

# hw8-report

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Let  $Y_1, Y_2, \dots, Y_n \sim_{iid} N(\theta, \sigma^2)$  where  $\sigma^2$  is known. Consider  $H_0 : \theta \in \Omega_0$  vs  $H_1 : \theta \in \Omega_1$  where

$$\Omega_0 = (-\infty, \theta_0], \quad \Omega_1 = (\theta_0, \infty), \quad \Omega = (-\infty, \infty) = \Omega_0 \cup \Omega_1$$

1. Write down the joint distribution  $f(y_1, \dots, y_n)$  and likelihood function  $L(\theta)$ .

$$= \left( \frac{1}{\sqrt{2\pi}\sigma} \right)^n e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \theta)^2}$$

The two equations have the same form, but the unknowns are not the same. For the joint distribution, the  $y$  values have not yet been observed and are unknown. On the other hand, the joint distribution *becomes* the likelihood function once these  $y$  values have been observed and the value of  $\theta$  is the primary unknown of the equation.

2. Given that the MLE is

$$\hat{\theta} = \bar{y} = \frac{1}{n} \sum_{i=1}^n y_i,$$

then what is the  $\max_{\theta \in \Omega} L(\theta)$ ?

$$\max L(\theta) = L(\bar{y}) = \left( \frac{1}{\sqrt{2\pi}\sigma} \right)^n e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \bar{y})^2}$$

3. Under  $H_0 : \theta \leq \theta_0$ , the restricted MLE is

$$\hat{\theta}_R = \begin{cases} \bar{y}, & \bar{y} \leq \theta_0 \\ \theta_0, & \bar{y} > \theta_0 \end{cases}$$

then what is  $\max_{\theta \in \Omega_0} L(\theta)$ ?

$$\max_{H_0} L(\theta) = L(\theta_0) = \begin{cases} \left( \frac{1}{\sqrt{2\pi}\sigma} \right)^n e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \bar{y})^2}, & \bar{y} \leq \theta_0 \\ \left( \frac{1}{\sqrt{2\pi}\sigma} \right)^n e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \theta_0)^2}, & \bar{y} > \theta_0 \end{cases}$$

4. The likelihood ratio test statistic is

$$\lambda(y_1, \dots, y_n) = \frac{\max_{\theta \in \Omega_0} L(\theta)}{\max_{\theta \in \Omega} L(\theta)}$$

Obtain this likelihood ratio test statistic.

5. The likelihood ratio test procedure is that we will reject  $H_0$  if  $\lambda(y_1, \dots, y_n) \leq k$ , where  $k < 1$ . Explain that this is equivalent to  $z \geq \sqrt{-2 \log k}$ , where

$$z = \frac{\bar{y} - \theta_0}{\sigma/\sqrt{n}}$$

6. Why must  $k < 1$ ?

One can tell that the value for comparison,  $k$ , must be less than 1 simply by understanding the terms in the ratio it is compared with. The denominator of  $\lambda(y_1, \dots, y_n)$  is  $\max_{\theta \in \Omega} L(\theta)$ . This term represents the global maximum of the likelihood with no restriction on the values that  $\theta$  can take. The numerator of  $\lambda(y_1, \dots, y_n)$  is the maximum of the same likelihood function, but with a restriction on the values that  $\theta$  can take. It therefore follows that the value numerator will be less than that of the denominator, leading to a value wherein  $k < 1$ .

7. (Optional) Derive both MLE  $\hat{\theta}$  and restricted MLE  $\hat{\theta}_R$ .