

# Theory of statistics 2

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# Sequential likelihood ratio test

All previous tests  $\gamma$  for the hypothesis  $H_0 : \theta = \theta_0$  vs  $H_1 : \theta = \theta_1$  use Likelihood Ratio (LR)  $\lambda$  under a fixed sample size  $n$  and rejects as  $\lambda \leq k$  where  $k$  is properly found. Whereas Sequential Likelihood Ratio Test (SLRT)  $\gamma_{SLRT}$  uses also LR  $\lambda_N$  but under a random sample size  $N$  and decides to reject or accept using two bands  $k_0 < k_1$ . Precisely,  $\gamma_{SLRT}$  computes from the one observation  $x_1$  the well-known LR:

$$\lambda_1 = \frac{\ell(x_1; \theta_0)}{\ell(x_1; \theta_1)}.$$

Then  $\gamma_{SLRT}$  rejects if  $\lambda_1 \leq k_0$  or accepts if  $\lambda_1 \geq k_1$  and stop.

If  $k_0 < \lambda_1 < k_1$ , then another observation is drawn

$$\lambda_2 = \frac{\ell(x_1, x_2; \theta_0)}{\ell(x_1, x_2; \theta_1)}.$$

is computed to **rejects** if  $\lambda_2 \leq k_0$  or **accepts** if  $\lambda_2 \geq k_1$  or to draw another more observation.

This sequential strategy makes the sample size  $N$  for which the decision of rejection or acceptance is random. This strategy may be summarize as following:

① To reject  $H_0$  at  $N$  means that:

$$\lambda_N \leq k_0; \quad k_0 < \lambda_j < k_1, j \in \{1, \dots, N-1\}.$$

② To accept  $H_0$  at  $N$  means that:

$$\lambda_N \geq k_1; \quad k_0 < \lambda_j < k_1, j \in \{1, \dots, N-1\}.$$

Where

$$\lambda_N = \frac{\ell(x_1, x_2, \dots, x_N; \theta_0)}{\ell(x_1, x_2, \dots, x_N; \theta_1)}.$$

The rejection area here is:

$$C = \bigcup_{N \geq 1} C_N,$$

where

$$C_N = \{x_1, x_2, \dots, x_n; k_0 < \lambda_i < k_1 \text{ for } i = 1, \dots, N-1; \lambda_N \leq k_0\}.$$

The acceptance area here is:

$$A = \bigcup_{N \geq 1} A_N,$$

where

$$A_N = \{x_1, x_2, \dots, x_n; k_0 < \lambda_i < k_1 \text{ for } i = 1, \dots, N-1; \lambda_N \geq k_1\}.$$

If the sizes of errors of  $\gamma_{SLRT}$  are  $\alpha^*$  and  $\beta^*$ , then

$$\alpha^* = \mathbb{P}(\text{Reject } H_0 | \theta_0) = \sum_{N \geq 1} \int_{C_N} \ell(x_1, x_2, \dots, x_N; \theta_0)$$

and

$$\beta^* = \mathbb{P}(\text{Accept } H_0 | \theta_1) = \sum_{N \geq 1} \int_{A_N} \ell(x_1, x_2, \dots, x_N; \theta_1).$$

## Theorem

- The test  $\gamma_{SLRT}$  of sizes of errors  $\alpha^*$  and  $\beta^*$  minimizes  $\mathbb{E}(N)$  among all other tests  $\gamma$  of sizes of errors  $\alpha \leq \alpha^*$  and  $\beta \leq \beta^*$ .
- The approximations

$$k_0 \approx k'_0 = \frac{\alpha^*}{1 - \beta^*} \text{ and } k_1 \approx k'_1 = \frac{1 - \alpha^*}{\beta^*}.$$

Satisfy

$$k'_0 \leq k_0 < k_1 \leq k'_1.$$



## Computation of $\mathbb{E}(N)$

$$\mathbb{E}(N|H_0) = \frac{\alpha^* \log(k'_0) + (1 - \alpha^*) \log(k'_1)}{\mathbb{E}(Z_i|H_0)}$$

$$\mathbb{E}(N|H_1) = \frac{(1 - \beta^*) \log(k'_0) + \beta^* \log(k'_1)}{\mathbb{E}(Z_i|H_1)},$$

where  $Z_i = \log \left( \frac{f(x_i; \theta_0)}{f(x_i; \theta_1)} \right)$

### Example

It was found  $\gamma_{MP}$  in testing  $H_0 : \theta = 3$  vs  $H_1 : \theta = 5$  for the normal distribution  $N(\theta, 16)$  using  $n = 25$ ,  $\alpha_{MP} = 0.05$  and  $\beta_{MP} = 0.1949$

- 1) Approximate the bands  $k_0$  and  $k_1$  of  $\gamma_{SLRT}$  with the errors of  $\gamma_{MP}$ :

$$k_0 \approx k'_0 = \frac{\alpha_{MP}}{1 - \beta_{MP}} = 0.0621 \text{ and } k_1 \approx k'_1 = \frac{1 - \alpha_{MP}}{\beta_{MP}} = 4.874.$$

- 2) Determine  $\mathbb{E}(Z_i|H_0)$  and  $\mathbb{E}(Z_i|H_1)$

$$Z_i = \log \left( \frac{f(x_i; \theta_0)}{f(x_i; \theta_1)} \right) = -\frac{1}{2 \times 16} [(x_i - 3)^2 - (x_i - 5)^2] = \frac{1}{8}(4 - x_i).$$

$$\mathbb{E}(Z_i|H_0) = \frac{1}{8}(4 - \mathbb{E}(x_i|H_0)) = \frac{1}{8}(4 - 3) = \frac{1}{8}.$$

$$\mathbb{E}(Z_i|H_1) = \frac{1}{8}(4 - \mathbb{E}(x_i|H_1)) = \frac{1}{8}(4 - 5) = -\frac{1}{8}.$$

3) Compute  $\mathbb{E}(N|H_0)$  and  $\mathbb{E}(N|H_1)$

$$\begin{aligned}\mathbb{E}(N|H_0) &= \frac{\alpha^* \log(k'_0) + (1 - \alpha^*) \log(k'_1)}{\mathbb{E}(Z_i|H_0)} \\ &= \frac{-2.779 \times 0.05 + 1.584 \times 0.95}{1/8} = 10.93\end{aligned}$$

$$\begin{aligned}\mathbb{E}(N|H_1) &= \frac{(1 - \beta^*) \log(k'_0) + \beta^* \log(k'_1)}{\mathbb{E}(Z_i|H_1)} \\ &= \frac{2.779 \times 0.8051 + 1.584 \times 0.1949}{(-1/8)} = 15.43\end{aligned}$$

## Homework

It was found  $\gamma_{MP}$  in testing  $H_0 : \theta = 2$  vs  $H_1 : \theta = 3$  for the exponential distribution  $f(x; \theta) = \theta e^{-\theta x}$ ,  $x > 0$  using  $n = 10$ ,  $\alpha_{MP} = 0.05$  and  $\beta_{MP} = 0.7$

- 1) Approximate the bands  $k_0$  and  $k_1$  of  $\gamma_{SLRT}$  with the errors of  $\gamma_{MP}$ :
- 2) Determine  $\mathbb{E}(Z_i|H_0)$  and  $\mathbb{E}(Z_i|H_1)$
- 3) Compute  $\mathbb{E}(N|H_0)$  and  $\mathbb{E}(N|H_1)$

## Homework

It was found  $\gamma_{MP}$  in testing  $H_0 : \sigma = 2$  vs  $H_1 : \sigma = 4$  for the normal distribution  $N(\theta, \sigma^2)$ , with  $\theta$  known, using  $n = 10$ ,  $\alpha_{MP} = 0.05$  and  $\beta_{MP} = 0.0824$

- 1) Approximate the bands  $k_0$  and  $k_1$  of  $\gamma_{SLRT}$  with the errors of  $\gamma_{MP}$ :
- 2) Determine  $\mathbb{E}(Z_i|H_0)$  and  $\mathbb{E}(Z_i|H_1)$
- 3) Compute  $\mathbb{E}(N|H_0)$  and  $\mathbb{E}(N|H_1)$

## Homework

It was found  $\gamma_{MP}$  in testing  $H_0 : \theta = 2$  vs  $H_1 : \theta = 3$  for the distribution  $f(x; \theta) = \theta x^{\theta-1}$ ,  $0 < x < 1$  using  $n = 10$ ,  $\alpha_{MP} = 0.05$  and  $\beta_{MP} = 0.23$

- 1) Approximate the bands  $k_0$  and  $k_1$  of  $\gamma_{SLRT}$  with the errors of  $\gamma_{MP}$ :
- 2) Determine  $\mathbb{E}(Z_i|H_0)$  and  $\mathbb{E}(Z_i|H_1)$
- 3) Compute  $\mathbb{E}(N|H_0)$  and  $\mathbb{E}(N|H_1)$

**Thank you**