

Stat 543

1-26-05

Recall $T(X)$ sufficient plus

log likelihoods
for x and y
with same
shape
gives $T(X)$ minimal sufficient

) $\Rightarrow T(x) = T(y)$

Example X_1, X_2, \dots, X_n iid Beta (α, β)

We've seen that $T(X) = \left(\prod_{i=1}^n X_i, \prod_{i=1}^n (1-X_i) \right)$
is sufficient for (α, β) - now argue that
it is minimal sufficient

Joint pdf on $[0,1]^n$ is

$$f(x|\alpha, \beta) = \frac{1}{(B(\alpha, \beta))^n} (\prod x_i)^{\alpha-1} (\prod (1-x_i))^{\beta-1}$$

$f(y|\alpha, \beta) = k(y, x) f(x|\alpha, \beta)$ as a function of $(\alpha, \beta) \in (0, \infty)^2$ requires

$$(\prod y_i)^{\alpha-1} (\prod (1-y_i))^{\beta-1} = k(y, x) (\prod x_i)^{\alpha-1} (\prod (1-x_i))^{\beta-1}$$

Set $\beta = 1$ and note that this requires

$$(\prod y_i)^{\alpha-1} = k(y, x) (\prod x_i)^{\alpha-1}$$

$$\left(\frac{\prod y_i}{\prod x_i} \right)^{\alpha-1} = k(y, x) \quad \text{as a function of } \alpha$$

and this requires $\frac{\pi y_i}{\pi x_i} = 1$ i.e. $\pi y_i = \pi x_i$

a similar argument setting $\alpha = 1$ requires
 $\pi(1-x_i) = \pi(1-y_i)$ i.e. $T(x) = T(y)$
 and by the theorem T is minimal sufficient

Are there any "mechanical" means of identifying a minimal sufficient statistic? Yes

- 1) In general ... but the "statistic" is a bit hard to think about
- 2) In special convenient families of dens called "exponential families"

General Result

Thm (Dykin/Lehmann/Scheffé) If $f(x|\theta) \forall \theta$ is either a pdf for X on \mathbb{R}^k or is a pmf for X and $\exists \theta_0$ s.t.

if $f(x|\theta) > 0$ for any θ , $f(x|\theta_0) > 0$
 then the (random function of θ)

$$R(x, \theta) = \frac{f(x|\theta)}{f(x|\theta_0)}$$

is a minimal sufficient statistic

Example $f(x|\theta)$ pmf

		x				
		0	1	2	3	4
θ	1	.1	.25	.2	.4	.05
	2	.2	.2	.4	.1	.1
	3	.1	.35	.2	.3	.05

$$\frac{f(x|\theta)}{f(x|3)} = R(x, \theta)$$

		x				
		0	1	2	3	4
1	1	5/7	1	4/3	1	
2	2	4/7	2	1/3	2	
3	1	1	1	1	1	

Note that the 3 functions of θ for $x=0, 2, 4$ are identical and are different from those of $x=1, 3$ (which are not the same) - so anything equivalent to e.g.

$$T(0) = T(2) = T(4) = 17$$

$$T(3) = 35$$

$$T(1) = 8$$

Example X_1, X_2, \dots, X_n iid $N(\mu, \sigma^2)$

$$R(x, \mu, \sigma^2) = \frac{f(x | \mu, \sigma^2)}{f(x | 0, 1)} = \frac{1}{\sigma^n} \exp \left[-\frac{1}{2} \left[\left(\frac{1}{\sigma^2} - 1 \right) \sum x_i^2 - \frac{2\mu \sum x_i}{\sigma^2} + n \frac{\mu^2}{\sigma^2} \right] \right]$$

Clearly if x and y have same $\sum x_i = \sum y_i$ and $\sum x_i^2 = \sum y_i^2$ then $R(x, \mu, \sigma^2) = R(y, \mu, \sigma^2)$ as functions of μ, σ^2

I want to argue that if $(\sum x_i, \sum x_i^2) \neq (\sum y_i, \sum y_i^2)$
 that they are different functions

If $\sum x_i \neq \sum y_i$ and consider $R(x, 1, 1)$ and $R(y, 1, 1)$

$$R(x, 1, 1) = \exp(\sum x_i + n) \neq$$

$$R(y, 1, 1) = \exp(\sum y_i + n)$$

and $R(x, \mu, \sigma^2)$ and $R(y, \mu, \sigma^2)$ are different functions

If $\sum x_i^2 \neq \sum y_i^2$ consider $R(x, 0, 2)$ and $R(y, 0, 2)$

$$R(x, 0, 2) = \left(\frac{1}{\sqrt{2}}\right)^n \exp\left(-\frac{1}{2}\left[-\frac{3}{4}\sum x_i^2\right]\right) \neq$$

$$R(y, 0, 2) = \left(\frac{1}{\sqrt{2}}\right)^n \exp\left(-\frac{1}{2}\left[-\frac{3}{4}\sum y_i^2\right]\right)$$

and $R(x, \mu, \sigma^2)$ and $R(y, \mu, \sigma^2)$ are different functions

i.e. $R(x, \mu, \sigma^2) = R(y, \mu, \sigma^2)$ as functions
of (μ, σ^2) iff $\sum x_i = \sum y_i$ and $\sum x_i^2 = \sum y_i^2$
i.e.

$$T(X) = (\sum X_i, \sum X_i^2)$$

is equivalent to the minimal sufficient statistic

$$R(X, \mu, \sigma^2)$$

Start reading section 1.6 on exponential families — They have many nice properties, among them is the fact that identifying a minimal sufficient statistic is easy

Handout