

Assignment - IV

T. Akash

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① $X \sim P(\lambda)$

$$H_0: \lambda = 1 \text{ vs } H_a: \lambda = 4$$

$$\phi(x) = \begin{cases} 1 & x > 2 \\ 0 & x \leq 2 \end{cases}$$

$P(\text{Type I error}) = \alpha = \text{Rejecting null when it is true.}$

$$P_{P=P_0}(\phi(x)=1) = P_{\lambda=1}(X>2) = 1 - P_{\lambda=1}(X \leq 2)$$

$$= 1 - P_{\lambda=1}(X=0) - P_{\lambda=1}(X=1) - P_{\lambda=1}(X=2)$$

$$\alpha = 1 - e^{-1} \left[1 + 1 + \frac{1}{2} \right] = 1 - \frac{5}{2e} = \underline{\underline{0.0803}}$$

$\beta = P(\text{Type II error}) = \text{Accepting null when it is false}$

$$\begin{aligned} P_{\lambda=4}(X \leq 2) &= \sum_{r=0}^2 \frac{e^{-\lambda} \lambda^r}{r!} = e^{-4} \left[\frac{4^0}{0!} + \frac{4^1}{1!} + \frac{4^2}{2!} \right] \\ &= e^{-4} [4 + 1 + 8] \end{aligned}$$

$$\frac{13}{e^4} = \underline{\underline{0.2381}}$$

①

$$(2) f_X(x) = \frac{1}{2\sigma} e^{-|x|/\sigma}$$

$$H_0: \sigma = 1 \text{ vs } H_a: \sigma > 1$$

$$\phi(x) = \begin{cases} 1 & |x| > 1 \\ 0 & |x| \leq 1 \end{cases}$$

$$\text{Size} = \alpha: P(\text{Type I error})$$

$$\Rightarrow P_{\sigma=1}(|x| > 1)$$

$$\Rightarrow \frac{1}{2} \left[\int_{-\infty}^{-1} e^x dx + \int_1^{\infty} e^{-x} dx \right] = \frac{1}{2} [2e^{-1}] = \underline{\underline{\frac{1}{e}}}$$

$$\text{Power: } P_{\theta}(X \in R) \quad \theta \in \Theta_1$$

$$\Rightarrow P_{\sigma > 1}(|x| > 1)$$

$$= \frac{1}{2\sigma} \left[\int_{-\infty}^{-1} e^{x/\sigma} dx + \int_1^{\infty} e^{-x/\sigma} dx \right]$$

$$= \frac{1}{2\sigma} \left[e^{x/\sigma} \Big|_{-\infty}^{-1} + e^{-x/\sigma} \Big|_1^{\infty} \right]$$

$$= \frac{1}{2} [2e^{-1/\sigma}] = \underline{\underline{e^{-1/\sigma}}}$$

$$\therefore \sigma > 1 \Rightarrow e^{-1/\sigma} > e^{-1} \quad \text{Hence proved.}$$

that Power > Size

③ $f_0(x) = \frac{\sigma}{\pi(\sigma^2 + x^2)}$ $H_0: \sigma = 1$ vs $H_a: \sigma = 2$

By NP Lemma Reject H_0 if $\frac{f_1(x)}{f_0(x)} \geq k$

$$\frac{f_1(x)}{f_0(x)} = \frac{\frac{\sigma_1}{\pi(\sigma_1^2 + x^2)}}{\frac{\sigma_0}{\pi(\sigma_0^2 + x^2)}} = \underbrace{2 \left[\frac{x^2 + 1}{x^2 + 4} \right]}_{Q(x)}$$

$Q'(x) > 0 \quad \forall x > 0 \Rightarrow$ Increasing function of x for $x > 0$
 $Q'(x) < 0 \quad \forall x < 0 \Rightarrow$ decreasing fn of x for $x < 0$
 $(\frac{x^2 + 1}{x^2 + 4})^2$

Max is attained at $x = \pm \infty \Rightarrow 2$

Min value $\Rightarrow x = 0 = \frac{1}{2}$

$k \leq \frac{1}{2} \rightarrow$ always reject H_0 & $\alpha = 1$

$k \geq 2 \rightarrow$ always accept H_0 & $\alpha = 0$

$\frac{1}{2} < k < 2 \rightarrow$ Reject H_0 if $Q(x) \geq k$

$$P_{\sigma=1} \left(\frac{2 \cdot x^2 + 1}{x^2 + 4} \geq k \right) = \alpha$$

By solving the size condition we get,

Reject H_0 if $|x| \geq \sqrt{\frac{4k-2}{2-k}} \Rightarrow k = \frac{4}{3 + 3 \cos(1-\alpha)}$

$$(4) \quad f_{\theta}(x) = \frac{2}{\theta^2} (1-x) \quad 0 \leq x \leq \theta$$

$$H_0: \theta = 1 \quad \text{vs} \quad H_a: \theta = 2$$

$$\frac{f_1(x)}{f_0(x)} > k \Rightarrow \frac{\frac{2}{\theta_1^2} (1-x)}{\frac{2}{\theta_0^2} (1-x)} > k \quad \text{--- (1)}$$

$$\Rightarrow \left(\frac{\theta_0}{\theta_1} \right)^2 (1-x) > k, \quad \theta_1 > \theta_0$$

$$\Rightarrow x > k'$$

Reject H_0 if $x > k'$, where k' is determined by
Size Condition

$$\begin{aligned} P_{\theta=1}(x > k) &= \int_k^1 2(1-x) dx \\ &= 2 \left[x - \frac{x^2}{2} \right] \Big|_k^1 \\ &\Rightarrow 2 \left[1 - \frac{1}{2} - \left[k - \frac{k^2}{2} \right] \right] \\ &\Rightarrow 1 - 2k + k^2 = (k-1)^2 \end{aligned}$$

$$(k-1)^2 = \alpha$$

$$\boxed{k' = 1 \pm \sqrt{\alpha}}$$

But in (1)

$$k' = \frac{4k-2}{4k-1} \leq 1 \Rightarrow k' = 1 - \sqrt{\alpha}$$

\therefore ~~Accept~~ H_0 when $x > 1 - \sqrt{\alpha}$
Reject

(4)

$$(5) \quad x_1, x_2, \dots, x_n \sim f_\theta(x) = \frac{1}{\Gamma(\theta)} x^{\theta-1} e^{-x}, \quad x > 0, \theta > 0$$

$$r(x) = \frac{f(x, \theta_1)}{f(x, \theta_2)}, \quad \theta_1 > \theta_2$$

$$r(x) = \frac{\left[\frac{1}{\Gamma(\theta_1)} \right]^n \prod_{i=1}^n x_i^{\theta_1-1} \left[e^{-\sum x_i} \right]}{\left[\frac{1}{\Gamma(\theta_2)} \right]^n \prod_{i=1}^n x_i^{\theta_2-1} \left[e^{-\sum x_i} \right]}$$

$r(x)$ is increasing function of $\prod_{i=1}^n x_i$.

Hence family has MLR in $(\theta, \prod_{i=1}^n x_i)$

UMP test for $H_0: \theta \leq 3$ vs $H_a: \theta > 3$

We know that $T(x) = \prod_{i=1}^n x_i$.

UMP test is given by,

$$\phi(x) = \begin{cases} 1 & \prod_{i=1}^n x_i > c \\ 0 & \prod_{i=1}^n x_i \leq c \end{cases}$$

Where c can be determined by Size Condition,

$$E_{\theta_0}(\phi(x)) = \alpha.$$

$$(6) \quad X_1, X_2, X_3, \dots, X_n \sim \text{Exp}(\lambda)$$

$$\text{UMP test } H_0: \lambda = 1 \text{ vs } H_a: \lambda \neq 1$$

$$f(x, \lambda) = \lambda^n e^{-\lambda \sum_{i=1}^n x_i}$$

We know that exponential distribution belongs to exponential family with $T(x) = \sum_{i=1}^n x_i$

$$\therefore \phi(x) = \begin{cases} 1 & \sum x_i < c_1 \text{ or } \sum x_i > c_2 \\ 0 & c_1 \leq \sum x_i \leq c_2 \end{cases}$$

$$E_{\lambda_0}(\phi(x)) = \alpha \quad (\lambda_0 = 1) \quad W = \sum x_i \sim \text{gamma}(n, 1)$$

$$E_{\lambda_0}(1 - \phi(x)) = 1 - \alpha$$

$$P_{\lambda_0}(c_1 \leq \sum x_i \leq c_2) = 1 - \alpha$$

$$\int_{c_1}^{c_2} \frac{w^{n-1} e^{-w}}{\Gamma(n)} dw = 1 - \alpha \quad \text{--- (1)}$$

$$E_{\lambda_0}[T(x) \phi(x)] = \alpha E_{\lambda_0}[T(x)]$$

$$E_{\lambda_0}[W(1 - \phi(x))] = (1 - \alpha) E_{\lambda_0}(W)$$

[Mean of gamma = $n\lambda$]

$$\int_{c_1}^{c_2} w \cdot f_{n,1}(w) dw = (1 - \alpha) n$$

$$= n(1) \\ = n$$

(7)

$$w g_{n,1} dw = w \cdot \frac{w^{n-1} e^{-w} \cdot n}{\Gamma(n) \cdot n} = n \cdot \frac{w^n e^{-w}}{\Gamma(n+1)}$$

$$\boxed{w g_{n,1}(w) = n \cdot g_{n+1,1}(w)}$$

$$\Rightarrow \int_{c_1}^{c_2} n \cdot g_{n+1,1}(w) dw = (1-\alpha)$$

$$\Rightarrow \int_{c_1}^{c_2} g_{n+1,1}(w) dw = 1-\alpha$$

$$\Rightarrow \frac{1}{\Gamma(n+1)} \int_{c_1}^{c_2} w^n e^{-w} dw = 1-\alpha$$

$$\Rightarrow \frac{1}{\Gamma(n+1)} \left[w^n e^{-w} \Big|_{c_2}^{c_1} + \underbrace{n \int_{c_1}^{c_2} w^{n-1} e^{-w} dw}_{(1)} \right] = 1-\alpha$$

$$c_1^n e^{-c_1} - c_2^n e^{-c_2} + n \Gamma(n) (1-\alpha) = \Gamma(n+1) [1-\alpha]$$

$$\Rightarrow \boxed{c_1^n e^{-c_1} = c_2^n e^{-c_2}}$$

By solving this equation we would get our c_1, c_2 .

$$(7) f_{\theta}(x) = \frac{2}{\theta^2} (\theta - x) \quad 0 < x < \theta \quad \text{LRT for } H_0: \theta = 2 \text{ vs } H_a: \theta \neq 2.$$

$$L(\theta, x) = \frac{2}{\theta^2} (\theta - x)$$

$$\log L = -2 \log(\theta) + 2 \log 2 + \log(\theta - x)$$

$$\frac{\partial}{\partial \theta} (\log(L(\theta))) = -\frac{2}{\theta} + \frac{1}{\theta - x} \leq 0$$

$$\Rightarrow 2x = \theta$$

$$L(\theta_L, x) = \frac{2}{(2x)^2} (2x - x) = \frac{1}{2x}$$

$$L(\theta_{LH}, x) = \frac{2}{4} (2 - x) = \frac{2-x}{2}$$

$$\lambda(x) = \frac{L(\theta_{LH}, x)}{L(\theta_L, x)} = \frac{\frac{2-x}{2}}{\frac{1}{2x}} = x(2-x)$$

According to LRT Reject H_0 if $\lambda(x) \leq c$.

$$x(2-x) \leq c$$

$$\Rightarrow (x-1)^2 \geq 1-c$$

$$\left. \begin{array}{l} \text{if } 0 < x \leq 1 \rightarrow x \leq 1 - \sqrt{1-c} \\ 1 \leq x \leq 2 \rightarrow x \geq 1 + \sqrt{1-c} \end{array} \right\} \rightarrow (1)$$

We know that

$$\sup P_0(\lambda(x) \leq c) = \alpha$$

$$\Rightarrow \int_0^{1-\sqrt{1-c}} \frac{1}{2}(2-x) dx + \int_{1+\sqrt{1-c}}^2 \frac{1}{2}(2-x) dx = \alpha$$

$$x - \frac{x^2}{4} \Big|_0^{1-\sqrt{1-c}} + x - \frac{x^2}{4} \Big|_{1+\sqrt{1-c}}^2 = \alpha$$

$$\Rightarrow 1 - \sqrt{1-c} = \alpha \quad \text{--- (2)}$$

Substituting (2) in (1) we get,

$$x \leq 1 - \sqrt{1-c} \Rightarrow x \leq \alpha$$

$$x > 1 + \sqrt{1-c} \Rightarrow x > 2 - \alpha$$

\therefore Reject H_0 if $x \leq \alpha$ or $x > 2 - \alpha$ else accept.

$$(8) \quad f(x) = \frac{1}{2\sigma} e^{-\left(\frac{|x|}{\sigma}\right)}$$

$$L(\sigma, x) = \left(\frac{1}{2\sigma}\right)^n e^{-\sum_{i=1}^n |x_i|/\sigma}$$

$$H_0: \sigma = 1 \quad \text{vs} \quad H_a: \sigma \neq 1$$

$$\log(L(\sigma, x)) = -n \log(2\sigma) - \sum \frac{|x_i|}{\sigma}$$

$$\frac{\partial \log(L)}{\partial \sigma} = -\frac{n}{2\sigma} \cdot 2 + \sum \frac{|x_i|}{\sigma^2} = 0$$

$$\Rightarrow \left[\frac{1}{\sigma} = \frac{\sum_{i=1}^n |x_i|/n}{\sigma^2} \right]$$

$$\Rightarrow \text{⑧ } L(\hat{\sigma}) = \frac{1}{(\frac{1}{2\hat{\sigma}})^n} \cdot e^{-\frac{\sum |x_i|}{\sum |x_i|/n}} = \frac{1}{(2\hat{\sigma})^n} e^{-n}$$

$$\hat{L}(\Lambda_H) = \frac{1}{2^n} e^{-\sum |x_i|} \quad [\sigma=1 \text{ in Null Hypothesis}]$$

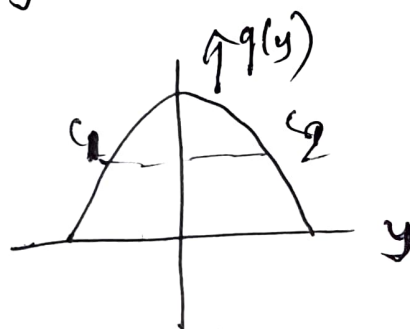
$$\lambda(x) = \frac{\hat{L}(\Lambda_H)}{\hat{L}(\Lambda)} = \frac{\frac{1}{2^n} \cdot e^{-\sum |x_i|}}{\left(\frac{1}{2\hat{\sigma}}\right)^n e^{-n}} = (\hat{\sigma})^n e^{n - \sum |x_i|}$$

$$\text{Let } y = \sum |x_i|$$

$$\lambda(y) = \frac{y^n}{n^n} e^{n-y}$$

$$g(y) = y^n e^{-y}, \quad g'(y) = ny^{n-1} e^{-y} - y^n e^{-y}$$

$$\Rightarrow e^{-y} \cdot y^{n-1} [n-y] \begin{cases} > 0 & y < n \\ < 0 & y > n \end{cases}$$



$$\Rightarrow g(y) < 0 \Rightarrow y < c_1 \text{ or } y > c_2$$

$$\sup_{\Lambda_H} [P(y < c_1) \text{ or } P(y > c_2)] = \alpha$$

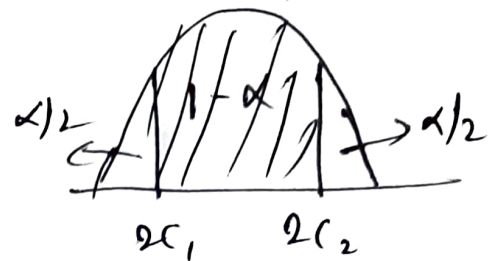
$$1 - \alpha = P_{\sigma=1}(c_1 \leq y \leq c_2)$$

$$1 - \alpha = P_{\sigma=1}\left(\frac{2c_1}{\sigma} \leq \frac{2y}{\sigma} \leq \frac{2c_2}{\sigma}\right) \quad \left[\frac{2\sum |x_i|}{\sigma} \sim \chi^2_{2n}\right]$$

$$\Rightarrow 1 - \alpha = P(2c_1 \leq W \leq c_2) \quad [\because \sigma=1, W = \frac{2\sum |x_i|}{\sigma}]$$

$$\Rightarrow 2C_2 = \chi^2_{2n, \alpha/2} \Rightarrow C_2 = \frac{\chi^2_{2n, \alpha/2}}{2}$$

$$2C_1 = \chi^2_{2n, 1-\alpha/2} \Rightarrow C_1 = \frac{\chi^2_{2n, 1-\alpha/2}}{2}$$



\therefore Reject H_0 if $y > C_2$ or $y < C_1$

$$\Rightarrow \sum |x_i| > \frac{\chi^2_{2n, \alpha/2}}{2} \quad (\text{or}) \quad \sum |x_i| < \frac{\chi^2_{2n, 1-\alpha/2}}{2} \quad [y = \sum |x_i|]$$