

Question 3

For some reason a lot of people like to assume measurement error is Normally distributed with a mean value μ and a standard deviation σ . Consider testing $H_0 : \mu = 0$ versus $H_a : \mu \neq 0$ based on $n = 16$ measurements.

Find the likelihood ratio test, λ , and its Rejection Region (RR) when $\alpha = 0.05$.

$$L(y_1, \dots, y_{16} | \mu) = \left(\frac{1}{\sqrt{32\pi}} \right)^{16} e^{-\frac{1}{32} \sum_{i=1}^{16} (y_i - \mu)^2}$$

find the MLE: $l(\mu) = -\frac{1}{2} \ln(32\pi) - \frac{1}{32} \sum_{i=1}^{16} (y_i - \mu)^2$

$$l'(\mu) = -\frac{1}{32} (2) \sum_{i=1}^{16} (y_i - \mu) (-1) \stackrel{\text{set}}{=} 0$$

$$\Rightarrow \frac{1}{16} \sum_{i=1}^{16} y_i - \frac{1}{16} \sum_{i=1}^{16} \mu = 0$$

$$\Rightarrow \frac{1}{16} \sum_{i=1}^{16} y_i = \mu \Rightarrow \hat{\mu} = \bar{y}$$

verify it's a maximum:

$$l''(\mu) = 0 - \frac{1}{16} \sum_{i=1}^{16} (1) = -1 < 0$$

thus $\hat{\mu} = \bar{y}$ is the MLE.

$$\begin{aligned} \text{Now, } \lambda &= \frac{L(\hat{\Omega}_0)}{L(\hat{\Omega})} = \frac{\left(\frac{1}{\sqrt{32\pi}} \right)^{16} e^{-\frac{1}{32} \sum_{i=1}^{16} y_i^2}}{\left(\frac{1}{\sqrt{32\pi}} \right)^{16} e^{-\frac{1}{32} \sum_{i=1}^{16} (y_i - \bar{y})^2}} \\ &= \frac{e^{-\frac{1}{32} \sum_{i=1}^{16} y_i^2}}{e^{-\frac{1}{32} \sum_{i=1}^{16} (y_i - \bar{y})^2}} \end{aligned}$$

$$\text{RMK: } y_i^2 = ((y_i - \bar{y}) + (\bar{y} - 0))^2 = (y_i - \bar{y})^2 + 2(y_i - \bar{y})(\bar{y}) + \bar{y}^2$$

$$\begin{aligned} \sum_{i=1}^{16} y_i^2 &= \sum_{i=1}^{16} (y_i - \bar{y})^2 + 2\bar{y} \sum_{i=1}^{16} (y_i - \bar{y}) + \sum_{i=1}^{16} \bar{y}^2 \\ &= \sum_{i=1}^{16} (y_i - \bar{y})^2 + 2\bar{y} \left(\sum_{i=1}^{16} y_i - 16\bar{y} \right) + 16\bar{y}^2 \\ &= \sum_{i=1}^{16} (y_i - \bar{y})^2 + 16\bar{y}^2 \end{aligned}$$

$$= \frac{e^{-\frac{1}{32} \left[\sum_{i=1}^{16} (y_i - \bar{y})^2 + 16\bar{y}^2 \right]}}{e^{-\frac{1}{32} \sum_{i=1}^{16} (y_i - \bar{y})^2}} = e^{-\frac{1}{2} \bar{y}^2}$$

$$\text{Hence, } \lambda = \ln(e^{-\frac{1}{2} \bar{y}^2}) \leq \ln(k) = k,$$

$$-\frac{1}{2} \bar{y}^2 \leq k \Rightarrow \bar{y}^2 \geq -2k = k^*$$

$$\text{Recall: } \left(\frac{\bar{y} - \mu}{\sqrt{\sigma^2/n}} \right)^2 \sim \chi^2_{(1)} \Rightarrow \frac{n\bar{y}^2}{\sigma^2} \sim \chi^2_{(1)}$$

Since $n=16$ and $\sigma^2=16$... we're good.

$$\alpha = P(\bar{y}^2 \geq k^* | \mu = 0)$$

$$\chi^2_{df=16, \alpha=0.05} = 3.84146$$

$$\text{Thus } RR = \{ \bar{y}^2 \geq 3.84146 \}$$