

Tutorial 8

Rui Qiu

2018-04-28

Q4.1

Use the Neyman-Pearson lemma to find the form of the critical region for the best test of H_0 against H_1 when

(c)

X_1, X_2, \dots, X_n are a random sample from the exponential distribution with p.d.f. $f(x; \theta) = \theta e^{-\theta x}$, $x > 0$ and $H_0 : \theta = \theta_0, H_1 : \theta = \theta_1, \theta_1 > \theta_0$.

Solution:

$$\begin{aligned} L(\theta) &= \prod_{i=1}^n \theta \cdot e^{-\theta x_i} \\ &= \theta^n \cdot e^{-\theta \sum_{i=1}^n x_i} \\ \lambda &= \frac{L(\theta_0 | \mathbf{x})}{L(\theta_1 | \mathbf{x})} \\ &= \frac{\theta_0^n \cdot e^{-\theta_0 \sum_{i=1}^n x_i}}{\theta_1^n \cdot e^{-\theta_1 \sum_{i=1}^n x_i}} \\ &= \exp((\theta_1 - \theta_0) \sum x_i) \cdot \left(\frac{\theta_0}{\theta_1} \right)^n \\ C &= \{\lambda \leq k\} \\ &= \{(\theta_1 - \theta_0) \cdot \sum x_i + n \log \left(\frac{\theta_0}{\theta_1} \right) \leq \log(k)\} \\ &= \left\{ \sum x_i \leq \frac{\log(k) + n \log \left(\frac{\theta_0}{\theta_1} \right)}{\theta_1 - \theta_0} \right\} \end{aligned}$$

(d)

$X_{11}, X_{12}, \dots, X_{1n_1} \sim N(\mu_1, \sigma_1^2), X_{21}, X_{22}, \dots, X_{2n_2} \sim N(\mu_2, \sigma_2^2)$, all X_{ij} are independent of each other, σ_1^2, σ_2^2 are known, and $H_0 : \mu_2 = \mu_1, H_1 : \mu_2 = \mu_1 + \delta$ with $\delta > 0$ (μ_1 and δ are both known constants).

Solution:

The following questions are written by hands, apologies for the inconsistency.

Tutorial 8.

Q4(c).

$$L(\theta) = \prod_{i=1}^n \theta \cdot e^{-\theta x_i} = \theta^n \cdot e^{-\theta \sum_{i=1}^n x_i}$$

$$\lambda = \frac{L(\theta_0)}{L(\theta_1)} = \frac{\theta_0^n \cdot e^{-\theta_0 \sum x_i}}{\theta_1^n \cdot e^{-\theta_1 \sum x_i}} = \exp((\theta_1 - \theta_0) \cdot \sum x_i) \cdot \left(\frac{\theta_0}{\theta_1}\right)^n$$

$$C = \{\lambda \leq k\}$$

$$= \left\{ (\theta_1 - \theta_0) \cdot \sum x_i + n \log\left(\frac{\theta_0}{\theta_1}\right) \leq \log k \right\}$$

$$= \left\{ \sum x_i \leq \frac{\log k + n \log\left(\frac{\theta_0}{\theta_1}\right)}{\theta_1 - \theta_0} \right\}$$

Q4(c.d).

$$L(\mu_1, \sigma_1^2, \mu_2, \sigma_2^2, \vec{x}_1, \vec{x}_2) = \left(\frac{1}{\sqrt{2\pi\sigma_1^2}}\right)^{n_1} \left(\frac{1}{\sqrt{2\pi\sigma_2^2}}\right)^{n_2} e^{-\frac{\sum_{i=1}^{n_1} (x_{1i} - \mu_1)^2}{2\sigma_1^2} - \frac{\sum_{j=1}^{n_2} (x_{2j} - \mu_2)^2}{2\sigma_2^2}}$$

$$\lambda = \frac{L(\mu_1 = \mu_2, \sigma_1^2, \sigma_2^2, \vec{x}_1, \vec{x}_2)}{L(\mu_1 = \mu_2 + \delta, \sigma_1^2, \sigma_2^2, \vec{x}_1, \vec{x}_2)} = \exp\left(\frac{1}{2\sigma_2^2} \left(-2\delta \sum x_{2j} + 2n_2\mu_1\delta + n_2\delta^2\right)\right)$$

$$C = \{\lambda(x) \leq k\}$$

$$= \left\{ \frac{1}{2\sigma_2^2} (-2\delta \sum x_{2j} + 2n_2\mu_1\delta + n_2\delta^2) \leq \log k \right\}$$

$$= \left\{ \sum x_{2j} > \frac{2\sigma_2^2 \log k - n_2(2\mu_1\delta + \delta^2)}{-2\delta} \right\}$$

$$= \left\{ \sum x_{2j} \geq \frac{\sigma_2^2 \log k}{\delta} + \frac{n_2(2\mu_1\delta + \delta^2)}{2\delta} \right\}$$

$$= \left\{ \sum x_{2j} \geq \frac{\sigma_2^2}{\delta} \log k + \frac{n_2\mu_1}{2} + \frac{n_2\delta}{2} \right\}$$

Q4.2. When $\sigma_1^2 = \sigma_2^2 = \sigma^2 = 1$, $n_1 = n_2 = n$

$$(a). C = \left\{ \sum X_i \geq -\log k + n\mu_1 + \frac{n}{2} \right\}$$

$$= \left\{ \bar{X} \geq -\frac{\log k}{n} + \mu_1 + \frac{1}{2} \right\}$$

$$P_{H_0}(C) = P_{H_0}(\bar{X} \geq k^{**}) = \alpha = 0.01$$

$$\Rightarrow P_{H_0}\left(\frac{\bar{X} - \mu_1}{1/\sqrt{n}} \geq k^{**}\right) = 0.01$$

$$\Rightarrow P_{H_0}(\bar{Z} \geq k^{**}) = 0.01$$

$$\text{so } k^{**} = 2.326348$$

Normal(0, 1)

$$\text{g}(\mu) = \text{Power} = P_{H_1}(C) = P_{H_1}(\bar{X} \geq k^{**})$$

$$= P_{H_1}\left(\bar{X} \geq k^{**} \cdot \frac{1}{\sqrt{n}} + \mu_1\right)$$

$$= P_{H_1}\left(\bar{X} \geq 2.326348 \cdot \frac{1}{\sqrt{10}} + \mu_1\right)$$

$$= P_{H_1}\left(\frac{\bar{X} - \mu_1 - 1}{\frac{1}{\sqrt{10}}} \geq \left(\frac{2.326348 \cdot \frac{1}{\sqrt{10}} + \mu_1 - \mu_1 - 1}{\frac{1}{\sqrt{10}}}\right)\right)$$

$$= P_{H_1}\left(\frac{\bar{X} - \mu_1 - 1}{\frac{1}{\sqrt{10}}} \geq 2.326348 - \sqrt{10}\right)$$

$$= 1 - P(\bar{Z} < 2.326348 - \sqrt{10})$$

$$= 0.7948$$

$$\frac{k^{**} - \mu_1}{1/\sqrt{n}} = k^{**}$$

$$k^{**} = k^{**} \cdot \frac{1}{\sqrt{n}} + \mu_1$$

(b).

$$\text{g}(\mu) = 1 - P(\bar{Z} < 2.326348 - \sqrt{n}) \geq 0.95$$

$$P(\bar{Z} < 2.326348 - \sqrt{n}) \leq 0.05$$

$$2.326348 - \sqrt{n} \leq -1.644854$$

$$\sqrt{n} \geq 15.77045$$

$$n = 16.$$

$$Q4.3: L(\vec{x}; \theta) = \frac{\theta^n}{\Gamma(n)} x^{(1-1)n} e^{-\theta \sum x_i}$$

$$\lambda = \frac{L(\theta_0)}{L(\theta_1)} = \left(\frac{\theta_0}{\theta_1}\right)^n e^{(\theta_1 - \theta_0)\sum x_i}$$

$$C = \{ \lambda \leq c \}$$

$$= \left\{ \sum x_i \leq \frac{1}{\theta_1 - \theta_0} (\log c - n \log \left(\frac{\theta_0}{\theta_1}\right)) \right\}$$

$$\text{When } n=\lambda=1, C = \left\{ \sum x_i \leq \frac{1}{\theta_1 - \theta_0} (\log c - \log \left(\frac{\theta_0}{\theta_1}\right)) \right\}$$

$$= \left\{ x \leq \underbrace{\frac{1}{(\theta_1 - \theta_0)} (\log c - \log \left(\frac{\theta_0}{\theta_1}\right))}_{c^*} \right\}$$

$$P_{H_0}(C) = P_{H_0}(x \leq c^*)$$

$$= \int_0^{c^*} \frac{\theta_0}{\Gamma(1)} e^{-\theta_0 x} dx$$

$$= \theta_0 - \frac{1}{\theta_0} e^{-\theta_0 x} \Big|_0^{c^*}$$

$$= -1(e^{-\theta_0 c^*} - 1)$$

$$= 1 - e^{-\theta_0 c^*} = \alpha$$

$$c^* = \frac{\log(1-\alpha)}{-\theta_0}$$

$$P_{H_1}(C) = P_{H_1}(x \leq c^*)$$

$$= \int_0^{c^*} \theta_1 \cdot e^{-\theta_1 x} dx$$

$$= 1 - e^{-\theta_1 c^*}$$

$$= 1 - (1-\alpha)^{\theta_1/\theta_0}$$