Probability And Mathmatical Statistics Homework 10

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3

解得

矩估计

$$EX^{2} = \int_{0}^{\infty} \frac{x}{\sqrt{2\pi}\sigma} e^{-\frac{(\ln x - \mu)^{2}}{2\sigma^{2}}} dx$$

$$\Leftrightarrow t = \frac{\ln x - \mu}{\sqrt{2}\sigma}$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{\pi}} e^{2\sqrt{2}\sigma t + 2\mu - t^{2}} dt$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{\pi}} e^{-(t - \sqrt{2}\sigma)^{2}} e^{2\sigma^{2} + 2\mu} dt$$

$$= \frac{e^{2\sigma^{2} + 2\mu}}{\sqrt{\pi}} \int_{-\infty}^{+\infty} e^{-(t - \sqrt{2}\sigma)^{2}} dt$$

$$= e^{2\sigma^{2} + 2\mu}$$

 $\hat{\mu} = \frac{4ln\bar{X} - ln\frac{\sum_{i=1}^{n} X_{i}^{2}}{n}}{2}$ $\hat{\sigma^{2}} = ln\frac{\sum_{i=1}^{n} X_{i}^{2}}{n} - 2ln\bar{X}$

极大似然估计

似然函数为

$$\begin{split} L(\mu,\sigma^2) &= \Pi_{i=1}^n (2\pi\sigma^2)^{-\frac{1}{2}} X_i^{-1} e^{-\frac{(\ln X_i - \mu)^2}{2\sigma^2}} \\ &= (2\pi\sigma^2)^{-\frac{n}{2}} \frac{1}{\Pi_{i=1}^n X_i} e^{-\frac{\sum_{i=1}^n (\ln X_i - \mu)^2}{2\sigma^2}} \\ \ln L(\mu,\sigma^2) &= -\frac{n}{2} ln(2\pi\sigma^2) + ln\frac{1}{\Pi_{i=1}^n X_i} - \frac{\sum_{i=1}^n (\ln X_i - \mu)^2}{2\sigma^2} \end{split}$$

$$\begin{cases} \frac{\partial lnL}{\partial \sigma^2} &= -\frac{n}{2} \frac{2\pi}{2\pi\sigma^2} + \frac{\sum_{i=1}^{n} (lnX_i - \mu)^2}{2(\sigma^2)^2} = 0 \\ \frac{\partial lnL}{\partial \mu} &= -\frac{\sum_{i=1}^{n} 2(\mu - lnX_i)}{2\sigma^2} = 0 \end{cases}$$
(1)

解得

$$\hat{\mu} = \frac{\sum_{i=1}^{n} ln X_i}{n}$$

$$\hat{\sigma^2} = \frac{\sum_{i=1}^{n} \left[ln X_i - \sum_{j=1}^{n} \frac{ln X_j}{n} \right]}{n}$$

4

使用样本矩代替标准矩,得

$$\begin{cases} \bar{X} &= e^{\mu + \frac{\sigma^2}{2}} \\ \frac{1}{n} \sum_{i=1}^n X_i^2 &= e^{2\mu + 2\sigma^2} \end{cases}$$

矩估计

极大似然估计

无偏估计意味着 $E[.] = \sigma^2$.

$$E[.] = c \sum_{i=1}^{n-1} E(X_{i+1} - X_i)^2$$

$$= c \sum_{i=1}^{n-1} \left[D(X_{i+1} - X_i) + (E(X_{i+1} - X_i))^2 \right]$$

$$= c(n-1)2\sigma^2$$

$$= \sigma^2$$

$$\therefore c = \frac{1}{2(n-1)}$$

$$\begin{split} L(\sigma^2) &= \Pi_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(X_i - \mu)^2}{2\sigma^2}} \\ lnL(\sigma^2) &= \sum_{i=1}^n -\frac{(X_i - \mu)^2}{2\sigma^2} + ln\frac{1}{\sqrt{2\pi\sigma^2}} \\ \frac{dlnL(\sigma^2)}{d\sigma^2} &= \sum_{i=1}^n \frac{(x - \mu)^2}{2\sigma^4} + \sqrt{2\pi\sigma^2} (-\frac{1}{2\pi\sigma^2}) \frac{\sqrt{2\pi}}{2\sqrt{\sigma^2}} \\ &= \sum_{i=1}^n \frac{(x - \mu)^2}{2\sigma^4} - \frac{1}{2\sigma^2} \\ & \Leftrightarrow \frac{dlnL(\sigma^2)}{d\sigma^2} = 0 \\ & \frac{\sum_{i=1}^n (X_i - \mu)}{2\sigma^2} = \frac{n}{2\sigma^2} \\ & \hat{\sigma^2} = \frac{1}{n} \sum_{i=1}^n (X_i - \mu)^2 \end{split}$$