + 1 - \cancel{\theta}}{\theta(1-\theta)} \right] \Rightarrow -\frac{n(1+\theta)}{\theta(1-\theta)} < 0$$

$$-\frac{1}{E \left[\frac{\partial^2 \log L(\theta)}{\partial \theta^2} \right]} = - \left[-\frac{n(1+\theta)}{\theta(1-\theta)} \right]$$

$$\frac{\theta(1-\theta)}{n(1+\theta)} = CRB$$

P03 Unit 1 Apr 08 Q8

$$A \sim N\left(0, \underbrace{9 \times 100}_{\sigma^2, n=100}\right) \quad A \sim N(0, 100)$$

$$P(A > 15) = 1 - P(A \leq 15)$$

$$\Rightarrow 1 - P\left(Z \leq \frac{15 - 0}{\sqrt{100}}\right) = 1 - P\left(Z \leq \frac{15}{10}\right)$$

$$1 - P(Z \leq 0.5) \Rightarrow \text{Answer}$$

$$\begin{aligned}
 & P(|Y| \geq 15) \\
 &= P(Y < -15) + P(Y > +15) \\
 &= P(Z < -0.5) + P(Z > 0.5) \\
 &\equiv P(Z > 0.5) + P(Z > 0.5)
 \end{aligned}$$

$$= 2(P(Z > 0.5))$$



$$\begin{aligned}
 &= 2(1 - P(Z \leq 0.5)) = 2(1 - 0.69146) \\
 &= 0.61708
 \end{aligned}$$