

Chapter 6

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Problem 1.

$$\begin{aligned}
 X_i &\sim \text{Gamma}(\alpha, \beta) \Rightarrow \mathbb{E}[X_i] = \frac{\alpha}{\beta} = \bar{X}_n \\
 \mathbb{E}[X_i^2] &= \text{Var}(X_i) + (\mathbb{E}[X_i])^2 = \frac{\alpha}{\beta^2} + \frac{\alpha^2}{\beta^2} = \frac{\alpha^2 + \alpha}{\beta^2} \\
 &\Rightarrow \frac{1}{n} \sum_{i=1}^n X_i^2 = \frac{\hat{\alpha}^2 + \hat{\alpha}}{\hat{\beta}^2} \\
 \Rightarrow \hat{\alpha} &= \frac{\bar{X}_n^2}{\frac{1}{n} \sum_{i=1}^n X_i^2 - \bar{X}_n^2} \text{ and } \hat{\beta} = \frac{\bar{X}_n}{\frac{1}{n} \sum_{i=1}^n X_i^2 - \bar{X}_n^2}
 \end{aligned}$$

Problem 2. a)

$$X_i \sim \text{Uniform}(a, b) \Rightarrow$$

$$\begin{cases} \mathbb{E}[X_i] = \frac{a+b}{2} \\ \mathbb{E}[X_i^2] = \frac{a^2+ab+b^2}{3} \end{cases} \quad (1)$$

$$\text{Let } \hat{\alpha}_1 = \mathbb{E}[X_i] \text{ and } \hat{\alpha}_2 = \mathbb{E}[X_i^2] \Rightarrow$$

$$\begin{cases} \hat{\alpha}_1 = \frac{a+b}{2} \\ \hat{\alpha}_2 = \frac{a^2+ab+b^2}{3} \end{cases} \quad (2)$$

Solve the above equations, we obtain:

$$\begin{cases} \hat{\alpha} = \hat{\alpha}_1 - \sqrt{3(\hat{\alpha}_2 - \hat{\alpha}_1^2)} \\ \hat{\beta} = \hat{\alpha}_1 + \sqrt{3(\hat{\alpha}_2 - \hat{\alpha}_1^2)} \end{cases} \quad (3)$$

b.)

The likelihood :

$$\mathbb{L}(a, b) = \prod_{i=1}^n \frac{1}{b-a} \mathbb{I}_{(a,b)} = \frac{1}{(b-a)^n} \mathbb{I}_{(-\infty, x_1)}(a) \mathbb{I}_{(x_n, +\infty)}(b)$$

To maximize the likelihood: $a = X_{(1)}$ and $b = X_{(n)}$

c.)

We have: $\tau = \int x dF(x) = \mathbb{E}[x] = \frac{a+b}{2}$, by the equivariance property:

$$\hat{\tau} = \frac{\hat{\alpha} + \hat{\beta}}{2} = \frac{X_{(1)} + X_{(2)}}{2}$$

d.) Update

Problem 4. $X_i \sim Uniform(0, \theta) \Rightarrow$ Likelihood of X:

$$\mathcal{L} = \prod_{i=1}^n f(x) = \prod_{i=1}^n \frac{1}{\theta} I(0 \leq X_i \leq \theta) = \frac{1}{\theta^n} I(0 \leq X_1, X_2, \dots, X_n \leq \theta)$$

$\Rightarrow \mathcal{L}$ max when $\theta = \hat{\theta} = X_{(n)}$

We have:

$$\mathbb{P}(|\hat{\theta} - \theta| < \epsilon) = \mathbb{P}(\hat{\theta} < \theta - \epsilon) + \mathbb{P}(\hat{\theta} > \theta + \epsilon) = \mathbb{P}(\hat{\theta} < \theta - \epsilon) = \left(\frac{\theta - \epsilon}{\theta}\right)^n \rightarrow 0$$

MLE is consistent