

HW 6

$$1 \quad E(\bar{x}) = \frac{E(x_1) + E(x_2) + \dots + E(x_n)}{n} = \frac{n \cdot \lambda}{n} = \lambda$$

$$\begin{aligned} E(s^2) &= E \left[\frac{1}{n-1} \sum (x_i - \bar{x})^2 \right] \\ &= \frac{1}{n-1} E \sum (x_i - \mu - (\bar{x} - \mu))^2 \\ &= \frac{1}{n-1} E \left[\sum (x_i - \mu)^2 + n(\bar{x} - \mu)^2 - 2(\bar{x} - \mu) \sum (x_i - \mu) \right] \\ &= \frac{1}{n-1} E \left[\sum (x_i - \mu)^2 + n(\bar{x} - \mu)^2 - 2n(\bar{x} - \mu)^2 \right] \\ &= \frac{1}{n-1} E \left[\sum (x_i - \mu)^2 - n(\bar{x} - \mu)^2 \right] \\ &= \frac{1}{n-1} \left[n\sigma^2 - n \frac{\sigma^2}{n} \right] = \sigma^2 \end{aligned}$$

$$\begin{aligned} \text{Var}(\bar{x}) &= \frac{\text{Var}(x_1) + \text{Var}(x_2) + \dots + \text{Var}(x_n)}{n} \\ &= \frac{n^2 \lambda}{n} = n\lambda \end{aligned}$$

$$\begin{aligned} \text{Var}(s^2) &= \text{Var} \left(\frac{(n-1) \cdot \frac{\sigma^2}{n-1} s^2}{(n-1)} \right) \\ &= \text{Var} \left(\frac{\sigma^2}{n-1} (x_{n-1}^2) \right) \\ &= \left(\frac{\sigma^2}{n-1} \right)^2 \cdot \text{Var}(x_{n-1}^2) = \left(\frac{\sigma^2}{n-1} \right)^2 \cdot 2(n-1) \\ &= \frac{\sigma^4}{(n-1)^2} \cdot 2(n-1) = \frac{2\sigma^4}{n-1} \end{aligned}$$

Since $n\lambda$ is closer to λ than $\frac{2\sigma^4}{n-1}$

then, sample mean \bar{x} is better estimator

b. \bar{X} and cS are unbiased estimator of θ

$$E(\bar{X}) = E(cS) = \theta$$

$$E(\alpha\bar{X} + (1-\alpha)cS) = \alpha E(\bar{X}) + (1-\alpha)E(cS)$$

$$= \alpha\theta + \theta - \theta\alpha = \theta$$

\hookrightarrow unbiased estimator

$$\text{Var}(\alpha\bar{X} + (1-\alpha)cS) = \alpha^2 \text{Var}(\bar{X}) + (1-\alpha)^2 c^2 \text{Var}(S)$$

$$\text{Var}(S) = \text{Var}\left[\frac{\sqrt{X_{n-1}^2} \sigma}{\sqrt{n-1}}\right] = \frac{\sigma^2}{n-1} \text{Var}(\sqrt{X_{n-1}})$$

$$\Gamma\left(\frac{n-1}{2}, 2\right) \quad k = 1/2$$

$$E(\sqrt{X_{n-1}}) = \frac{\Gamma\left(\frac{n-1}{2} + \frac{1}{2}\right) \sqrt{2}}{\Gamma\left(\frac{n-1}{2}\right)} = \frac{\sqrt{2} \Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)}$$

$$\text{Var}(\sqrt{X_{n-1}}) = 2 E(\sqrt{X_{n-1}}) = \frac{2\sqrt{2} \Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)}$$

$$\text{Var}[cS] = c^2 \text{Var}[S]$$

$$= c^2 \text{Var}\left[\frac{\sqrt{X_{n-1}^2} \sigma}{\sqrt{n-1}}\right]$$

$$= \frac{c^2 \sigma^2}{n-1} \left(\frac{2\sqrt{2} \Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)} \right)$$

$$= \frac{\sigma^2}{n-1} \left(\frac{\sqrt{n-1} \Gamma\left(\frac{n-1}{2}\right)}{\Gamma\left(\frac{n}{2}\right)} \right)^2 \frac{2\sqrt{2} \Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)}$$

$$= \frac{2\sqrt{2} \sigma^2 \Gamma\left(\frac{n-1}{2}\right)}{\Gamma\left(\frac{n}{2}\right)}$$

$$\text{Var}[cS] = c^2 \text{Var}[S] = c^2 \text{Var}\left[\frac{\sqrt{X_{n-1}^2} \sigma}{\sqrt{n-1}}\right]$$

$$= \frac{c^2}{n-1} \sigma^2 \text{Var}(\sqrt{X_{n-1}^2})$$

$$= \frac{\theta^2}{n-1} \left(\frac{\Gamma\left(\frac{n-1}{2}\right) (n-1)}{2 \Gamma\left(\frac{n}{2}\right)} \right)^2 \frac{2\sqrt{2} \Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)} = \frac{\theta^2 (n-1) \Gamma\left(\frac{n-1}{2}\right)}{2 \Gamma(n/2)} \sqrt{2}$$

$$= \frac{\theta^2 c \sqrt{2}}{2}$$

HW 6

$$c \quad E\left(\frac{1}{\bar{x}}\right) = E\left(\frac{1}{\frac{x_1 + x_2 + \dots + x_n}{n}}\right) = E\left(\frac{n}{x_1 + x_2 + \dots + x_n}\right) = n \times \frac{1}{E(x_1) + E(x_2) + \dots + E(x_n)}$$

$$= \frac{n}{n(\alpha\beta)} = \frac{1}{\alpha\beta}$$

$$E\left(\frac{\alpha}{\bar{x}}\right) = \alpha E\left(\frac{1}{\bar{x}}\right) = \alpha \cdot \frac{1}{\alpha\beta} = \frac{1}{\beta}$$

Thus, $\frac{\alpha}{\bar{x}}$ is unbiased estimator of $\frac{1}{\beta}$

$$d \quad \sum_{i=1}^4 (x_i - \bar{x})^2 = \frac{(x_1 - x_2)^2}{2} + \frac{\left(x_3 - \frac{(x_1 + x_2)}{2}\right)^2}{\frac{3}{2}} + \frac{\left(x_4 - \frac{(x_1 + x_2 + x_3)}{3}\right)^2}{\frac{4}{3}}$$

$$\bullet E(x_1 - x_2) = E(x_1) - E(x_2) = 0 \quad \text{since } x_1, x_2 \text{ iid} \quad \left. \begin{array}{l} \\ \end{array} \right\} \frac{(x_1 - x_2)^2}{2} \sim \chi_1^2$$

$$\bullet \text{Var}(x_1 - x_2) = \text{Var}(x_1) + \text{Var}(x_2) = 2$$

$$\bullet E\left(x_3 - \frac{(x_1 + x_2)}{2}\right) = E(x_3) - \frac{1}{2}E(x_1) - \frac{1}{2}E(x_2) = 0$$

$$\text{Var}\left(x_3 - \frac{(x_1 + x_2)}{2}\right) = \text{Var}(x_3) + \frac{1}{4}\text{Var}(x_1) + \frac{1}{4}\text{Var}(x_2) = \frac{3}{2}$$

$$\left. \begin{array}{l} \\ \end{array} \right\} \frac{\left(x_3 - \frac{(x_1 + x_2)}{2}\right)^2}{\frac{3}{2}} \sim \chi_1^2$$

$$\bullet E\left(x_4 - \frac{(x_1 + x_2 + x_3)}{3}\right) = 0$$

$$\text{Var}\left(x_4 - \frac{(x_1 + x_2 + x_3)}{3}\right) = \text{Var}(x_4) + \frac{1}{9}(\text{Var}(x_1) + \text{Var}(x_2) + \text{Var}(x_3)) = \frac{4}{3}$$

$$\left. \begin{array}{l} \\ \end{array} \right\} \frac{\left(x_4 - \frac{(x_1 + x_2 + x_3)}{3}\right)^2}{\frac{4}{3}} \sim \chi_1^2$$

$$\rightarrow XX' = \begin{bmatrix} 1 & -1 & 0 & 0 \\ -1/2 & -1/2 & 1 & 0 \\ -1/3 & -1/3 & -1/3 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & -1/2 & -1/3 \\ -1 & -1/2 & -1/3 \\ 0 & 1 & -1/3 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 3/2 & -1/3 \\ 0 & -1/3 & 4/3 \\ 0 & 0 & 1 \end{bmatrix} \frac{4}{3}$$

→ Independence