

Homework 11

Xupeng Chen

2017/12/19

1.(i)

We can assume that $X_i \sim B(p)$, p is unknown.

So the number of young adults listening to that program $\sum_{i=1}^n X_i \sim B(n, p)$.

ii.

binomial test: $H_0 : p \leq p_0, H_1 : p > p_0$.

Use the program we can get $p\text{-value} = 0.0001$ in this test, supports H_1 .

So the claim made by the manager is supported.

2

i.

X are counts of the head and $X \sim B(n, \theta)$.

$$L(\theta_0; x) = \binom{n}{x} \theta_0^x (1 - \theta_0)^{n-x}, \because \theta_{MLE} = \frac{x}{n}, \therefore \sup_{\theta \in \Theta} L(\theta; x) = \binom{n}{x} \left(\frac{x}{n}\right)^x \left(\frac{n-x}{n}\right)^{n-x},$$

$$\lambda = \frac{n^n \theta_0^x (1 - \theta_0)^{n-x}}{x^x (n-x)^{n-x}},$$

$$\frac{\partial(-\log \lambda)}{\partial x} = \log x + 1 - \log(n-x) - 1 - \log \theta_0 + \log(1 - \theta_0) = \log \left[\frac{x(1 - \theta_0)}{\theta_0(n-x)} \right],$$

$$\text{so we have: } \begin{cases} \frac{\partial(-\log \lambda)}{\partial x} < 0, & x < n\theta_0 \\ \frac{\partial(-\log \lambda)}{\partial x} = 0, & x = n\theta_0 \\ \frac{\partial(-\log \lambda)}{\partial x} > 0, & x > n\theta_0 \end{cases}$$

$$\therefore \lambda < \lambda_0 \Leftrightarrow -\log \lambda > C_1 \Leftrightarrow |x - n\theta_0| > C_2.$$

$$\text{Let } P_{\theta_0}(|X - n\theta_0| > C_2) = \alpha, P_{\theta_0}(X > n\theta_0 + C_2) = \frac{\alpha}{2}.$$

$$\text{then find the smallest } C_2 \text{ s.t. } P_{\theta_0}(X > n\theta_0 + C_2) = \sum_{X=n\theta_0+C_2}^n \binom{n}{X} \theta_0^X (1 - \theta_0)^{n-X} \leq \frac{\alpha}{2}.$$

$$P_{\theta}(X > 57) = 0.067, \text{ and } P_{\theta}(X > 58) = 0.044, \therefore C_2 = 8.$$

$$\text{The likelihood ratio test is } \begin{cases} 1, & X > 58 \text{ or } X < 42 \\ 0, & \text{otherwise} \end{cases}$$

$$\because x = 60 > 58, \therefore H_0 \text{ is rejected.}$$

(ii)

$$\text{According to CLT, } \frac{X - n\theta}{\sqrt{n\theta(1 - \theta)}} \xrightarrow[n \rightarrow \infty]{d} N(0, 1).$$

$$\text{Let } P_{\theta_0}(|X - n\theta_0| > C_2) = \alpha, \text{ then } C_2 = z_{\alpha/2} \sqrt{n\theta_0(1 - \theta_0)} = 8.224268.$$

$$\because x = 60 > 8.224268, \therefore H_0 \text{ is rejected.}$$

3.

Let X_i be the i -th women's blood pressure difference after and before the usage of the pill.

According to CLT, $\frac{\sqrt{n}(\bar{X} - \mu)}{S} \xrightarrow[n \rightarrow \infty]{d} N(0, 1), S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$.

Assume X_i approxiamtely $i. i. d. \sim N(\mu, \sigma^2)$.

$$L(\mu, \sigma^2; \mathbf{x}) = (2\pi\sigma^2)^{-n/2} \exp\left[\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2\right],$$

$$\begin{cases} \frac{\partial \log L(\mu, \sigma^2; \mathbf{x})}{\partial \mu} = \frac{n}{\sigma^2} (\bar{x} - \mu) \\ \frac{\partial \log L(\mu, \sigma^2; \mathbf{x})}{\partial (\sigma^2)} = -\frac{n}{2\sigma^2} + \frac{1}{2\sigma^4} \sum_{i=1}^n (x_i - \mu)^2. \end{cases}$$

It is easy to get that

$$\sup_{\theta \in \Theta} L(\theta; \mathbf{x}) = \left[\frac{2\pi}{n} \sum_{i=1}^n (x_i - \bar{x})^2 \right]^{-n/2} \exp\left(-\frac{n}{2}\right)$$

$$\sup_{\theta \in \Theta_0} L(\theta; \mathbf{x}) = \begin{cases} \left[\frac{2\pi}{n} \sum_{i=1}^n (x_i - \bar{x})^2 \right]^{-n/2} \exp\left(-\frac{n}{2}\right), & \bar{x} \geq \mu_0 \\ \left[\frac{2\pi}{n} \sum_{i=1}^n (x_i - \mu_0)^2 \right]^{-n/2} \exp\left(-\frac{n}{2}\right), & \bar{x} < \mu_0 \end{cases}$$

$$\lambda = \frac{\sup_{\theta \in \Theta_0} L(\theta; \mathbf{x})}{\sup_{\theta \in \Theta} L(\theta; \mathbf{x})} = \begin{cases} 1, & \bar{x} \geq \mu_0 \\ \left[\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{\sum_{i=1}^n (x_i - \mu_0)^2} \right]^{-n/2}, & \bar{x} < \mu_0 \end{cases}.$$

$$\lambda < \lambda_0 \Leftrightarrow \frac{\sqrt{n}(\bar{x} - \mu_0)}{\sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}} < C.$$

$$\text{Let } P_{\mu_0} \left(\frac{\sqrt{n}(\bar{x} - \mu_0)}{\sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}} < C \right) = \alpha, \text{ then } C = -t_{n-1, \alpha}.$$

$$\text{The likelihood ratio test is } \varphi(\mathbf{x}) = \begin{cases} 1, & \frac{\sqrt{n}(\bar{x} - \mu_0)}{\sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}} < -t_{n-1, \alpha/2} \\ 0, & \text{otherwise} \end{cases}.$$

(ii)

$$\frac{\sqrt{n}(\bar{x} - \mu_0)}{\sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}} = \frac{\sqrt{90}\bar{x}}{s}, t_{n-1, \alpha/2} = t_{89, 0.05}.$$

If $\frac{\sqrt{90}\bar{x}}{s} < -t_{89, 0.05}$, then H_0 is rejected. Else, H_0 is accepted.

4.

i.

$$H_0 : \mu = 2.5, H_1 : \mu \neq 2.5.$$

$$L(\mu_0; \mathbf{x}) = (2\pi\sigma^2)^{-n/2} \exp\left[-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu_0)^2\right],$$

$$\sup_{\theta \in \Theta} L(\theta; \mathbf{x}) = (2\pi\sigma^2)^{-n/2} \exp\left[-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \bar{x})^2\right],$$

$$\lambda = \frac{L(\theta_0; \mathbf{x})}{\sup_{\theta \in \Theta} L(\theta; \mathbf{x})} = \exp\left[-\frac{n}{2\sigma^2} (\bar{x} - \mu_0)^2\right],$$

$$\lambda < \lambda_0 \Leftrightarrow \frac{n}{2\sigma^2} (\bar{x} - \mu_0)^2 > C_1 \Leftrightarrow \frac{\sqrt{n}|\bar{x} - \mu_0|}{\sigma} > C_2$$

$$\text{Let } P_{\mu_0}\left(\frac{\sqrt{n}|\bar{x} - \mu_0|}{\sigma} > C_2\right) = \alpha, \text{ then } C_2 = z_{\alpha/2}.$$

$$\text{The likelihood ratio test is } \varphi(\mathbf{x}) = \begin{cases} 1, & \frac{\sqrt{n}|\bar{x} - \mu_0|}{\sigma} > z_{\alpha/2} \\ 0, & \text{otherwise} \end{cases}$$

$$\frac{\sqrt{n}|\bar{x} - \mu_0|}{\sigma} = 0.8 < 1.959964 = z_{\alpha/2}, \therefore H_0 \text{ is accepted.}$$

(ii)

The power function is

$$\begin{aligned} \beta(\mu) &= P_{\mu}\left(\frac{\sqrt{n}|\bar{x} - \mu_0|}{\sigma} > z_{\alpha/2}\right) \\ &= 1 + \Phi\left(-z_{\alpha/2} + \frac{\sqrt{n}(\mu_0 - \mu)}{\sigma}\right) - \Phi\left(z_{\alpha/2} + \frac{\sqrt{n}(\mu_0 - \mu)}{\sigma}\right) \\ &= 1 + \Phi(98.04 - 40\mu) - \Phi(101.96 - 40\mu) \end{aligned}$$

5.

i.

The joint likelihood function is

$$L(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2; \mathbf{x}, \mathbf{y}) = (2\pi)^{-\frac{m+n}{2}} (\sigma_1^2)^{-\frac{m}{2}} (\sigma_2^2)^{-\frac{n}{2}} \exp\left[-\frac{1}{2\sigma_1^2} \sum_{i=1}^m (x_i - \mu_1)^2 - \frac{1}{2\sigma_2^2} \sum_{i=1}^n (y_i - \mu_2)^2\right],$$

$$\log L(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2; \mathbf{x}, \mathbf{y}) = -\frac{m+n}{2} \log 2\pi - \frac{m}{2} \log \sigma_1^2 - \frac{n}{2} \log \sigma_2^2 - \frac{1}{2\sigma_1^2} \sum_{i=1}^m (x_i - \mu_1)^2 - \frac{1}{2\sigma_2^2} \sum_{i=1}^n (y_i - \mu_2)^2$$

$$\text{Let } \begin{cases} \frac{\partial \log L(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2; \mathbf{x}, \mathbf{y})}{\partial \mu_1} = 0 \\ \frac{\partial \log L(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2; \mathbf{x}, \mathbf{y})}{\partial \mu_2} = 0 \\ \frac{\partial \log L(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2; \mathbf{x}, \mathbf{y})}{\partial (\sigma_1^2)} = 0 \\ \frac{\partial \log L(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2; \mathbf{x}, \mathbf{y})}{\partial (\sigma_2^2)} = 0 \end{cases},$$

we can get the only maximum point $(\bar{x}, \bar{y}, \frac{1}{m} \sum_{i=1}^m (x_i - \bar{x})^2, \frac{1}{n} \sum_{i=1}^n (y_i - \bar{y})^2)$,

$$\therefore \sup_{\sigma_1 \neq \sigma_2} L(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2; \mathbf{x}, \mathbf{y}) = (2\pi)^{-\frac{m+n}{2}} \left[\frac{1}{m} \sum_{i=1}^m (x_i - \bar{x})^2 \right]^{-\frac{m}{2}} \left[\frac{1}{n} \sum_{i=1}^n (y_i - \bar{y})^2 \right]^{-\frac{n}{2}} \exp\left(-\frac{m+n}{2}\right).$$

$$\text{Similarly, } \sup_{\sigma_1 = \sigma_2 = \sigma} L(\mu_1, \mu_2, \sigma^2; \mathbf{x}, \mathbf{y}) = (2\pi)^{-\frac{m+n}{2}} \left\{ \frac{1}{m+n} \left[\sum_{i=1}^m (x_i - \bar{x})^2 + \sum_{i=1}^n (y_i - \bar{y})^2 \right] \right\}^{-\frac{m+n}{2}} \exp\left(-\frac{m+n}{2}\right),$$

$$\lambda = \frac{\left\{ \frac{1}{m+n} \left[\sum_{i=1}^m (x_i - \bar{x})^2 + \sum_{i=1}^n (y_i - \bar{y})^2 \right] \right\}^{-\frac{m+n}{2}}}{\left[\frac{1}{m} \sum_{i=1}^m (x_i - \bar{x})^2 \right]^{-\frac{m}{2}} \left[\frac{1}{n} \sum_{i=1}^n (y_i - \bar{y})^2 \right]^{-\frac{n}{2}}}.$$

$$\lambda < \lambda_0 \Leftrightarrow \frac{(n-1) \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sum_{i=1}^n (y_i - \bar{y})^2} < C_1 \text{ or } \frac{(n-1) \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sum_{i=1}^n (y_i - \bar{y})^2} > C_2.$$

$$\text{Under assumption of } H_0, \frac{(n-1) \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sum_{i=1}^n (y_i - \bar{y})^2} \sim F_{m-1, n-1},$$

$$\therefore C_1 = F_{m-1, n-1, 1-\alpha/2}, C_2 = F_{m-1, n-1, \alpha/2} \text{ make } P(\lambda < \lambda_0) = \alpha.$$

The likelihood ratio test is

$$\varphi(\mathbf{x}) = \begin{cases} 1, & \frac{(n-1) \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sum_{i=1}^n (y_i - \bar{y})^2} < F_{m-1, n-1, 1-\alpha/2} \text{ or } \frac{(n-1) \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sum_{i=1}^n (y_i - \bar{y})^2} > F_{m-1, n-1, \alpha/2} \\ 0, & \text{otherwise} \end{cases}$$

$$\text{In this question, } \frac{s_x^2}{s_y^2} = \frac{4}{9} > 0.2295034 = F_{8,9,0.025}, \therefore H_0 \text{ is accepted.}$$

(ii)

$F_{m-1, n-1}(x)$ is the c.d.f. of F-distribution $F_{m-1, n-1}$.

$$\therefore \frac{(n-1) \sigma_2^2 \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sigma_1^2 \sum_{i=1}^n (y_i - \bar{y})^2} \sim F_{m-1, n-1},$$

\therefore the power function is

$$\beta(\sigma_1, \sigma_2) = P_{\sigma_1, \sigma_2} \left(\frac{(n-1) \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sum_{i=1}^n (y_i - \bar{y})^2} < F_{m-1, n-1, 1-\alpha/2} \right) + P_{\sigma_1, \sigma_2} \left(\frac{(n-1) \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sum_{i=1}^n (y_i - \bar{y})^2} > F_{m-1, n-1, \alpha/2} \right)$$

$$= P_{\sigma_1, \sigma_2} \left(\frac{(n-1) \sigma_2^2 \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sigma_1^2 \sum_{i=1}^n (y_i - \bar{y})^2} < \frac{\sigma_2^2}{\sigma_1^2} F_{m-1, n-1, 1-\alpha/2} \right) + P_{\sigma_1, \sigma_2} \left(\frac{(n-1) \sigma_2^2 \sum_{i=1}^m (x_i - \bar{x})^2}{(m-1) \sigma_1^2 \sum_{i=1}^n (y_i - \bar{y})^2} > \frac{\sigma_2^2}{\sigma_1^2} F_{m-1, n-1, \alpha/2} \right)$$

$$= 1 + F_{m-1, n-1} \left(\frac{\sigma_2^2}{\sigma_1^2} F_{m-1, n-1, 1-\alpha/2} \right) - F_{m-1, n-1} \left(\frac{\sigma_2^2}{\sigma_1^2} F_{m-1, n-1, \alpha/2} \right).$$

$$\text{so } \beta(2, 3) = 1 + F_{8,9} \left(\frac{9}{4} F_{8,9,0.025} \right) - F_{8,9} \left(\frac{9}{4} F_{8,9,0.975} \right) = 0.1838922.$$

6.

i.

The likelihood function is $L(\theta; \mathbf{x}) = \prod_{i=1}^n \frac{1}{1-\theta} I_{\theta \leq x_i \leq 1} = \frac{1}{(1-\theta)^n} I_{\theta \leq x_{(1)}}$

$$\sup_{\theta \in \Theta} L(\theta; \mathbf{x}) = \frac{1}{(1-x_{(1)})^n},$$

$$\sup_{\theta \in \Theta_0} L(\theta; \mathbf{x}) = \begin{cases} \frac{1}{(1-x_{(1)})^n}, & \theta_0 \leq x_{(1)} \\ 0, & \theta_0 > x_{(1)} \end{cases},$$

$$\lambda = \frac{\sup_{\theta \in \Theta_0} L(\theta; x)}{\sup_{\theta \in \Theta} L(\theta; x)} = \begin{cases} 1, & \theta_0 \leq x_{(1)} \\ 0, & \theta_0 > x_{(1)} \end{cases}.$$

$$\text{The likelihood ratio test is } \varphi(\mathbf{x}) = \begin{cases} 1, & \theta_0 > x_{(1)} \\ 0, & \theta_0 \leq x_{(1)} \end{cases}$$

ii.

The p.d.f. of $X_{(1)}$ is $f(x) = n(1 - \frac{x-\theta}{1-\theta})^{n-1} \cdot \frac{1}{1-\theta} = \frac{n}{(1-\theta)^n} (1-x)^{n-1}, \theta < x < 1$.

$$\text{If } \theta_0 > \theta, \text{ then } P_{\theta}(\theta_0 > X_{(1)}) = \int_{\theta}^{\theta_0} f(x) dx = 1 - \left(\frac{1-\theta_0}{1-\theta}\right)^n.$$

$$\text{If } \theta_0 \leq \theta, \text{ then } P_{\theta}(\theta_0 > X_{(1)}) = 0.$$

$$\text{The power function is } \beta(\theta) = P_{\theta}(\theta_0 > X_{(1)}) = \begin{cases} 1 - \left(\frac{1-\theta_0}{1-\theta}\right)^n, & \theta < \theta_0 \\ 0, & \theta \geq \theta_0 \end{cases}.$$