

# Likelihood Ratio Test

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## 1 Likelihood Ratio Test

This test is designed for the case when we want to test Null Hypothesis  $H_0 : \mu = \mu_0$  against  $H_1 : \mu \neq \mu_0$ . As using UMP/MP level  $\alpha$  tests does not exist in this problem.

### 1.1 Algorithm

1. We want to test  $H_0 : \theta \in \Theta_0$  versus  $H_1 : \theta \in \Theta_1$ .

2. Consider:

$$\Lambda(\mathbf{x}) = \frac{\sup_{\theta \in \Theta_0} L(\theta, \mathbf{x})}{\sup_{\theta \in \Theta_0 \cup \Theta_1} L(\theta, \mathbf{x})}.$$

where  $\Lambda(\mathbf{x})$  is called **Likelihood Ratio Test Statistic**.

3. Likelihood level  $\alpha$  test is given by,

$$\varphi(\mathbf{x}) = \begin{cases} 1 & \text{if } \Lambda(\mathbf{x}) < k \\ \gamma & \text{if } \Lambda(\mathbf{x}) = k \\ 0 & \text{if } \Lambda(\mathbf{x}) > k \end{cases}$$

where  $\gamma$  and  $k$  are such that  $\mathbb{E}_\theta(\varphi(\mathbf{X})) \leq \alpha$  for all  $\theta \in \Theta_0$ .

### 1.2 Discussion

- $\sup_{\theta \in \Theta_0} L(\theta, \mathbf{x})$  can be considered as **the maximum value of the Likelihood Function over  $\Theta_0$  when  $\mathbf{X} = \mathbf{x}$  is observed**
- Similarly,  $\sup_{\theta \in \Theta_0 \cup \Theta_1} L(\theta, \mathbf{x})$  can be considered as **the maximum value of the Likelihood Function over  $\Theta_0 \cup \Theta_1$  when  $\mathbf{X} = \mathbf{x}$  is observed.**
- Clearly  $\Lambda(\mathbf{x}) \in [0, 1]$ .
- The main point of the whole algorithm/test is to *reject  $H_0$  when  $\Lambda(\mathbf{x})$  is small*. This is when likelihood under  $\Theta_0$  is lower than that of the likelihood under  $\Theta_0 \cup \Theta_1$ . This means that *observed values are more likely under  $\Theta_1$  rather than  $\Theta_0$ . Hence we reject  $H_0$ .*

## 2 Examples

**Question:** Let  $X_1, X_2, \dots, X_n \stackrel{i.i.d.}{\sim} N(\mu, \sigma^2)$ , where  $\sigma$  is known. Let  $\mu_0$  be a real number. We are interested to test  $H_0 : \mu = \mu_0$  against  $H_1 : \mu \neq \mu_0$ .

In this question,  $\Theta_0 = \{\mu_0\}$  and  $\Theta_1 = \mathbb{R}/\{\mu_0\}$ . Hence  $\Theta_0 \cup \Theta_1 = \mathbb{R}$ .

Now, step 2 of the algorithm,

$$\sup_{\mu \in \Theta_0} L(\mu) = L(\mu_0) = \left( \frac{1}{\sigma\sqrt{2\pi}} \right)^n \exp \left[ -\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu_0)^2 \right].$$

and,

$$\sup_{\mu \in \Theta_0 \cup \Theta_1} L(\mu) = L(\bar{x}) = \left( \frac{1}{\sigma\sqrt{2\pi}} \right)^n \exp \left[ -\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \bar{x})^2 \right].$$

Therefore, we can calculate  $\Lambda(\mathbf{x})$

$$\Lambda(\mathbf{x}) = \frac{\sup_{\theta \in \Theta_0} L(\theta, \mathbf{x})}{\sup_{\theta \in \Theta_0 \cup \Theta_1} L(\theta, \mathbf{x})} = \exp \left[ -\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu_0)^2 - (x_i - \bar{x})^2 \right].$$

Expanding it,

$$\begin{aligned} \Lambda(\mathbf{x}) &= \exp \left[ -\frac{1}{2\sigma^2} \sum_{i=1}^n (2x_i - \bar{x} - \mu_0)(\bar{x} - \mu_0) \right] \\ &= \exp \left[ -\frac{1}{2\sigma^2} \left( \sum_{i=1}^n 2x_i(\bar{x} - \mu_0) - \sum_{i=1}^n (\bar{x} + \mu_0)(\bar{x} - \mu_0) \right) \right] \\ &= \exp \left[ -\frac{1}{2\sigma^2} (2\bar{x}n(\bar{x} - \mu_0) - n\bar{x}^2 + n\mu_0^2) \right] \end{aligned}$$

Hence,

$$\Lambda(\mathbf{x}) = \exp \left[ -\frac{n}{2\sigma^2} (\bar{x} - \mu_0)^2 \right] < k \iff |\bar{x} - \mu_0| > k_1 \text{ for some } k_1 \in \mathbb{R}.$$

Therefore, now we can write the Likelihood Ratio Level  $\alpha$  test as:

$$\varphi(\mathbf{x}) = \begin{cases} 1 & \text{if } |\bar{x} - \mu_0| > k_1 \\ 0 & \text{otherwise} \end{cases}$$

remember that  $X_i \sim N(\mu, \sigma^2)$

where  $k_1$  is such that,

$$\begin{aligned} \mathbb{E}_{\mu_0}(\varphi(\mathbf{x})) &= P_{\mu_0} [|\bar{X} - \mu_0| > k_1] = \alpha \\ &= P_{\mu_0} \left[ \frac{\sqrt{n}}{\sigma} |\bar{X} - \mu_0| > \frac{\sqrt{n}}{\sigma} k_1 \right] = \alpha \\ &= P_{\mu_0} \left[ \frac{\sqrt{n}}{\sigma} (\bar{X} - \mu_0) > \frac{\sqrt{n}}{\sigma} k_1 \right] = \frac{\alpha}{2} \end{aligned}$$

which implies that

$$\frac{\sqrt{n}}{\sigma} k_1 = z_{\frac{\alpha}{2}}.$$

Note that we followed the same procedure as in finding UMP/MP test from previous topic.

Therefore, **the Likelihood Ratio Level  $\alpha$  test is given by:**

$$\varphi(\mathbf{x}) = \begin{cases} 1 & \text{if } \frac{\sqrt{n}}{\sigma} (\bar{X} - \mu_0) > z_{\frac{\alpha}{2}} \\ 0 & \text{otherwise} \end{cases}.$$