

# Math 741 Assignment 11 (Quiz)

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6.5.1. solution:

$$L(p) = \prod_{i=1}^n (1-p)^{k_i-1} p = (1-p)^{\sum_{i=1}^n k_i - n} p^n$$

$$\begin{aligned} L'(p) &= \left( \sum_{i=1}^n k_i - n \right) \cdot (1-p)^{\sum_{i=1}^n k_i - n - 1} \cdot (-1) \cdot p^n + (1-p)^{\sum_{i=1}^n k_i - n} (np^{n-1}) \\ &= \left[ (1-p)^{\sum_{i=1}^n k_i - n - 1} p^{n-1} \right] \cdot \left[ p(n - \sum_{i=1}^n k_i) + n(1-p) \right] \end{aligned}$$

Solve for  $p$  for  $L'(p) = 0$ , then

$$p(n - \sum_{i=1}^n k_i) + n(1-p) = 0 \Rightarrow p = \frac{n}{\sum_{i=1}^n k_i}$$

Therefore,

$$\begin{aligned} \max_{\Omega} [L(p)] &= \left( 1 - \frac{n}{\sum_{i=1}^n k_i} \right)^{\sum_{i=1}^n k_i - n} \left( \frac{n}{\sum_{i=1}^n k_i} \right)^n \\ \max_{\Omega_0} [L(p)] &= (1-p_0)^{\sum_{i=1}^n k_i - n} p_0^n \\ \lambda &= \frac{\max_{\Omega_0} [L(p)]}{\max_{\Omega} [L(p)]} = \frac{(1-p_0)^{\sum_{i=1}^n k_i - n} p_0^n}{\left( 1 - \frac{n}{\sum_{i=1}^n k_i} \right)^{\sum_{i=1}^n k_i - n} \left( \frac{n}{\sum_{i=1}^n k_i} \right)^n} \end{aligned}$$

6.5.2.(H) solution: Given  $y_1, \dots, y_{10} \sim EXP(\lambda)$ , iid. Then

$$f_Y(y; \lambda) = \begin{cases} \lambda e^{-\lambda y} & x > 0, \lambda > 0 \\ 0 & o.w. \end{cases}$$

Therefore,

$$L(\lambda) = \prod_{i=1}^{10} \lambda e^{-\lambda y_i} = \lambda^{10} e^{-\lambda \sum_{i=1}^{10} y_i}$$

$$\ln[L(\lambda)] = 10 \ln \lambda - \lambda \sum_{i=1}^{10} y_i$$

$$\frac{\partial}{\partial \lambda} [\ln[L(\lambda)]] = \frac{10}{\lambda} - \sum_{i=1}^{10} y_i$$

Let  $\frac{\partial}{\partial \lambda} [\ln[L(\lambda)]] = 0$ , then

$$\frac{10}{\hat{\lambda}} - \sum_{i=1}^{10} y_i = 0 \implies \hat{\lambda} = \frac{10}{\sum_{i=1}^{10} y_i}$$

Therefore,

$$\max_{\Omega} [L(\lambda)] = \left( \frac{10}{\sum_{i=1}^{10} y_i} \right)^{10} e^{-10}$$

$$\max_{\Omega_0} [L(\lambda)] = \lambda_0^{10} e^{-\lambda_0 \sum_{i=1}^{10} y_i}$$

Let

$$\Lambda = \frac{\lambda_0^{10} e^{-\lambda_0 \sum_{i=1}^{10} y_i}}{\left( \frac{10}{\sum_{i=1}^{10} y_i} \right)^{10} e^{-10}}$$

Moreover, it is a two-sided test. The integral should be

$$\begin{aligned} P(0 \leq \Lambda \leq \lambda^* | H_0 \text{ is true}) &= 0.05 \\ &= \int_0^{\lambda^*} f_{\Lambda}(t | H_0 \text{ is true}) dt \end{aligned}$$

6.5.3. solution: Given  $y_1, \dots, y_n \sim N(\mu, 1)$  and

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu \neq \mu_0$$

We know that MLE is  $\mu = \bar{y}$ , so

$$L(\lambda) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi}} e^{-\frac{(y_i - \mu)^2}{2}} = \left( \frac{1}{\sqrt{2\pi}} \right)^n e^{-\frac{\sum_{i=1}^n (y_i - \mu)^2}{2}}$$

$$\begin{aligned}
\max_{\Omega} [L(\lambda)] &= \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{2}} \\
\max_{\Omega_0} [L(\lambda)] &= \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \mu_0)^2}{2}} \\
\Lambda &= \frac{\left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \mu_0)^2}{2}}}{\left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{2}}} = e^{-\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{2} + \frac{\sum_{i=1}^n (y_i - \mu_0)^2}{2}} \\
&= e^{-\frac{1}{2} \left( \frac{\sum_{i=1}^n (\bar{y} - \mu_0)}{1/\sqrt{n}} \right)^2}
\end{aligned}$$

The generalized likelihood ratio test is one that rejects the null hypothesis when ever  $0 < \lambda \leq \lambda^*$  so that

$$P(0 \leq \Lambda \leq \lambda^* | H_0 \text{ is true}) = \alpha$$

6.5.4.(H) solution: Given  $y_1, \dots, y_n \sim N(\mu, 1)$  and

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu = \mu_1$$

We know that MLE is  $\mu = \bar{y}$ , so

$$\begin{aligned}
L(\lambda) &= \prod_{i=1}^n \frac{1}{\sqrt{2\pi}} e^{-\frac{(y_i - \mu)^2}{2}} = \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \mu)^2}{2}} \\
\max_{\Omega} [L(\lambda)] &= \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{2}} \\
\max_{\Omega_0} [L(\lambda)] &= \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \mu_0)^2}{2}} \\
\Lambda &= \frac{\left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \mu_0)^2}{2}}}{\left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{2}}} = e^{-\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{2} + \frac{\sum_{i=1}^n (y_i - \mu_0)^2}{2}} \\
&= e^{-\frac{1}{2} n (\bar{y} - \mu_0)^2} = e^{-\frac{1}{2} \left( \frac{(\bar{y} - \mu_0)}{1/\sqrt{n}} \right)^2}
\end{aligned}$$

where  $z = \frac{(\bar{y} - \mu_0)}{1/\sqrt{n}}$ . The generalized likelihood ratio test is one that rejects the null hypothesis whenever  $0 < \lambda \leq \lambda^*$  so that

$$P(0 \leq \Lambda \leq \lambda^* | H_0 \text{ is true}) = \alpha$$

The likelihood ratio test didn't change in this case.

The critical region depends on the particular value of  $\mu_1$  since  $\alpha = P(\mu = \mu_1 | \mu = \mu_0)$

6.5.5. solution: