

Example (Continuation from last lecture)

$X_1, \dots, X_n \sim N(\mu, 36)$, $H_0: \mu = 50$ vs $H_1: \mu = 55$.

b) For $n=16$, $G: \bar{X} \geq 53$, find α .

Solution

$$\alpha = P(\text{Type I error}) = P(\bar{X} \geq 53 \mid H_0) = P\left(\frac{\bar{X}-50}{\sqrt{36/16}} > \frac{53-50}{\sqrt{36/16}}\right) =$$

$$= P(Z > 2) = 1 - P(Z \leq 2) = 1 - \Phi(2) = 0.0228$$

Example

Let X_1, \dots, X_n r.s of $N(\theta, 1)$. $H_0: \theta = 0$ vs $H_1: \theta = -1$

Show that if $n=25$, $\alpha=0.05$, then the power of the test is 0.999

Solution

$$L(\theta) = \prod_{i=1}^n \lambda(x_i; \theta) = (2\pi)^{-n/2} \cdot \exp\left\{-\frac{1}{2} \sum_{i=1}^n (x_i - \theta)^2\right\}$$

$$\frac{L_0}{L_1} = \frac{L(\theta_0)}{L(\theta_1)} = \frac{(2\pi)^{-n/2} \cdot \exp\left\{-\frac{1}{2} \sum_{i=1}^n (x_i - 0)^2\right\}}{(2\pi)^{-n/2} \cdot \exp\left\{-\frac{1}{2} \sum_{i=1}^n (x_i + 1)^2\right\}} = \exp\left\{-\frac{1}{2} \left[\sum_{i=1}^n x_i^2 - \sum_{i=1}^n (x_i + 1)^2 \right]\right\}$$

$$= \exp\left\{-\frac{1}{2} \left[2\sum x_i^2 - 2\sum x_i - n \right]\right\} = \exp\left\{\frac{n}{2} + \sum x_i\right\}$$

$$\text{so } \frac{L_0}{L_1} \leq k \iff \exp\left\{\frac{n}{2} + \sum x_i\right\} \leq k \iff \frac{n}{2} + \sum x_i \leq \log k \iff$$

$$\iff \bar{X} \leq \frac{\log k}{n} - \frac{1}{2} = w. \text{ So } G = \{(x_1, \dots, x_n) : \bar{X} \leq w\}$$

$$\begin{aligned} P(G \mid H_0) &= P(\bar{X} \leq w \mid \theta=0) = P\left(\frac{\bar{X}-0}{\sqrt{1/25}} \leq \frac{w-0}{\sqrt{1/25}}\right) = 0.05 \\ P(Z \leq -Z_{0.05}) &= 0.05 \end{aligned} \Rightarrow$$

$$\Rightarrow 5w = -Z_{0.05} \Rightarrow w = -0.329$$

Power Function: $\Pi(\theta) = P(G \mid \theta) = P_\theta(G) \underbrace{\underbrace{\quad}_{H_1} \quad}_{1-\theta}$

$$\Pi(\theta_1) = P_{\theta_1}(G) = P(\bar{X} \leq -0.329 \mid \theta = -1) =$$

$$= P\left(\frac{\bar{X}+1}{\sqrt{1/25}} \leq \frac{-0.329+1}{\sqrt{1/25}}\right) = P(Z \leq 3.3) = \Phi(3.3) = 0.999$$

Generalization

H_0 and H_1 do not necessarily belong to the same distribution, nor are the random variables X_1, \dots, X_v independent. Hence, if H_0 is the simple hypothesis that the joint PDF of the sample is $g(\underline{x})$ and H_1 is the simple hypothesis that the joint PDF is $h(\underline{x})$, then for the m.o.c.r. G of the test H_0 vs H_1 we have:

$$i) \frac{g(\underline{x})}{h(\underline{x})} \leq K \text{ for } (x_1, \dots, x_v) \in G$$

$$ii) \frac{g(\underline{x})}{h(\underline{x})} > K \text{ for } (x_1, \dots, x_v) \notin G$$

$$iii) P((x_1, \dots, x_v) \in G | H_1) = \alpha$$

Example

Let X_1, \dots, X_v r.s. with PDF $f(x) > 0$, $x = 0, 1, \dots$

Test: $H_0: g(x) = \frac{e^{-1}}{x!}, x = 0, 1, \dots$

$H_1: h(x) = (\frac{1}{2})^{x+1}, x = 0, 1, \dots$

$$\text{We have: } \frac{g(\underline{x})}{h(\underline{x})} = \frac{\prod_{i=1}^v e^{-1}/x_i!}{\prod_{i=1}^v (\frac{1}{2})^{x_i+1}} = \frac{(2e^{-1})^v \cdot 2^{\sum x_i}}{\prod_{i=1}^v x_i!} \leq K \Leftrightarrow$$

$$\Leftrightarrow v \log(2e^{-1}) + \sum_{i=1}^v x_i \cdot \log 2 - \log \prod_{i=1}^v x_i! \leq \log K \Leftrightarrow$$

$$\Leftrightarrow \sum_{i=1}^v x_i \cdot \log 2 - \sum_{i=1}^v \log(x_i!) \leq \log K - v \log(2e^{-1})$$

Application: $K = 1, v = 1$

$$\text{We have } \frac{g(x)}{h(x)} < 1 \Leftrightarrow \frac{2^{x_1}}{x_1!} \leq \frac{1}{2}, \text{ so } G = \{X_1 : X_1 = 0, 3, 4, 5, \dots\}$$

$$\begin{aligned} \alpha &= P(\text{Type I error}) = P(G | H_0) = 1 - P(G' | H_0) = 1 - P(X_1 = 1, 2 | H_0) = \\ &= 1 - e^{-1} - \frac{e^{-2}}{2!} = 0,445 \end{aligned}$$

$$\begin{aligned} \text{The power of the test is } P(G | H_1) &= 1 - P(G' | H_1) = \\ &= 1 - P(X_1 = 0, 1 | H_1) = 1 - \frac{1}{2^2} - \frac{1}{2^3} = 0,625 \end{aligned}$$

Simple vs Complex

Definition

If we are testing $H_1: \theta = \theta_0$ (simple) versus a complex H_1 , the test is called uniformly most powerful test (u.m.p.t) if it is the most powerful test versus every simple alternative.

Example (One-sided test)

Let X_1, \dots, X_n r.s. $\sim N(\theta, \sigma^2)$, $\theta > 0$. Prove that there exists a u.m.p.t. with given error α for $H_0: \theta = \theta_0$ vs $H_1: \theta > \theta_0$.

Solution

Suppose $\theta_1 > \theta_0$. We test $H_0: \theta = \theta_0$ vs $H_1: \theta = \theta_1$

$$L(\theta) = \prod_{i=1}^n f(x_i; \theta) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\theta}} e^{-\frac{1}{2\theta}x_i^2} = (2\pi\theta)^{-n/2} \exp\left\{-\frac{1}{2\theta} \sum_{i=1}^n x_i^2\right\}$$

$$\text{N-P Lemma: } \frac{\theta_0}{\theta_1} \leq K \iff \frac{(2\pi\theta_0)^{-n/2} \exp\left\{-\frac{1}{2\theta_0} \sum_{i=1}^n x_i^2\right\}}{(2\pi\theta_1)^{-n/2} \exp\left\{-\frac{1}{2\theta_1} \sum_{i=1}^n x_i^2\right\}} =$$

$$= \left(\frac{\theta_0}{\theta_1}\right)^{n/2} \exp\left\{-\frac{1}{2\theta_0} \sum_{i=1}^n x_i^2 - \frac{1}{2\theta_1} \sum_{i=1}^n x_i^2\right\} = \left(\frac{\theta_0}{\theta_1}\right)^{n/2} \exp\left\{-\frac{(\theta_1 - \theta_0)}{2\theta_1\theta_0} \sum_{i=1}^n x_i^2\right\} \leq K$$

$$\iff \frac{n}{2} \log \frac{\theta_0}{\theta_1} - \frac{\theta_1 - \theta_0}{2\theta_1\theta_0} \sum_{i=1}^n x_i^2 \leq \log K \iff$$

$$\iff \sum_{i=1}^n x_i^2 \geq \frac{2\theta_0\theta_1}{\theta_1 - \theta_0} \left[\frac{n}{2} \log\left(\frac{\theta_0}{\theta_1}\right) - \log K \right] = w$$

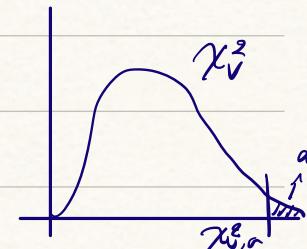
So $G = \{(x_1, \dots, x_n) : \sum_{i=1}^n x_i^2 \geq w\}$ and we need:

$$\begin{aligned} P((x_1, \dots, x_n) \in G \mid H_0) &= \alpha \iff X_i \sim N(0, \theta) \iff \frac{x_i}{\sqrt{\theta}} \sim N(0, 1) \\ \iff \frac{\sum x_i^2}{\theta} &\sim \chi_1^2 \iff Y = \frac{\sum x_i^2}{\theta} = \frac{\sum x_i^2}{\theta_0} \sim \chi_v^2 \end{aligned}$$

and so $P(G \mid H_0) = P(\sum x_i^2 \geq w \mid H_0) = \alpha \iff$

$$\iff P\left(\frac{\sum x_i^2}{\theta} \geq \frac{w}{\theta} \mid H_0\right) = \alpha \iff P\left(\frac{\sum x_i^2}{\theta_0} \geq \frac{w}{\theta_0} \mid H_0\right) = \alpha \iff$$

$$\left. \begin{aligned} \Rightarrow P(Y \geq \frac{w}{\theta_0}) &= \alpha \\ P(Y \geq \chi_{v,\alpha}^2) &= \alpha \end{aligned} \right\} \Rightarrow \frac{w}{\theta_0} = \chi_{v,\alpha}^2 \Rightarrow w = \theta_0 \cdot \chi_{v,\alpha}^2$$



Finally, $G = \{(x_1, \dots, x_v) : \sum_{i=1}^v x_i^2 \geq \theta_0 \chi_{v,a}^2\}$

The c.r. G is independent of θ_1 , hence the test is the u.m.g.t.