

Question 1

有

$$\begin{aligned} L(\theta) &= \prod_{i=1}^n 1_{x_i \geq \theta} \cdot \frac{\theta}{x_i^2} \\ &= 1_{x_1 \geq \theta, \dots, x_n \geq \theta} \cdot \frac{\theta^n}{\prod x_i^2}, \end{aligned}$$

于是 θ 的最大似然估计为 $\hat{\theta} = \min\{x_1, \dots, x_n\}$.

对于任意 $\varepsilon > 0$, 考虑 $\lim_{n \rightarrow \infty} P(|\hat{\theta}_n - \theta| > \varepsilon)$. 记 $p := P(X \leq \theta + \varepsilon) > 0$, 则

$$P(|\hat{\theta}_n - \theta| > \varepsilon) = (1 - p)^n \rightarrow 0,$$

于是 $\hat{\theta}$ 是渐进无偏估计量. 显然 $\mathbb{E}(\hat{\theta}) > \theta$, $\hat{\theta}$ 不是无偏估计量.

下面计算均方误差, 先算 $\mathbb{E}(\hat{\theta})$ 和 $\mathbb{E}(\hat{\theta}^2)$.

$$\begin{aligned} F_{\hat{\theta}}(t) &= 1 - P(\hat{\theta} \geq t) = 1 - \left(\frac{\theta}{t}\right)^n, \\ f_{\hat{\theta}}(t) &= F'_{\hat{\theta}}(t) = -n \left(\frac{\theta}{t}\right)^{n-1} \cdot \left(-\frac{\theta}{t^2}\right) = \frac{n\theta^n}{t^{n+1}}, \\ \mathbb{E}(\hat{\theta}) &= \int_{\theta}^{+\infty} t f(t) dt = n\theta^n \int_{\theta}^{+\infty} t^{-n} dt = \frac{n}{n-1} \theta, \\ \mathbb{E}(\hat{\theta}^2) &= \int_{\theta}^{+\infty} t^2 f(t) dt = n\theta^n \int_{\theta}^{+\infty} t^{-n+1} dt = \frac{n}{n-2} \theta^2. \end{aligned}$$

于是

$$\begin{aligned} \text{MSE}(\hat{\theta}) &= \mathbb{E}((\hat{\theta} - \theta)^2) = \mathbb{E}(\hat{\theta}^2) - 2\theta \cdot \mathbb{E}(\hat{\theta}) + \theta^2 \\ &= \left(1 + \frac{n}{n-2} - \frac{2n}{n-1}\right) \theta^2 \\ &= \frac{2\theta^2}{n^2 - 3n + 2}. \end{aligned}$$

由 $\text{MSE}(\hat{\theta}) \rightarrow 0$ 知 $\hat{\theta}$ 是一致估计量.

考虑枢轴量 $G = \hat{\theta}/\theta$,

$$F_G(t) = F_{\hat{\theta}}(\theta t) = 1 - t^{-n}, \quad t \geq 1.$$

取 $t = \alpha^{-\frac{1}{n}}$ 即有 $P(\theta > t^{-1}\hat{\theta}) = F_G(t) = 1 - \alpha$, $[\alpha^{\frac{1}{n}}\hat{\theta}, \hat{\theta}]$ 即为置信度为 $1 - \alpha$ 的置信区间.

Question 2

1. 已知 $\sum x_i \sim \Gamma(n, \lambda)$, 于是有

$$\begin{aligned}\mathbb{E}(\hat{\lambda}_0) &= n \cdot \int_0^{+\infty} \frac{\lambda^n}{\Gamma(n)} x^{n-1} e^{-\lambda x} \cdot \frac{1}{x} dx \\ &= n\lambda \cdot \int_0^{+\infty} \frac{\lambda^{n-1}}{\Gamma(n)} x^{n-2} e^{-\lambda x} dx \\ &= n\lambda \cdot \frac{\Gamma(n-1)}{\Gamma(n)} \\ &= \frac{n}{n-1} \lambda.\end{aligned}$$

于是 $\hat{\lambda}_0$ 不是无偏估计量, 是渐进无偏估计量.

2. $\hat{\lambda}_1 := \frac{n-1}{n} \hat{\lambda}_0$ 是无偏估计量.

3. $\hat{\lambda}_1$ 的均方误差即为自身方差. 考虑计算 $\mathbb{E}(\hat{\lambda}_0^2)$.

$$\begin{aligned}\mathbb{E}(\hat{\lambda}_1^2) &= n^2 \cdot \int_0^{+\infty} \frac{\lambda^n}{\Gamma(n)} x^{n-1} e^{-\lambda x} \cdot \frac{1}{x^2} dx \\ &= \frac{n^2}{(n-1)(n-2)} \lambda^2.\end{aligned}$$

于是

$$\begin{aligned}\text{MSE}(\hat{\lambda}_1) &= \frac{(n-1)^2}{n^2} \text{Var}(\hat{\lambda}_0^2) \\ &= \frac{(n-1)^2}{n^2} \left(\frac{n^2}{(n-1)(n-2)} \lambda^2 - \frac{n^2}{(n-1)^2} \lambda^2 \right) \\ &= \frac{1}{n-2} \lambda^2.\end{aligned}$$

$\text{MSE}(\hat{\lambda}_1) \rightarrow 0$, 于是它是一致估计量.

4. 考虑 $2\lambda X_i$ 服从的分布 $g: g(x) = \frac{f(\frac{x}{2\lambda})}{2\lambda} = \frac{1}{2} e^{-\frac{x}{2}}$, 即 $2\lambda X_i \sim \text{Exp}(\frac{1}{2})$, $2\lambda \sum X_i \sim \chi^2(n)$. 令 $t_0 := F^{-1}(1-\alpha)$, 则

$$P\left(0 \leq \lambda \leq \frac{t_0}{2n\bar{X}}\right) = P\left(0 \leq 2\lambda \sum X_i \leq t_0\right) = 1 - \alpha.$$

$\left[0, \frac{F^{-1}(1-\alpha)}{2n\bar{X}}\right]$ 即为所求.

Question 3

1. 有

$$F_X(x) = \int_0^x (\theta + 1)t^\theta dt = x^{\theta+1}, \quad \forall 0 < x < 1. \text{ 于是}$$

$$F_X(x) = \begin{cases} 0 & \text{if } x \leq 0, \\ x^{\theta+1} & \text{if } 0 < x < 1, \\ 1 & \text{if } x \geq 1. \end{cases}$$

2. 有

$$\mathbb{E}(X) = \int_0^1 (\theta + 1)x^\theta \cdot x dx = \frac{\theta + 1}{\theta + 2},$$

即

$$\theta = \frac{1}{1 - \mathbb{E}(X)} - 2.$$

替换后即得

$$\hat{\theta}_1 = \frac{1}{1 - \bar{X}} - 2.$$

3. 对概率密度函数取对数:

$$\begin{aligned} \sum_{i=1}^n \ln f(x_i) &= \sum_{i=1}^n \ln(\theta + 1) + \theta \ln x_i \\ &= n \ln(\theta + 1) + \theta \sum \ln x_i. \end{aligned}$$

取导数得到

$$\frac{n}{\theta + 1} + \sum \ln x_i = 0,$$

即

$$\hat{\theta}_2 = -\frac{n}{\sum \ln x_i} - 1.$$

考虑 $\ln X_i$ 服从的分布: $f_{\ln X}(x) = f_X(e^x) \cdot e^x = (\theta + 1)e^{(\theta+1)x}$, 于是 $-\ln X$ 服从 $\text{Exp}(\theta + 1)$, $-\sum \ln X_i$ 服从 $\Gamma(n, \theta + 1)$. 代入计算有

$$\begin{aligned} \mathbb{E}(\hat{\theta}_2) &= -1 + \int_0^{+\infty} \frac{n}{x} \cdot \frac{(\theta + 1)^n}{\Gamma(n)} x^{n-1} e^{-(\theta+1)x} dx \\ &= -1 + n(\theta + 1) \cdot \frac{\Gamma(n-1)}{\Gamma(n)} \\ &= -1 + \frac{n}{n-1}(\theta + 1) \\ &= \frac{n}{n-1}\theta + \frac{1}{n-1}. \end{aligned}$$

于是 $\hat{\theta}_2$ 不是无偏估计量, 是渐进无偏估计量.

4. 约定 $K := -\sum \ln X_i \sim \Gamma(n, \theta + 1)$. 有

$$\mathbb{E}\left((\hat{\theta}_2 - \theta)^2\right) = \mathbb{E}\left((\hat{\theta}_2)^2\right) - 2\theta\mathbb{E}(\hat{\theta}_2) + \theta^2,$$

其中

$$\begin{aligned}\mathbb{E}\left((\hat{\theta}_2)^2\right) &= \mathbb{E}\left(\frac{n^2}{K^2}\right) - 2\mathbb{E}\left(\frac{n}{K}\right) + 1 \\ &= \mathbb{E}\left(\frac{n^2}{K^2}\right) - 2\frac{n(\theta+1)}{n-1} + 1,\end{aligned}$$

而

$$\begin{aligned}\mathbb{E}\left(\frac{n^2}{K^2}\right) &= n^2 \int_0^{+\infty} \frac{1}{k^2} \cdot \frac{(\theta+1)^n}{\Gamma(n)} k^{n-1} e^{-(\theta+1)k} dk \\ &= n^2 \cdot \frac{(\theta+1)^2}{(n-2)(n-1)}.\end{aligned}$$

于是

$$\begin{aligned}\text{MSE}(\hat{\theta}_2) &= \frac{n^2}{(n-2)(n-1)}(\theta+1)^2 - \left(\frac{2n(\theta+1)}{n-1}\right) + 1 - 2\theta \cdot \left(\frac{n}{n-1}\theta + \frac{1}{n-1}\right) + \theta^2 \\ &= \left(\frac{n^2}{(n-2)(n-1)} - \frac{2n}{n-1} + 1\right)\theta^2 + \left(\frac{2n^2}{(n-2)(n-1)} - \frac{2n}{n-1} - \frac{2}{n-1}\right)\theta \\ &\quad + \left(\frac{n^2}{(n-2)(n-1)} - \frac{2n}{n-1} + 1\right) \\ &= \frac{(\theta+1)^2(n+2)}{(n-2)(n-1)}.\end{aligned}$$

有 $\text{MSE}(\hat{\theta}_2) \rightarrow 0$, 于是它是一致估计量.

Question 4

1. 有

$$\begin{aligned}\sum \ln P(X = x_i) &= \sum \ln\left(\frac{\lambda^{x_i}}{x_i!} e^{-\lambda}\right) \\ &= \ln \lambda \cdot \sum x_i - n\lambda - \sum \ln(x_i!),\end{aligned}$$

求导有

$$\frac{1}{\lambda} \cdot \sum x_i - n = 0,$$

于是 \bar{X} 确实是 λ 最大似然估计, 于是 \hat{p}_1 亦是 $e^{-\lambda}$ 最大似然估计.

记 $Y := \sum_{i=1}^n X_i$, 则 $Y \sim \pi(n\lambda)$. 利用提示中的结论有

$$\mathbb{E}(\hat{p}_1) = \mathbb{E}\left(e^{-\frac{1}{n}Y}\right) = e^{n\lambda\left(e^{-\frac{1}{n}}-1\right)}.$$

于是 \hat{p}_1 不是无偏估计量. 利用泰勒展开

$$e^{-\frac{1}{n}} - 1 = -\frac{1}{n} + O\left(\frac{1}{n^2}\right)$$

可知 \hat{p}_1 是渐进无偏估计量.

计算 \hat{p}_1 的均方误差:

$$\begin{aligned}\mathbb{E}\left((\hat{p}_1 - p)^2\right) &= \mathbb{E}(\hat{p}_1^2) - 2p\mathbb{E}(\hat{p}_1) + p^2 \\ &= e^{n\lambda\left(e^{-\frac{2}{n}}-1\right)} - 2e^{-\lambda+n\lambda\left(e^{-\frac{1}{n}}-1\right)} + e^{-2\lambda}.\end{aligned}$$

当 $n \rightarrow \infty$ 时 $\text{MSE}(\hat{p}_1) \rightarrow 0$, 于是 \hat{p}_1 是一致估计量.

2. 有

$$\mathbb{E}(\hat{p}_2) = \mathbb{E}(1_{X_i=0}) = e^{-\lambda},$$

于是 \hat{p}_2 是无偏估计量, 自然也是渐进无偏估计量. 而

$$\begin{aligned}\text{MSE}(\hat{p}_2) &= \text{Var}(\hat{p}_2) \\ &= \frac{1}{n} \text{Var}(1_{X_i=0}) \\ &= \frac{1}{n} e^{-\lambda}(1 - e^{-\lambda}).\end{aligned}$$

于是 \hat{p}_2 是一致估计量.