

$$F_1(x) = \frac{\sum_{i=1}^n \mathbb{1}_{x_i \leq x}}{n}$$

$$\theta_0 = \theta(F) = \mu^3$$

Question 1:

$$1): \quad \hat{\theta} = \theta(F_1) = \left[\int x dF_1(x) \right]^3 = \left[\int x \left(\frac{1}{n} \sum_{i=1}^n \delta_{(x_i - x)} dx \right) \right]^3 = \left[\frac{1}{n} \sum_{i=1}^n x_i \right]^3$$

where F_1 is empirical distribution. $= \bar{x}^3$

2):

$$b_1 = \mathbb{E}_{F_0}(\hat{\theta} - \theta_0) = \mathbb{E}_{F_0}(\hat{\theta}) - \theta_0 = \mathbb{E}_{F_0}(\bar{x}^3) - \theta_0$$

$$\text{Let } \sigma^2 = \text{Var}(x_i) = \mathbb{E}[(x_i - \mu)^2] \quad [I \text{ drop } F_0 \text{ for convenience}]$$

$$\mu = \mathbb{E}[(x_i - \mu)^3]$$

$$\mathbb{E}[\bar{x}^3] = \mathbb{E}[(x_i - \mu + \mu)^3]$$

$$\begin{aligned} &= \mathbb{E}[\mu^3 + 3\mu^2(x_i - \mu) + 3\mu(\bar{x} - \mu)^2 + (\bar{x} - \mu)^3] \\ &= \underbrace{\mathbb{E}[\mu^3]}_{\mu^3} + 3\mu^2 \underbrace{\mathbb{E}[x_i - \mu]}_{=0} + 3\mu \underbrace{\mathbb{E}[(\bar{x} - \mu)^2]}_A + \underbrace{\mathbb{E}[(\bar{x} - \mu)^3]}_B \end{aligned}$$

$$A = \text{Var}(\bar{x}) = \text{Var}\left(\frac{1}{n} \sum_{i=1}^n x_i\right) = \frac{1}{n^2} \times n \times \text{Var}(x_i) = \frac{\sigma^2}{n}$$

$$\begin{aligned} B &= \frac{\mathbb{E}[(n\bar{x} - n\mu)^3]}{n^3} = \frac{1}{n^3} \mathbb{E}\left[\left(\sum_{i=1}^n x_i - n\mu\right)^3\right] \\ &= \frac{1}{n^3} \mathbb{E}\left[\left(\sum_{i=1}^n (x_i - \mu)\right)^3\right] \end{aligned}$$

$$\begin{aligned}
&= \frac{1}{n^3} \mathbb{E} \left(\sum_{i=1}^n (x_i - u)^3 + \sum_{i \neq j}^n (x_i - u)^2 (x_j - u) \right) \\
&= \frac{1}{n^3} \sum_{i=1}^n \mathbb{E} [(x_i - u)^3] + \sum_{i \neq j} \mathbb{E} (x_i - u)^2 \underbrace{\mathbb{E} (x_j - u)}_{=0} \\
&= \frac{1}{n^3} \sum_{i=1}^n r \\
&= \frac{r}{n^2}
\end{aligned}$$

$$\Rightarrow b_1 = u^3 + \frac{3u\sigma^2}{n} + \frac{r}{n^2} - u^3 = \frac{3u\sigma^2}{n} + \frac{r}{n^2}$$

3): Bootstrap estimate of b_1 ,

$$\hat{b}_1 = \mathbb{E}_{F_1} [\theta(F_2) - \theta(F_1)]$$

$$= \mathbb{E}_{F_1} [\theta(F_2)] - \theta(F_1)$$

$$= \frac{3\bar{x}\hat{\sigma}^2}{n} + \frac{\hat{r}}{n^2} \quad (\text{just replace } u \mapsto \bar{x}, \sigma^2 \mapsto \hat{\sigma}^2, n \mapsto \hat{n}.)$$

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i, \quad \hat{\sigma}^2 = \int (x - \bar{x})^2 dF_1 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$$

$$\hat{r} = \int (x - \bar{x})^3 dF_1 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^3$$

4):

Bootstrap principle:

the population equation: $\mathbb{E}[f_t(F_0, F_1) | F_0] = 0$

↓ we solve via

the sample equation: $\mathbb{E}\{f_t(F_1, F_2) | F_1\} = 0$

For Bias correction:

$$f_t(F_0, F_1) = \theta(F_1) - \theta(F_0) + t_0$$

$$f_t(F_1, F_2) = \theta(F_2) - \theta(F_1) + t_1$$

$$\mathbb{E}\{f_t(F_1, F_2) | F_1\} = 0$$

$$\Rightarrow t_1 = \hat{t}_0 = \theta(F_1) - \mathbb{E}\{\theta(F_2) | F_1\}.$$

Then the bootstrap bias-reduced estimate is:

$$\theta(F_1) + \hat{t}_0 = 2\theta(F_1) - \mathbb{E}\{\theta(F_2) | F_1\}.$$

$$= \theta(F_1) - \mathbb{E}\{\theta(F_2) - \theta(F_1) | F_1\}$$

$$= \bar{x}^3 - \hat{b}_1$$

$$= \bar{x}^3 - \frac{3\bar{x}s^2}{n} - \frac{\hat{r}}{n^2}$$

$\hat{\theta}_1$

$$\underline{\underline{5)}}: b_2 = \mathbb{E}_{F_0}(\hat{\theta}_1 - \theta_0)$$

$$= \mathbb{E}(\hat{\theta}_1) - \theta_0$$

$$= \mathbb{E}\left(\bar{x}^3 - \frac{3\bar{x}\hat{\sigma}^2}{n} - \frac{\hat{r}}{n^2}\right) - \eta^3$$

$$\textcircled{1} \mathbb{E}(\bar{x}^3) = \eta^3 + \frac{3\eta\sigma^2}{n} + \frac{r}{n^2}$$

$$\textcircled{2} \mathbb{E}(\bar{x} \hat{\sigma}^2) = \mathbb{E}\left[\bar{x} \cdot \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2\right]$$

$$\textcircled{3} \mathbb{E}(\hat{r})$$

$$\begin{aligned} \text{For } \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2 &= \frac{1}{n} \left[\sum_{i=1}^n x_i^2 - 2x_i \bar{x} + \bar{x}^2 \right] \\ &= \frac{1}{n} \left[\sum_{i=1}^n x_i^2 - 2n\bar{x}^2 + n\bar{x}^2 \right] \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 - \frac{1}{n} n \bar{x}^2 \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 - \bar{x}^2 \end{aligned}$$

$$\begin{aligned} \Rightarrow \textcircled{2} &= \mathbb{E}[\bar{x} \hat{\sigma}^2] = \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n x_i^2 \bar{x} - \bar{x}^3\right] \\ &= \frac{1}{n} \left\{ \mathbb{E}\left[\bar{x} \cdot \sum_{i=1}^n x_i^2\right] \right\} - \mathbb{E}[\bar{x}^3] \end{aligned}$$

$$\text{For } E \left[\bar{x} \cdot \sum_{i=1}^n x_i^2 \right] = \frac{1}{n} \sum_{j=1}^n E \left\{ x_j \sum_{i=1}^n x_i^2 \right\}$$

$$\text{for } E \left(x_j \sum_{i=1}^n x_i^2 \right)$$

$$= E \left(x_j^3 + \sum_{i \neq j} x_j x_i^2 \right)$$

$$= E(x_j^3) + \sum_{i \neq j} E(x_j) E(x_i^2) = (6^2 + u^2)$$

using $n=1$ in the expression of $E\{\bar{X}^3\}$ gives $E\{X^3_j\}$

$$= u^3 + 3u6^2 + r + (n-1)u(u^2 + 6^2)$$

Not dependent on \tilde{j}

$$\Rightarrow E \left(\bar{x} \cdot \sum_{i=1}^n x_i^2 \right) = u^3 + 3u6^2 + r + nu^3 - u^3 + nu6^2 - u6^2$$

$$= 3n6^2 + r + nu^3 + nu6^2 - u6^2$$

$$\textcircled{2} = \frac{3n6^2}{n} + \frac{r}{n} + \underline{u^3} + u6^2 - \frac{u6^2}{n} - \underline{u^3} - \frac{3u6^2}{n} - \frac{r}{n^2}$$

$$= u6^2 + \frac{1}{n} (r - u6^2) - \frac{r}{n^2}$$

$$\textcircled{3} \mathbb{E}(\hat{\tau}) = \mathbb{E}\left(\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^3\right)$$

$$= \frac{1}{n} \sum_{i=1}^n \mathbb{E}(x_i - \bar{x})^3$$

$$= \mathbb{E}[(x_i - \bar{x})^3]$$

$$= \mathbb{E}[(x_i - u + u - \bar{x})^3]$$

$$= \mathbb{E}\left\{(x_i - u)^3 - 3(x_i - u)^2(u - \bar{x}) + 3(x_i - u)(\bar{x} - u)^2 - (\bar{x} - u)^3\right\}$$

$$= \mathbb{E}\{(x_i - u)^3\} - 3 \underbrace{\mathbb{E}\{(x_i - u)^2(u - \bar{x})\}}_C + 3 \underbrace{\mathbb{E}\{(x_i - u)(\bar{x} - u)^2\}}_D - \mathbb{E}\{(\bar{x} - u)^3\}$$

$$= n - 3C + 3D - \frac{n}{n^2}$$

$$C = \mathbb{E}\{(x_j - u)^2(\bar{x} - u)\}$$

$$= \frac{1}{n} \mathbb{E}\left[(x_j - u)^2 \left(\sum_{i=1}^n x_i - u\right)\right]$$

$$= \frac{1}{n} \mathbb{E}\left[(x_j - u)^3 + \sum_{i \neq j}^n \overset{\text{idpt}}{(x_j - u)^2 (x_i - u)}\right]$$

$$= \frac{1}{n} n + \sum_{i \neq j}^n \mathbb{E}(x_j - u)^2 \underbrace{\mathbb{E}(x_i - u)}_{=0}$$

$$= \frac{n}{n}$$

$$D = E \{ (x_j - u) (\bar{x} - u)^2 \}$$

$$= \frac{1}{n} \sum_{j=1}^n E [(x_j - u) (\bar{x} - u)^2]$$

$$= \frac{1}{n} \cdot E \left(\sum_{j=1}^n (x_j - u) (\bar{x} - u)^2 \right)$$

$$= \frac{1}{n} \cdot E \left[\left(\sum_{j=1}^n x_j - nu \right) (\bar{x} - u)^2 \right]$$

$$= \frac{1}{n} E [(n\bar{x} - nu) (\bar{x} - u)^2]$$

$$= E (\bar{x} - u)^3$$

$$= \frac{r}{n^2}$$

$$b_2 = u^3 + \frac{3u\sigma^2}{n} + \frac{r}{n^2} - \frac{3}{n} (u\sigma^2 + \frac{r}{n} - \frac{u\sigma^2}{n} - \frac{r}{n^2})$$

$$- \frac{1}{n^2} r (1 - \frac{3}{n} + \frac{2}{n^2}) - u^3$$

$$= \frac{3}{n^2} (u\sigma^2 - r) + \frac{6r}{n^3} - \frac{2r}{n^4}$$

6). Compare b_1 and b_2

• The Bias of the estimator $\hat{\theta}_1$ is $O(\frac{1}{n})$

• The Bias of the estimator $\hat{\theta}$ is $O(\frac{1}{n^2})$

\Rightarrow Bootstrap Bias correction reduces the order
magnitude of the Bias by the factor $\frac{1}{n}$