补充题

Assignment: X_1, \ldots, X_n i.i.d. $\sim Exp(\lambda)$. Solve the level $\alpha = 0.1$ UMPT for $H_0: \lambda \leq 1$ or $\lambda \geq 2, H_1: 1 < \lambda < 2$.

Solve: Notice that the joint p.d.f. is

$$f(\mathbf{x}|\lambda) = \lambda^n e^{-\lambda \sum_i x_i} I_{(0,\infty)}(x_{(1)}),$$

where λ is strictly increasing with respect to λ and $T(\mathbf{X}) := -\sum_i X_i$. Then the reject region for a level α UMPT should be $\{t_1 < T < t_2\}$. Similar to Ex. 5.48, we know that $-2\lambda T \sim \chi_{2n}^2$, thus t_1, t_2 can be solved by

$$\begin{cases} \alpha = P(t_1 < T < t_2 | \lambda = 1) = P(-2t_1 > \chi_{2n}^2 > -2t_2) = \chi_{2n}^2(-2t_1) - \chi_{2n}^2(-2t_2) \\ \alpha = P(t_1 < T < t_2 | \lambda = 2) = P(-4t_1 > \chi_{2n}^2 > -4t_2) = \chi_{2n}^2(-4t_1) - \chi_{2n}^2(-4t_2) \end{cases}$$

where $P(\chi_{2n}^2 > \chi_{2n}^2(\alpha)) = \alpha$.

Thus, the UMPT rejects H_0 if $x_1 < \sum_i X_i < x_2$, where x_1, x_2 satisfy

$$\begin{cases} 0.1 = \chi_{2n}^2(2x_2) - \chi_{2n}^2(2x_1) \\ 0.1 = \chi_{2n}^2(4x_2) - \chi_{2n}^2(4x_1) \end{cases}$$

4.1

[Wei] 4.1. Notice that $Q(\mathbf{X}, \mu) = \frac{\sqrt{n}(\bar{X}-\mu)}{\sigma} \sim N(0,1)$ is a pivot. Then a 0.95 confidence interval is $\{-z_{0.05/2} \leq Q \leq z_{0.05/2}\}$. With $\bar{X} = 4.7832$, $\sigma = 0.01$, n = 5, the confidence interval is

$$[\bar{X} - \frac{\sigma}{\sqrt{n}}z_{0.025}, \bar{X} + \frac{\sigma}{\sqrt{n}}z_{0.025}] = [4.774, 4.792].$$

4.3

[Wei] 4.3. Similar to Exercise 4.2, a 0.05 confidence interval is $[\bar{X} - \frac{S}{\sqrt{n}}t_{n-1}(\alpha/2), \bar{X} + \frac{S}{\sqrt{n}}t_{n-1}(\alpha/2)] = [1784, 2116].$

4.4

[Wei] 4.4. Here we have

$$\frac{\sqrt{n}(\bar{X}-\mu)}{4} \sim N(0,1).$$

and we may transform the probability as

$$P(\bar{X}-1 < \mu < \bar{X}+1) = P(-\frac{\sqrt{n}}{4} < Z < \frac{\sqrt{n}}{4}) \ge 0.9.$$

where $Z \sim N(0,1)$. We also note that $z_{\alpha/2} = 1.65$, which implies that $\frac{\sqrt{n}}{4} \geq 1.65$, thus we have $n \geq 44$.

4.10

[Wei] 4.10. The problem should be revised as "how many should n be at most?" Notice that $\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$, and from the definition of lower confidence limit,

$$0.95 \le P\left(\sigma \ge \frac{\sqrt{(n-1)S^2}}{4}\right) = P(\chi_{n-1}^2 \le 16).$$

Thus we have $n \leq 9$.

$$\begin{split} m &= 5, n = 7, \overline{X} = 10.06, \overline{Y} = 12.1857, S_1^2 = 0.093, S_2^2 = 0.1248, \overline{X} - \overline{Y} = -2.1257, \\ S_\omega^2 &= \frac{(m-1)S_1^2 + (n-1)S_2^2}{n+m-2} = 0.1121, S_\omega = 0.3348,$$
 查表得 $t_{m+n-2}(\alpha/2) = 2.2281$,

算得
$$S_{\omega}t_{m+n-2}(lpha/2)\sqrt{rac{1}{n}+rac{1}{m}}=0.4368$$
,则置信区间为

$$\left[\overline{X}-\overline{Y}-S_{\omega}t_{m+n-2}(\alpha/2)\sqrt{\frac{1}{n}+\frac{1}{m}}, \overline{X}-\overline{Y}+S_{\omega}t_{m+n-2}(\alpha/2)\sqrt{\frac{1}{n}+\frac{1}{m}}\right]$$

$$= [-2.5625, -1.6889]$$

4.14

[Wei] 4.14. Facts:

(1)
$$\frac{\bar{Y}-\bar{X}-(b-a)}{\sqrt{\sigma_1^2/m+\sigma_2^2/n}} \sim N(0,1).$$

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$$\frac{\bar{Y} - \bar{X} - (b - a)}{\sqrt{\sigma_1^2 / m + \sigma_2^2 / n}} \sim N(0, 1).$$
(2)
$$\frac{(m - 1)S_1^2}{\sigma_1^2} + \frac{(n - 1)S_2^2}{\sigma_2^2} \sim \chi_{m+n-2}^2.$$

Since
$$\sigma_2^2 = \lambda \sigma_1^2$$
, $Q = \frac{\frac{\bar{Y} - \bar{X} - (b-a)}{\sqrt{\sigma_1^2/m + \sigma_2^2/n}}}{\sqrt{\frac{1}{m+n-2} \left(\frac{(m-1)S_1^2}{\sigma_1^2} + \frac{(n-1)S_2^2}{\sigma_2^2}\right)}} = \frac{\frac{\bar{Y} - \bar{X} - (b-a)}{\sqrt{1/m + \lambda/n}}}{\sqrt{\frac{1}{m+n-2} \left((m-1)S_1^2 + (n-1)S_2^2/\lambda\right)}} \sim t_{m+n-2}$

is a pivot. Then a $1-\alpha$ confidence interval for b-a is

$$\bar{Y} - \bar{X} \pm \sqrt{\frac{(1/m + \lambda/n)((m-1)S_1^2 + (n-1)S_2^2/\lambda)}{(m+n-2)}} t_{m+n-2}(\alpha/2).$$

4.16

$$S_1^2/S_2^2 = 0.893$$
,查表得 $F_{m-1,n-1}(\alpha/2) = F_{9,9}(0.025) = 4.03$,

$$F_{m-1,n-1}(1-\alpha/2) = 1/F_{n-1,m-1}(\alpha/2) = 1/F_{9,9}(0.025) = 1/4.03,$$

则 σ_a^2/σ_b^2 的置信系数为95%的置信区间为

$$\left[\frac{S_1^2}{S_2^2} \cdot \frac{1}{F_{m-1,n-1}(\alpha/2)}, \frac{S_1^2}{S_2^2} \cdot F_{n-1,m-1}(\alpha/2)\right] = [0.222, 3.599]$$

4.19

[Wei] 4.19. We only consider the case n > 1.

(1) Let
$$Y_i = \frac{X_i - \theta_1}{\theta_2 - \theta_1} \sim U(0, 1), \ Z_i = \frac{\theta_2 - X_i}{\theta_2 - \theta_1} \sim U(0, 1), \ Q = \frac{(\theta_2 - \theta_1) - (X_{(n)} - X_{(1)})}{\theta_2 - \theta_1}$$
 then $Q = Y_{(1)} + Z_{(1)}$ and

$$f_{Y_{(1)},Z_{(1)}}(y,z) = f_{Y_{(1)},Y_{(n)}}(y,1-z) = n(n-1)(1-z-y)^{n-2}I_{(0,1)}(y,z,y+z).$$

Therefore, Q is a pivot with p.d.f.

$$f_Q(x) = \int f_{Y_{(1)}, Z_{(1)}}(y, x - y) dy = n(n - 1)x(1 - x)^{n-2} I_{(0,1)}(x).$$

We can observe that $Q \sim Beta(2, n-1)$.

Suppose $P(Q \in (1-c,1)) = 1-\alpha$ (this assumption is convenient for following analysis, you can generally suppose a confidence interval (a,b) instead), then $c = Beta_{n-1,2}(\alpha)$. Also notice that $Q = 1 - \frac{(X_{(n)} - X_{(1)})}{\theta_2 - \theta_1}$,

$$Q \in (1-c,1) \Leftrightarrow \theta_2 - \theta_1 \in ((X_{(n)} - X_{(1)})/c, \infty).$$

As a result, a $1 - \alpha$ confidence interval for $\theta_2 - \theta_1$ is

$$((X_{(n)}-X_{(1)})/Beta_{n-1,2}(\alpha),\infty).$$

(2) Hint: the objective statistic can be rewritten as $R := \frac{Y_{(1)} + Y_{(n)} - 1}{Y_{(n)} - Y_{(1)}}$, which is auxiliary. Notice that $(X_{(1)}, X_{(n)})$ is sufficient and complete for (θ_1, θ_2) . Basu's theorem tolds us $S := Y_{(n)} - Y_{(1)}$ is independent with R. Now write the p.d.f of $(Y_{(1)}, Y_{(n)})$. With the relationship

$$\begin{cases} Y_{(1)} = \frac{1}{2}(RS - S + 1) \\ Y_{(n)} = \frac{1}{2}(RS + S + 1) \end{cases}$$

write the p.d.f of (R, S). Consequently, $f_R = f_{(R,S)}/f_S$. The explicit form is very complicated, and should be discussed with nodes $0, \pm 1$.

(3) Using R as the pivot.

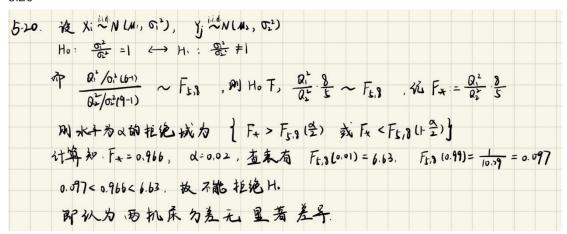
4.23

[Wei] 4.23.

- (1) $X_{(1)} \theta \sim Exp(n)$.
- (2) Using $Q := X_{(1)} \theta$ in (1) as a pivot, suppose the interval as (0, c) for Q. After calculation, the interval for θ is $(X_{(1)} + \log \alpha/n, X_{(1)})$.

12.1 作业

5.20



5.56

解: 对于
$$H_0: \sigma_1^2 = \sigma_2^2 \leftrightarrow H_1: \sigma_1^2 \neq \sigma_2^2, \diamondsuit F = \frac{S_1^2}{S_2^2} \cdot \frac{\sigma_2^2}{\sigma_1^2},$$
则有接受域

检验问题 $H_0:\sigma^2=\sigma_0^2\leftrightarrow H_1:\sigma^2\neq\sigma_0^2$ 的水平为 α 的接受域为

$$\overline{D} = \left\{ \boldsymbol{X} : \chi_{n-1}^2 \left(1 - \frac{\alpha}{2} \right) \leqslant \frac{(n-1)S^2}{\sigma_0^2} \leqslant \chi_{n-1}^2 \left(\frac{\alpha}{2} \right) \right\}$$

 H_0 成立时 $\sigma^2 = \sigma_0^2$,故可解得置信区间为

$$\left[\frac{(n-1)S^2}{\chi_{n-1}^2\left(\frac{\alpha}{2}\right)}, \frac{(n-1)S^2}{\chi_{n-1}^2\left(1-\frac{\alpha}{2}\right)}\right].$$

同理可解得置信上、下限为 $\frac{(n-1)S^2}{\chi^2_{n-1}(1-\alpha)}$, $\frac{(n-1)S^2}{\chi^2_{n-1}(\alpha)}$

对于检验问题 $H_0:\sigma_1^2/\sigma_2^2=c \leftrightarrow H_1:\sigma_1^2/\sigma_2^2\neq c$, $\frac{S_2^2}{S_1^2}\cdot c\sim F_{n-1,m-1}$,接受域为

$$\overline{D} = \left\{ (\boldsymbol{X}, \boldsymbol{Y}) : F_{n-1,m-1} \left(1 - \frac{\alpha}{2} \right) \leq \frac{S_2^2}{S_1^2} \cdot c \leq F_{n-1,m-1} \left(\frac{\alpha}{2} \right) \right\}$$

故置信区间为

$$\begin{bmatrix} S_1^2 \\ S_2^2 \end{bmatrix} \cdot F_{n-1,m-1}(1-\alpha/2), \frac{S_1^2}{S_2^2} \cdot F_{n-1,m-1}(\alpha/2)$$

对于检验问题 $H_0:\sigma_1^2/\sigma_2^2 \leq c \leftrightarrow H_1:\sigma_1^2/\sigma_2^2 > c$,接受域为

$$\overline{D} = \left\{ (\boldsymbol{X}, \boldsymbol{Y}) : \frac{S_2^2}{S_1^2} \leqslant \frac{F_{n-1,m-1}(\alpha)}{c} \right\} \Rightarrow 1 - \alpha = P\left(c \leqslant \frac{S_1^2}{S_2^2} \cdot F_{n-1,m-1}(\alpha)\right)$$

故置信上限为 $\frac{S_1^2}{S_2^2} \cdot F_{n-1,m-1}(\alpha)$

同理可得置信下限为 $\frac{S_1^2}{S_2^2} \cdot F_{n-1,m-1}(1-\alpha)$.

12.6 作业

作业 1

补充 1:
$$X_1$$
, ···; X_n i.i.d. $\sim N(M_1 \circ^2)$, $n > 4$, $M \in \mathbb{R}$, $\circ^* > 0$ 未知、 \overline{k}_M , $\overline{l}_1 = \overline{l}_2 = \overline{l}_1$ 的 $\overline{l}_1 = \overline{l}_2 = \overline{l}_1$ 的 $\overline{l}_2 = \overline{l}_2 = \overline{l}_2 = \overline{l}_2$ 的 $\overline{l}_3 = \overline{l}_4 = \overline{l}_4 = \overline{l}_4$ 的 $\overline{l}_4 = \overline{l}_4 = \overline{l}_4$

作业 2

补充
$$\geq$$
: $\chi_{i,i}$, χ_{n} i.i.d. $\sim f(x_{i})\lambda_{i} = e^{-(x_{i} \cdot \lambda_{i})} 1(x_{i} \cdot \lambda_{i})$, $\chi \in \mathbb{R}$. 成 $\chi \in \mathbb{R}$. $\chi \in$