

8. Let $X_1, X_2, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$ where both μ, σ^2 are unknown.

Test: $H_0: \mu = \mu_0, 0 < \sigma^2 < \infty$ vs $H_1: \mu \neq \mu_0, 0 < \sigma^2 < \infty$

Use the Likelihood ratio test to obtain the BCR.

$X_1, X_2, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$, find the MLE of μ and σ^2 .

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}, \quad -\infty < x < \infty.$$

$$f(x_i) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x_i-\mu}{\sigma}\right)^2}, \quad -\infty < x_i < \infty \quad \forall i.$$

$$\begin{aligned} \therefore \text{Likelihood fn } L(\mu, \sigma^2 | \underline{x}) &= \prod_{i=1}^n f(x_i) \\ &= \prod_{i=1}^n \left(\frac{1}{\sigma\sqrt{2\pi}} \right) e^{-\frac{1}{2}\left(\frac{x_i-\mu}{\sigma}\right)^2} \\ &= \frac{1}{(\sigma\sqrt{2\pi})^n} e^{-\frac{1}{2\sigma^2} \sum (x_i-\mu)^2} \end{aligned}$$

$$\begin{aligned} \text{Log-likelihood fn: } l(\mu, \sigma^2 | \underline{x}) &= -n \ln(\sigma\sqrt{2\pi}) - \frac{1}{2\sigma^2} \sum (x_i-\mu)^2 \\ &= \left\{ -n \ln \sigma - \frac{1}{2\sigma^2} \sum (x_i-\mu)^2 + c \right\} \end{aligned}$$

For MLE:

$$\frac{\partial l}{\partial \mu} = 0 \Rightarrow 0 - \frac{1}{2\sigma^2} \cdot 2 \sum (x_i-\mu) = 0 \Rightarrow \sum (x_i-\mu) = 0 \Rightarrow \hat{\mu}_{MLE} = \bar{x}$$

$$\frac{\partial l}{\partial \sigma} = 0 \Rightarrow -\frac{n}{\sigma} - \frac{1}{\sigma^3} \sum (x_i-\mu)^2 = 0$$

$$\Rightarrow \frac{n}{\sigma} = \frac{1}{\sigma^3} \sum (x_i-\mu)^2 \Rightarrow \hat{\sigma}_{MLE}^2 = \frac{1}{n} \sum (x_i-\mu)^2$$

↳ if μ is known.

But, if μ is unknown $\hat{\sigma}^2 = \frac{1}{n-1} \sum (x_i - \hat{\mu})^2$

↳ If μ is known.

But, if μ is unknown, $\hat{\sigma}_{MLE}^2 = \frac{1}{n} \sum (x_i - \hat{\mu}_{MLE})^2$

$\Rightarrow \hat{\sigma}_{MLE}^2 = \frac{1}{n} \sum (x_i - \bar{x})^2$

To test: $H_0: \mu = \mu_0, 0 < \sigma^2 < \infty$ vs $H_1: \mu \neq \mu_0, 0 < \sigma^2 < \infty$.

$\Omega_0 = \{ \mu = \mu_0, 0 < \sigma^2 < \infty \}$

$\Omega = \{ -\infty < \mu < \infty, 0 < \sigma^2 < \infty \}$

∴ Maximized value of L under Ω

$= L(\hat{\mu}_{MLE}, \hat{\sigma}_{MLE}^2)$

Hence $L(\mu, \sigma^2) = \frac{1}{(\sigma\sqrt{2\pi})^n} e^{-\frac{1}{2\sigma^2} \sum (x_i - \mu)^2}$

$L(\hat{\mu}_{MLE}, \hat{\sigma}_{MLE}^2) = \frac{1}{(\hat{\sigma}\sqrt{2\pi})^n} e^{-\frac{1}{2\hat{\sigma}^2} \sum (x_i - \hat{\mu})^2}$

$\hat{\mu}_{MLE} = \bar{x}$

$\hat{\sigma}_{MLE}^2 = \frac{1}{n} \sum (x_i - \bar{x})^2 = s^2$

$\sum (x_i - \bar{x})^2 = ns^2$

$= \frac{1}{(\hat{\sigma}^2 2\pi)^{n/2}} e^{-\frac{1}{2\hat{\sigma}^2} \sum (x_i - \hat{\mu})^2}$

$= \frac{1}{(\hat{\sigma}^2 2\pi)^{n/2}} e^{-\frac{1}{2\hat{\sigma}^2} \sum (x_i - \bar{x})^2}$

$L^{\max}(\Omega) = \frac{1}{(s^2 2\pi)^{n/2}} e^{-\frac{n}{2}}$ → Maximized value of L under Ω .

To find the maximized value of L under $\Omega_0 = \{ \mu = \mu_0, 0 < \sigma^2 < \infty \}$

$L^{\max}(\Omega_0) = L(\mu_0, \hat{\sigma}_{MLE}^2)$ [since μ is known under Ω_0]

$= \frac{1}{\hat{\sigma}^2} e^{-\frac{1}{2\hat{\sigma}^2} \sum (x_i - \mu_0)^2}$ where, $\hat{\sigma}_{MLE}^2 = \frac{1}{n} \sum (x_i - \mu_0)^2 = s^2$

$$= \frac{1}{(\hat{\sigma}^2 2\pi)^{n/2}} e^{-\frac{1}{2\hat{\sigma}^2} \sum (x_i - \mu_0)^2} \quad \left. \begin{array}{l} \text{where,} \\ \hat{\sigma}_{MLE}^2 = \frac{1}{n} \sum (x_i - \mu_0)^2 = s_0^2 \end{array} \right\}$$

$$= \frac{1}{(s_0^2 2\pi)^{n/2}} e^{-n/2}$$

For Likelihood Ratio Test $\lambda = \frac{L^{\max}(\Omega_0)}{L^{\max}(\Omega)} = \left(\frac{s^2}{s_0^2} \right)^{n/2}$

For BCR under Likelihood Ratio Test, find $\lambda < \lambda_0 \Rightarrow$ BCR for the test.

$$P[\lambda < \lambda_0 | H_0] = \alpha.$$

Q. Let $x_1, x_2, \dots, x_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$. Find the BCR using the likelihood ratio test for testing: $H_0: \mu = \mu_0, 0 < \sigma^2 < \infty$ vs $H_1: \mu > \mu_0, 0 < \sigma^2 < \infty$ where both μ, σ^2 are unknown.

$$\Omega = \{ -\infty < \mu < \infty, 0 < \sigma^2 < \infty \}$$

$$\Omega_0 = \{ \mu = \mu_0, 0 < \sigma^2 < \infty \}$$

$$\hat{\mu}_{MLE} = \begin{cases} \bar{x}, & \mu > \mu_0 \\ \mu_0, & \mu \leq \mu_0 \end{cases}$$

$$\hat{\sigma}_{MLE}^2 = \begin{cases} s^2, & \mu > \mu_0 \\ s_0^2, & \mu \leq \mu_0 \end{cases}$$

where $s^2 = \frac{1}{n} \sum (x_i - \bar{x})^2, s_0^2 = \frac{1}{n} \sum (x_i - \mu_0)^2$

HW. Find $L^{\max}(\Omega), L^{\max}(\Omega_0), \lambda$.