

- a) $(Y_{(1)}, Y_{(n)})$ is complete.
- b) $Y_{(1)}$ is complete if θ_2 is known
- c) $Y_{(n)}$ is complete if θ_1 is known.
- iv) If $\underline{T}(Y)$ is not minimal suff, then
 $\underline{T}(Y)$ is not complete.

30] know ^{can't use for a proof, but generally true} rule of thumb p 137: If $\underline{T}(Y)$ is $1 \times k$ and $\underline{\theta}$ is $1 \times j$ where $k > j$, generally \underline{T} is not complete, especially if $k=2$ and $j=1$.
 Watch out for $N(\theta, \theta^2)$, $N(\sigma, \sigma^2)$,
 $U(a(\theta), b(\theta))$ data.

31] Showing $\underline{T}(Y)$ is complete when Y is not from a k dim exp fam where Ω contains a k dim rectangle (eg a k PREF), is generally very hard. You need to find every g such that $E_{\underline{\theta}}[g(\underline{T})] = 0 \forall \underline{\theta} \in \Omega$

32] Showing $\underline{T}(Y)$ is not complete is usually very hard. You need to show that \underline{T} is not minimal sufficient or find a g where $E_{\underline{\theta}}[g(\underline{T})] = 0 \forall \underline{\theta} \in \Theta$

but $P[g(\underline{T}) \neq 0] > 0$ or $P[g(\underline{T}) = 0] < 1$.
 Use point 30] to get some partial credit.

33] Know for final: often if \underline{T} is not complete, there is some restriction that gives a clue to θ .

$$T = (\bar{Y}, s^2)$$

ex] $Y \sim N(\mu, \mu^2)$ or $N(\sigma^2, \sigma^2)$, try $\bar{Y} - s^2$

ex] $\begin{pmatrix} X \\ Y \end{pmatrix} \sim N_2 \left[\begin{pmatrix} \mu \\ \mu \end{pmatrix}, \begin{pmatrix} \sigma_x^2 & 0 \\ 0 & \sigma_y^2 \end{pmatrix} \right], \theta = (\mu, \sigma_x^2, \sigma_y^2)$

$$f(x, y) = \frac{1}{2\pi \sigma_x \sigma_y} \exp \left\{ -\frac{1}{2} \left[\left(\frac{x-\mu}{\sigma_x} \right)^2 + \left(\frac{y-\mu}{\sigma_y} \right)^2 \right] \right\} =$$

$$\frac{1}{2\pi \sigma_x \sigma_y} \exp \left[-\frac{1}{2} \left(\frac{\mu^2}{\sigma_x^2} + \frac{\mu^2}{\sigma_y^2} \right) \right] \exp \left[\frac{\mu}{\sigma_x^2} x + \frac{\mu}{\sigma_y^2} y - \frac{1}{2\sigma_x^2} x^2 - \frac{1}{2\sigma_y^2} y^2 \right]$$

is a 4 parameter exp fam. Note $\dim(\underline{T}) = 4 > 3 = \dim(\theta)$.

$$\underline{T}(x, y) = \left(\sum_{i=1}^n x_i, \sum_{i=1}^n y_i, \sum_{i=1}^n x_i^2, \sum_{i=1}^n y_i^2 \right)$$

\underline{T} is min suff i) because the n_i do not satisfy a linearity constraint

ii) By LSM $R[(x, y), (w, z)] = f(x, y) / f(w, z) =$

$$\exp \left[\frac{\mu}{\sigma_x^2} (\sum x_i - \sum w_i) \right] \exp \left[\frac{\mu}{\sigma_y^2} (\sum y_i - \sum z_i) \right] \exp \left[-\frac{1}{2\sigma_x^2} (\sum x_i^2 - \sum w_i^2) \right] \exp \left[-\frac{1}{2\sigma_y^2} (\sum y_i^2 - \sum z_i^2) \right]$$

is constant $\forall \theta$ iff $\underline{T}(x, y) = \underline{T}(w, z)$ since $e^a > 0$.

The clue that \underline{T} is not complete is $\mu_x = \mu_y = \mu$.

$$E_{\theta} [g(\underline{T})] = E_{\theta} (\sum x_i - \sum y_i) = n\mu - n\mu = 0 \quad \forall \theta$$

So $g(\underline{T}) = \sum x_i - \sum y_i$, but $P_{\theta} (\sum x_i - \sum y_i = 0) = P_{\theta} (g(\underline{T}) = 0)$

$\approx 0 < 1$, so \underline{T} is not complete.

X_1, \dots, X_n are iid $U(a(\theta), b(\theta))$. Find a min suff stat T . Determine whether T is complete. Typically $T_2 = (X_{(1)}, X_{(n)})$ is 2 dimensional but θ 1 dimensional and T_2 is not complete. Here is an exception.

X_1, \dots, X_n are iid $U(1-\theta, 1+\theta)$, $\theta > 0$.

so $f(x) = \frac{1}{2\theta}$, $1-\theta < x < 1+\theta$.

$T_1 = \max(X_{(n)} - 1, 1 - X_{(1)})$ is min suff

$$P(X_{(n)} \leq t) = P(X_1 \leq t, \dots, X_n \leq t) = [F(t)]^n$$

$$= \int_{1-\theta}^t \frac{1}{2\theta} dx = \frac{t - 1 + \theta}{2\theta}$$

$$f(t) = \frac{n}{2\theta} \left(\frac{t + \theta - 1}{2\theta} \right)^{n-1}, \quad 1-\theta < t < 1+\theta$$

$$E_{\theta}[X_{(n)}] = \int_{1-\theta}^{1+\theta} x \frac{n}{2\theta} \left(\frac{x + \theta - 1}{2\theta} \right)^{n-1} dx.$$

$$\text{Let } U = \frac{x + \theta - 1}{2\theta}, \quad x = 2\theta U + 1 - \theta. \text{ So } x = 1 + \theta \rightarrow U = 1$$

$$\text{and } x = 1 - \theta \rightarrow U = 0, \quad dx = 2\theta dU.$$

$$\text{So } E_{\theta}(X_{(n)}) = n \int_0^1 \frac{2\theta U + 1 - \theta}{2\theta} U^{n-1} 2\theta dU$$

$$= 2\theta n \int_0^1 U^n dU + n(1-\theta) \int_0^1 U^{n-1} dU$$

$$= 2\theta n \frac{U^{n+1}}{n+1} \Big|_0^1 + n(1-\theta) \frac{U^n}{n} \Big|_0^1 =$$

$$2\theta \frac{n}{n+1} + \frac{n(1-\theta)}{n} = 1 - \theta + 2\theta \frac{n}{n+1} \approx 1 + \theta \quad \text{as you should expect}$$

$$\text{Now } P(X_{(n)} \leq t) = 1 - [P(X_1 > t)]^n = 1 - [1 - F(t)]^n = 1 - \left[1 - \frac{t - 1 + \theta}{2\theta} \right]^n, \quad 1 - \theta < t < 1 + \theta.$$

$$\text{So } f_{X_{(1)}}(x) = -n \left(1 - \frac{x+\theta-1}{2\theta}\right)^{n-1} \left(\frac{-1}{2\theta}\right) = \frac{n}{2\theta} \left(\frac{\theta-x+1}{2\theta}\right)^{n-1}$$

for $1-\theta < x < 1+\theta$. So

$$E_{\theta}[X_{(1)}] = \int x f_{X_{(1)}}(x) dx = \int_{1-\theta}^{1+\theta} \frac{n}{2\theta} x \left(\frac{\theta-x+1}{2\theta}\right)^{n-1} dx$$

$$\text{Let } u = \frac{\theta-x+1}{2\theta}, \quad 2\theta u = \theta-x+1 \text{ or } x = \theta+1-2\theta u$$

$$\text{So } x=1-\theta \rightarrow u=1, \quad x=1+\theta \rightarrow u=0, \quad dx = -2\theta du$$

$$E_{\theta}(X_{(1)}) \stackrel{\downarrow}{=} \int_{+1}^0 \frac{n}{2\theta} (\theta+1-2\theta u) u^{n-1} (-2\theta) du$$

$$= n \int_0^1 (\theta+1-2\theta u) u^{n-1} du =$$

$$n(\theta+1) \int_0^1 u^{n-1} du - 2\theta n \int_0^1 u^n du =$$

$$n(\theta+1) \frac{u^n}{n} \Big|_0^1 - 2\theta n \frac{u^{n+1}}{n+1} \Big|_0^1 = n(\theta+1) \frac{1}{n} - 2\theta n \frac{1}{n+1}$$

$$E_{\theta}[X_{(1)}] = \theta+1 - 2\theta \frac{n}{n+1} \quad \left(\approx 1-\theta \text{ as expected} \right)$$

$$\text{Now } f(x) = \prod_{i=1}^n \frac{1}{2\theta} I[1-\theta < x_i < 1+\theta]$$

$$= \left(\frac{1}{2\theta}\right)^n I[1-\theta < x_{(1)} < x_{(n)} < 1+\theta] \quad \frac{1}{h(x)}$$

$$g(T(x)|\theta) = \left(\frac{1}{2\theta}\right)^n I[\max(x_{(n)}-1, 1-x_{(1)}) < \theta]$$

So $(X_{(1)}, X_{(n)})$ is suff by factorization.

(This could be used to show T_2 is not min suff, see notes 32.)

$(\max(x_{(1)}-1, 1-x_{(1)}))$
is minsuft

Now $E_{\theta} [a x_{(1)} + b x_{(n)} + c] = 0 \quad \forall \theta$

if $a=b=1$ and $c=-2$ since

$$1+\theta - 2\theta \frac{n}{n+1} + 1-\theta + 2\theta \frac{n}{n+1} = 2,$$

So if $g(T_2) = g(x_{(1)}, x_{(n)}) = x_{(1)} + x_{(n)} - 2$,
then $E_{\theta} [g(T_2)] = 0 \quad \forall \theta$, but

$P_{\theta} (g(T_2) = 0) = 0 < 1$. So T_2 is not complete

$(x_{(1)}, x_{(n)})$ have a pdf so $x_{(1)} + x_{(n)} - 2$ has a pdf

Easier: $(x_{(1)}, x_{(n)})$ is not a function of $\max(x_{(n)}-1, 1-x_{(1)})$ so not minsuft so not complete.

ex) Suppose Y_1, \dots, Y_n are iid $N(\mu, 1)$. Then
 $T = \sum Y_i \sim N(n\mu, n)$ and

$$E_{\mu} [\sum Y_i - n\mu] = 0 \quad \forall \mu.$$

Can I say $g(T) = \sum Y_i - n\mu$ shows
 T is not complete?

No, $g(T)$ needs to be free of unknown parameters μ .

35) P119 * Let Y_1, \dots, Y_n have pdf or pmf $f(y|\theta)$.
A statistic $w(Y)$ whose distribution does not depend on θ is an ancillary statistic

36) know Basu's Lemma Let Y_1, \dots, Y_n have pdf or pmf 32.5

$f(\underline{y} | \theta)$. If $\underline{T}(\underline{y})$ is complete sufficient,

then $\underline{T}(\underline{y})$ is independent of every ancillary statistic.

ex] For any location family, $Y_{(n)} - Y_{(1)}$ is ancillary and s^2 is ancillary.

For any scale family, $\underline{w}(\underline{y}) = a\left(\frac{y_1}{y_n}, \dots, \frac{y_{n-1}}{y_n}\right)$ is ancillary.

37] If $\underline{T}(\underline{y})$ is not ind of an ancillary statistic, then $\underline{T}(\underline{y})$ is not complete. See ex 4.17.

38] If $\underline{w}(\underline{y})$ is ancillary and $\underline{T}(\underline{y})$ complete, Basu's theorem implies $\underline{w}(\underline{y}) \perp\!\!\!\perp \underline{T}(\underline{y})$

Without finding the joint and marginal dist's of $\underline{w}(\underline{y})$ and $\underline{T}(\underline{y})$,

ex] p 119 Y_1, \dots, Y_n iid $N(\mu, \sigma^2)$. For fixed σ^2 , this is a location family with $\sum Y_i$ complete. Thus $\bar{Y} = \frac{1}{n} \sum Y_i \perp\!\!\!\perp s^2$ by Basu's theorem for any $\sigma^2 > 0$ known. ancillary
Thus $\bar{Y} \perp\!\!\!\perp s^2$ for $\sigma^2 > 0$ unknown.

Problems 4.31, 4.32 have solutions. end exam 2 material