

Neyman Pearson Lemma

Jacques Amsel

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1 Statement

If a simple null and alternative hypothesis is used ($H_o : \theta = \theta_o$ vs $H_a : \theta = \theta_a$) with a specific α , any other test with $\alpha^* \leq \alpha$ will have $1 - \beta^* \leq LR$

$$LR = \frac{L(\theta_o)}{L(\theta_a)} < k$$

(Note: here, k is related to the rejection region of the test)

2 Proof

A test for $H_o : \theta = \theta_o$ vs $\theta = \theta_a$ means to use

$$d(y) = \begin{cases} 0 & H_o \text{ fails to be rejected} \\ 1 & H_o \text{ is rejected} \end{cases}$$

$$E[d(Y)|H_o \text{ true}]$$

$$\begin{aligned} &= 1 \cdot P(H_o \text{ rejected}|H_o \text{ true}) + 0 \cdot P(H_o \text{ fails to be rejected}|H_o \text{ true}) \\ &= \alpha \end{aligned}$$

$$E[d(Y)|H_a \text{ true}]$$

$$\begin{aligned} &= 1 \cdot P(H_o \text{ rejected}|H_a \text{ true}) + 0 \cdot P(H_o \text{ fails to be rejected}|H_a \text{ true}) \\ &= P(H_o \text{ rejected}|H_a \text{ true}) = 1 - P(H_o \text{ fails to be rejected}|H_a \text{ true}) \\ &= 1 - \beta \end{aligned}$$

let $d(Y)$ be associated with the LR test (i.e. $H_o : \theta = \theta_o$, etc),

$$d(Y) = 1 \Rightarrow \frac{L(\theta_o)}{L(\theta_a)} < k \Rightarrow L(\theta_o) < k \cdot L(\theta_a)$$

let $d^*(Y)$ be the decision function of any other test such that

$$\alpha^* = E[d^*(Y)|H_o \text{ true}] \leq E[d(Y)|H_o \text{ true}] = \alpha \quad (1)$$

We must show that

$$1 - \beta^* = E[d^*(Y)|H_a \text{ true}] \leq E[d(Y)|H_a \text{ true}] = 1 - \beta$$

$$\text{If } d(Y) = 1 \Rightarrow k \cdot L(\theta_a) > L(\theta_o) \Rightarrow 0 < k \cdot L(\theta_a) - L(\theta_o)$$

$$\text{If } d(Y) = 0 \Rightarrow k \cdot L(\theta_a) < L(\theta_o) \Rightarrow 0 > k \cdot L(\theta_a) - L(\theta_o)$$

$$d^*[k \cdot L(\theta_a) - L(\theta_o)] \leq d(Y)[k \cdot L(\theta_a) - L(\theta_o)] \quad (2)$$

START NOT IN PROOF:

Why?

If $d(Y) = 1$:

$$\frac{L(\theta_o)}{L(\theta_a)} < k \Rightarrow k \cdot L(\theta_a) - L(\theta_o) > 0 \text{ (proven previously)}$$

Since $d^*(Y) = 0$ or 1 and $d(Y) = 1$

$$d^*(Y) \cdot \text{positive} \leq d(Y) \cdot \text{positive}$$

If $d(Y) = 0$:

$$\frac{L(\theta_o)}{L(\theta_a)} > k \Rightarrow k \cdot L(\theta_a) - L(\theta_o) < 0 \text{ (proven previously)}$$

Since $d^*(Y) = 0$ or 1 and $d(Y) = 1$

$$d^*(Y) \cdot \text{negative} \leq d(Y) \cdot \text{negative}$$

END NOT IN PROOF

Integrate (2) on both sides dy .

$$\int (k \cdot d^*(y)L(\theta_a) - d^*(y)L(\theta_o))dy \leq \int (k \cdot d(Y)L(\theta_a) - d(Y)L(\theta_o))dy$$

Since $L(\theta_x) = f(y|\theta_x)$

$$\int (k \cdot d^*(y)f(y|\theta_a) - d^*(y)f(y|\theta_o))dy \leq \int (k \cdot d(Y)f(y|\theta_a) - d(Y)f(y|\theta_o))dy$$

$$k \int d^*(y)f(y|\theta_a)dy - \int d^*(y)f(y|\theta_o)dy \leq k \int d(Y)f(y|\theta_a)dy - \int d(Y)f(y|\theta_o)dy$$

$$kE[d^*(y)|H_a \text{ true}] - E[d^*(y)|H_o \text{ true}] \leq kE[d(Y)|H_a \text{ true}] - E[d(y)|H_o \text{ true}]$$

$$k(1 - \beta^*) - \alpha^* \leq k(1 - \beta) - \alpha$$

$$k(1 - \beta^*) - \alpha^* \leq k(1 - \beta) - \alpha$$

$$\alpha - \alpha^* \leq k(\beta^* - \beta)$$

Since, from (1), $\alpha \geq \alpha^*$

$$0 \leq k(\beta^* - \beta) \Rightarrow \beta \leq \beta^* \Rightarrow 1 - \beta \geq 1 - \beta^*$$