

STAT5703 HW1 Exercise 1

Wen Fan(wf2255), Banruo Xie(bx2168), Hanjun Li(hl3339)

Exercise 1.

Question 1.

To compute p^{th} population quantile, we need to find point $Q_D(p)$ such that $P(D \leq Q_D(p)) = p$. So for the exponential distribution:

$$\begin{aligned}\int_0^{Q_D(p)} \lambda e^{-\lambda D} dD &= p = 1 - e^{-\lambda Q_D(p)} \\ 1 - p &= e^{-\lambda Q_D(p)} \\ \ln(1 - p) &= -\lambda Q_D(p) \\ Q_D(p) &= -\frac{1}{\lambda} \ln(1 - p)\end{aligned}$$

Question 2.

To find a method of moments-based estimator of $Q_D(p)$, we need to find MM of λ first, because $Q_D(p) = -\frac{1}{\lambda} \ln(1 - p)$. The first moment of D is:

$$\mu_1 = \mathbf{E}[D] = \frac{1}{\lambda} = \psi(\lambda)$$

The first empirical moment is:

$$\hat{\mu}_1 = \bar{D}_n$$

The MM estimator of λ is:

$$\hat{\lambda}^{MM} = \psi^{-1}(\bar{D}_n) = \frac{1}{\bar{D}_n}$$

Therefore the MM-based estimator of $Q_D(p)$:

$$\hat{Q}_D(p)^{MM} = -\frac{1}{\hat{\lambda}^{MM}} \ln(1 - p) = -\bar{D}_n \ln(1 - p)$$

.

Question 3.

We already know $D_1, \dots, D_n \stackrel{i.i.d.}{\sim} \text{Exp}(\lambda)$, hence by CLN,

$$\begin{aligned}\sqrt{n}(\bar{D}_n - \mu) &\xrightarrow[n \rightarrow \infty]{\mathcal{D}} \mathcal{N}(0, \sigma^2) \\ \sqrt{n}(\bar{D}_n - \frac{1}{\lambda}) &\xrightarrow[n \rightarrow \infty]{\mathcal{D}} \mathcal{N}(0, \frac{1}{\lambda^2})\end{aligned}$$

Then by Delta Method,

$$\sqrt{n}(\ln(1 - p)\bar{D}_n + Q_D(p)) \xrightarrow[n \rightarrow \infty]{\mathcal{D}} \mathcal{N}(0, \frac{(\ln(1 - p))^2}{\lambda^2})$$

From question 2, we already obtain the method of moments estimator of $Q_D(p)$ is $\hat{Q}_D(p)^{MM} = -\bar{D}_n \ln(1 - p)$, hence

$$\frac{\sqrt{n}\lambda(Q_D(p) - \hat{Q}_D(p)^{MM})}{\ln(1 - p)} \xrightarrow[n \rightarrow \infty]{\mathcal{D}} \mathcal{N}(0, 1)$$

For *approximate* $(1 - \alpha)$ -confidence interval, we can get:

$$L(D) = \hat{Q}_D(p)^{MM} - \frac{z_{1-\alpha/2} \ln(1-p)}{\lambda\sqrt{n}}$$

$$U(D) = \hat{Q}_D(p)^{MM} + \frac{z_{1-\alpha/2} \ln(1-p)}{\lambda\sqrt{n}}$$

the *approximate* $(1 - \alpha)$ -confidence interval for $Q_D(p)$ is $[\hat{Q}_D(p)^{MM} - \frac{z_{1-\alpha/2} \ln(1-p)}{\lambda\sqrt{n}}, \hat{Q}_D(p)^{MM} + \frac{z_{1-\alpha/2} \ln(1-p)}{\lambda\sqrt{n}}]$

Question 4.

Since $D_1, \dots, D_n \stackrel{i.i.d.}{\sim} \text{Exp}(\lambda)$, from addition rule for the exponential distribution, $\sum_{i=1}^n D_i \sim \Gamma(n, \lambda)$, then

$$\lambda \bar{D}_n = \frac{\lambda}{n} \sum_{i=1}^n D_i \sim \Gamma(n, n)$$

Because distribution of $\lambda \bar{D}_n$ is independent of the parameter λ , $\lambda \bar{D}_n$ is an exact pivot.

To construct an exact confidence interval of median, we can get the confidence interval of λ first. We know that

$$2n\lambda\bar{D}_n \sim \Gamma(n, \frac{1}{2}) \sim \chi_{2n}^2$$

Therefore, $P(\chi_{1-\alpha/2, 2n}^2 < 2n\lambda\bar{D}_n < \chi_{\alpha/2, 2n}^2) = 1 - \alpha$ for any $\alpha \in (0, 1)$

$$P(\frac{\chi_{1-\alpha/2, 2n}^2}{2n\bar{D}_n} < \lambda < \frac{\chi_{\alpha/2, 2n}^2}{2n\bar{D}_n}) = 1 - \alpha$$

Because $Q_D(0.5) = -\frac{1}{\lambda} \ln(0.5)$,

$$P(\frac{-2n\bar{D}_n \ln(0.5)}{\chi_{1-\alpha/2, 2n}^2} < Q_D(0.5) < \frac{-2n\bar{D}_n \ln(0.5)}{\chi_{\alpha/2, 2n}^2}) = 1 - \alpha$$

Hence, the $(1 - \alpha)$ exact confidence interval of the median $Q_D(0.5)$ is,

$$[\frac{-2n\bar{D}_n \ln(0.5)}{\chi_{1-\alpha/2, 2n}^2}, \frac{-2n\bar{D}_n \ln(0.5)}{\chi_{\alpha/2, 2n}^2}]$$