

Theorem 8.2.4 If $T(\underline{X})$ is a sufficient statistic for Θ , $\lambda(\underline{x})$ is the LRT statistic based on \underline{X} , and $\lambda^*(T(\underline{x}))$ is the LRT statistic based on $T(\underline{x})$, then $\lambda^*(T(\underline{x})) = \lambda(\underline{x}) \quad \forall \underline{x}$ in the sample space.

proof: see textbook

Remark: Using the Factorization Theorem, $\lambda(\underline{x})$ and hence the rejection region should depend on \underline{x} only through a sufficient statistic.

Ex: $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$, $-\infty < \mu < \infty$, σ^2 known, $(H) = \mathbb{R}$.

$$H_0: \mu = \mu_0$$

$$(H)_0 = \{\mu_0\}$$

$$H_1: \mu \neq \mu_0$$

$$(H)_1 = \mathbb{R} - \{\mu_0\}$$

$$L(\mu | \underline{x}) = e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2}$$

Recall that the MLE of $\mu = \bar{X}$ which is a sufficient statistic for μ .

Then

$$\begin{aligned} \lambda(\underline{x}) &= \frac{e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu_0)^2}}{e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \bar{x})^2}} \\ &= e^{-\frac{n}{2\sigma^2} (\bar{x} - \mu_0)^2} \end{aligned}$$

For some $C \in (0, 1)$, the LRT rejects H_0 if $e^{-\frac{n}{2\sigma^2} (\bar{x} - \mu_0)^2} \leq C$

$$\Leftrightarrow (\bar{x} - \mu_0)^2 \geq -\frac{2\sigma^2}{n} \log C.$$

Hence reject H_0 if $|\bar{x} - \mu_0| \geq \sqrt{-\frac{2\sigma^2}{n} \log C}$

$$\Leftrightarrow \bar{x} \geq \mu_0 + \sqrt{-\frac{2\sigma^2}{n} \log C} \quad \text{or} \quad \bar{x} \leq \mu_0 - \sqrt{-\frac{2\sigma^2}{n} \log C}.$$

Bayesian Tests (Section 8.2.2)

$\pi(\theta)$: prior distribution of $\theta \in \Theta$

$f(\underline{x}|\theta)$: joint distribution of sample

Combine prior on θ and sample distribution: to form posterior distribution of θ given \underline{x} .

where $\pi(\theta|\underline{x}) \propto L(\theta|\underline{x}) \pi(\theta)$.

Hypothesis test: $H_0: \theta \in \Theta_0$

$H_1: \theta \in \Theta_1 = \Theta_0^C$

In the Bayesian framework, the posterior distribution $\pi(\theta|\underline{x})$ is used to calculate the probabilities that H_0 and H_1 are true.

$$P(H_0 \text{ is true} | \underline{x}) = P(\theta \in \Theta_0 | \underline{x}) = \int_{\theta \in \Theta_0} \pi(\theta|\underline{x}) d\theta$$

$$P(H_1 \text{ is true} | \underline{x}) = P(\theta \in \Theta_1 | \underline{x}) = 1 - P(\theta \in \Theta_0 | \underline{x}) = 1 - P(H_0 \text{ is true} | \underline{x}),$$

Ex: $X_1, \dots, X_n \sim \text{Bernoulli}(\theta)$, $\Theta = \{\frac{1}{4}, \frac{3}{4}\}$.

$$H_0: \theta = \frac{1}{4}$$

$$H_1: \theta = \frac{3}{4}$$

$$\text{Suppose } \pi(\theta) = \begin{cases} a, & \theta = \frac{1}{4} \\ 1-a, & \theta = \frac{3}{4} \end{cases} \quad \text{for } 0 \leq a \leq 1.$$

$$\text{i.e., } P(\theta \in \Theta_0) = a, \quad P(\theta \in \Theta_1) = 1-a,$$

$$\pi(\theta|\underline{x}) \propto \theta^{\sum x_i} (1-\theta)^{n-\sum x_i} \pi(\theta) = (1-\theta)^n \left(\frac{\theta}{1-\theta}\right)^{\sum x_i} \cdot \pi(\theta)$$

$$= \begin{cases} a \left(\frac{3}{4}\right)^n \left(\frac{1}{3}\right)^s, & \theta = \frac{1}{4} \\ (1-a) \left(\frac{1}{4}\right)^n 3^s, & \theta = \frac{3}{4} \end{cases}$$

$$= \begin{cases} a \left(\frac{1}{4}\right)^n 3^{n-s}, & \theta = \frac{1}{4} \\ (1-a) \left(\frac{1}{4}\right)^n 3^s, & \theta = \frac{3}{4} \end{cases}$$

One can show that

$$\text{i.e., } P(\theta \in H_0 | X) = \frac{a}{a + (1-a) 3^{2s-n}}, \quad s = 0, 1, \dots, n$$

$$P(\theta \in H_1 | X) = \frac{(1-a) 3^{2s-n}}{a + (1-a) 3^{2s-n}}, \quad s = 0, 1, \dots, n$$

Some Options for rejecting H_0 :

1. For given values of a , n , and $s = \sum_{i=1}^n x_i$, one could reject H_0 if

$$\frac{P(\theta \in H_1 | X)}{P(\theta \in H_0 | X)} = \frac{(1-a) 3^{2s-n}}{a} = \left(\frac{1-a}{a}\right) 3^{n(\bar{x}-1)} > K \quad \text{for some } K \geq 1$$

$$\hat{\theta} = \bar{x} = \frac{s}{n} = \frac{\sum_{i=1}^n x_i}{n}$$

$$\Leftrightarrow \text{reject } H_0 \text{ for } \bar{x} > \frac{1}{2} \left(\frac{1}{n} \frac{\log(K \frac{a}{1-a})}{\log 3} + 1 \right)$$

2. Reject H_0 if $P(\theta \in H_1 | X) > \frac{1}{2}$.

3. To guard against falsely rejecting H_0 , however, consider rejecting H_0

if $P(\theta \in H_1 | X) > K$ for some $K \gg \frac{1}{2}$.

Ex: $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} f(x_i | \theta)$ when $f(x_i | \theta) = \frac{1}{\theta} e^{-x_i/\theta} I_{(0, \infty)}(x_i)$, $\theta > 0$.

Thus $H = (0, \infty)$.

$H_0: \theta \leq \theta_0$

$H_1: \theta > \theta_0$

Suppose $\pi(\theta) = \frac{1}{\Gamma(\alpha) \beta^\alpha} \theta^{-(\alpha+1)} e^{-1/(\beta\theta)} I_{(0, \infty)}(\theta)$, i.e., $\theta \sim \text{Inverse Gamma}(\alpha, \beta)$ with α, β known.

$$\text{Then } \pi(\theta | X) \propto \theta^{-n} e^{-\frac{n\bar{x}}{\theta}} \cdot \pi(\theta)$$

$$\text{so } \pi(\theta | X) \sim \text{Inverse Gamma}(\alpha+n, (n\bar{x} + 1/\beta)^{-1})$$

$$\text{Then } P(\theta \in H_1 | X) = \int_{\theta_0}^{\infty} \pi(\theta | X) d\theta$$

EXTRA:

$$L(\theta | X) \propto \frac{1}{\theta^n} e^{-\frac{1}{\theta} \sum_{i=1}^n x_i} I_{(0, \infty)}(x) = \theta^{-n} e^{-\frac{n\bar{x}}{\theta}} I_{(0, \infty)}(x)$$

$$\pi(\theta | X) \propto \theta^{-n} e^{-\frac{n\bar{x}}{\theta}} \cdot \theta^{-(\alpha+1)} e^{-\frac{1}{\beta\theta}} I_{(0, \infty)}(\theta) = \theta^{-(n+\alpha+1)} e^{-\frac{1}{\theta} (n\bar{x} + 1/\beta)} I_{(0, \infty)}(\theta)$$