

The Rao-Blackwell Theorem

Stat 563
4-8-24

Let $X_1, \dots, X_n \sim \text{iid } f(x|\theta)$

①

Let T be a sufficient statistic for θ .

Let W be an unbiased estimator of $\tau(\theta)$.

Let $\phi(T) = E[W|T]$

Then $\phi(T)$ is an unbiased estimator of $\tau(\theta)$

and $V[\phi(T)] \leq V[W] \quad \forall \theta$.

Proof: $E[\phi(T)] = E[E[W|T]]$
 $= E[W] = \tau(\theta)$

②

$$V[W] = \underbrace{E[V[W|T]]}_{\geq 0} + \underbrace{V[E[W|T]]}_{V[\phi(T)]}$$

$$V[W] \geq V[\phi(T)]$$

Also, $f(x|T)$ is free of θ , so
distribution of $W|T$ is free of θ ,
as is $\phi(T) = E(W|T)$.

Example: $X_1, X_2 \sim \text{iid } \frac{1}{\theta} e^{-x/\theta} \quad x > 0$ (3)

Let $W = X_1$ $E[W] = \theta$ $V[W] = \theta^2$

Find a sufficient statistic for θ .

$$L(\theta) = \frac{1}{\theta} e^{-\frac{x_1}{\theta}} \frac{1}{\theta} e^{-\frac{x_2}{\theta}}$$

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

So $T = X_1 + X_2$ is sufficient for θ .

Find $E[W|T]$ (4)

Start with the joint distribution of W, T .

$W = X_1$	$X_1 = W$	$X_1 > 0$
$T = X_1 + X_2$	$X_2 = T - W$	$X_2 > 0$
		$W > 0$
		$t - w > 0$
		$t > w$

$$|J| = \begin{vmatrix} 1 & 0 \\ -1 & 1 \end{vmatrix} = 1$$

$$g(w, t) = f(x_1, x_2) \cdot 1 = \frac{1}{\theta^2} e^{-\frac{1}{\theta}(x_1+x_2)}$$

$$= \frac{1}{\theta^2} e^{-t/\theta} \quad 0 < w < t$$

$$g(w|t) = \frac{g(w,t)}{h(t)} = \frac{\frac{1}{\theta^2} e^{-t/\theta}}{\frac{1}{\theta^2} t e^{-t/\theta}} \quad (5)$$

Because $T \sim \text{Gamma}(\alpha=2, \beta=\theta)$

$$= \frac{1}{t}, \quad 0 < w < t$$

That is, $W|T \sim \text{Unif}(0, T)$

$$\text{So } E[W|T] = \frac{0+T}{2} = \frac{T}{2} = \frac{X_1+X_2}{2} = \bar{X}$$

Note $E[\bar{X}] = \theta$ and $V[\bar{X}] = \frac{\theta^2}{2}$

A simpler approach might be to (6)

① Find a sufficient statistic for θ , called T

② Find a function of T that is
an unbiased estimator of θ .

Defn: W is a minimum variance unbiased (MVUB) estimator of θ if $E[W] = \theta$ and its variance is less than or equal to the variance of any other unbiased estimator of θ , $\forall \theta$.

⑦

Theorem: If W is a MVUE of θ ,
then W is unique.

Proof: Let W' also be MVUE of θ .

$$\text{Let } W^* = \frac{1}{2}(W + W')$$

$$E[W^*] = \frac{1}{2}[E(W) + E(W')] = \theta$$

$$\begin{aligned} V[W^*] &= \frac{1}{4} [\underline{V(W)} + V(W') + 2 \text{Cov}(W, W')] \\ &= \frac{1}{4} [2V(W) + 2 \text{Cov}(W, W')] \end{aligned}$$

$$V[W^*] = \frac{1}{2}V(W) + \frac{1}{2}\text{Cov}(W, W')$$

⑧

Recall $\rho^2 = \frac{\text{Cov}^2(W, W')}{V(W)V(W')} \leq 1$

$$\text{So } \text{Cov}^2(W, W') \leq (V(W))^2$$

$$|\text{Cov}(W, W')| \leq V(W) \quad \star$$

$$\text{Now } V[W^*] \leq \frac{1}{2}V(W) + \frac{1}{2}V(W)$$

$$V[W^*] \leq V(W)$$

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If, for any θ , $V[W^*] < V[W]$,
our assumption that W was MVUE
would be violated.

$$\therefore V[W^*] = V[W] \quad \forall \theta$$

Equivalently, $\rho = \text{Cor}(W, W') = 1$

Cauchy-Schwarz says $W' = aW + b$

By \star , $\text{Cov}(W, W') = V[W]$

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$$\begin{aligned}\text{Cov}(W, W') &= \text{Cov}(W, aW + b) \\ &= a \text{Cov}(W, W) + \text{Cov}(W, b) \\ &= a V[W] + 0\end{aligned}$$

$$\therefore a = 1$$

$$\text{So } W' = W + b$$

$$E[W'] = E[W] + b$$

$$\theta = \theta + b \quad \therefore b = 0$$

$$\therefore W = W'$$

The Lehmann-Scheffé Theorem

(11)

Let T be a complete sufficient statistic for θ . If there exists W , a function of T , that is an unbiased estimator of $\tau(\theta)$, then W is the unique MVUE of $\tau(\theta)$.

Proof: Assume W exists.

Let W' be another unbiased estimator of $\tau(\theta)$.

By the Rao-Blackwell Theorem,

(12)

$E[W'|T]$ will be an unbiased estimator of $\tau(\theta)$ and its variance will be $\leq V[W']$.

Also note that $E[W'|T]$ is a function of T .

Let $\phi(T)$ and $\psi(T)$ be functions of T that are both unbiased estimators of $\tau(\theta)$.

$$\begin{aligned} E[\varphi(T) - \psi(T)] &= E[\varphi(T)] - E[\psi(T)] \quad (13) \\ &= \tau(\theta) - \tau(\theta) = 0 \quad \forall \theta \end{aligned}$$

By completeness, this implies

$$\varphi(T) - \psi(T) = 0$$

So W and $E[W|T]$ must be equal.

$$\text{Then } V[W] = V[E[W|T]] \leq V[W']$$

So W is MVUE and is unique.