

$$S_n^{OR} = \frac{1}{\sqrt{n}} \sum_{i=1}^n \left[\left(\frac{x_i - \mu}{\sigma} \right)^3 - sk \right]$$

$$S_n = \frac{1}{\sqrt{n}} \sum_{i=1}^n \left[\left(\frac{x_i - \hat{\mu}}{\hat{\sigma}} \right)^3 - sk \right] \quad \hat{\mu} = \bar{x} \quad \hat{\sigma}^2 = \bar{x}^2 - (\bar{x})^2$$

1)

$$\begin{aligned} S_n^{Ho} &= \frac{1}{\sqrt{n} \hat{\sigma}^3} \sum_{i=1}^n (x_i - \hat{\mu})^3 = \frac{1}{\hat{\sigma}^3 \sqrt{n}} \sum_{i=1}^n (x_i^3 - 3x_i^2 \hat{\mu} + 3x_i \hat{\mu}^2 - \hat{\mu}^3) \\ &= \frac{1}{\sqrt{n}} \cdot \frac{[\bar{x}^3 - 3\bar{x}\bar{x}^2 + 3\bar{x}(\bar{x})^2 - (\bar{x})^3]}{(\bar{x}^2 - (\bar{x})^2)^{\frac{3}{2}}} \end{aligned}$$

$$= \sqrt{n} g(\bar{x}^3, \bar{x}^2, \bar{x})$$

$$g(a, b, c) = \frac{a - 3bc + 2c^3}{(b - c)^{\frac{3}{2}}}$$

we have $\sqrt{n} \left[\begin{pmatrix} \bar{x}^3 \\ \bar{x}^2 \\ \bar{x} \end{pmatrix} - \begin{pmatrix} E(x^3) \\ E(x^2) \\ E(x) \end{pmatrix} \right] \xrightarrow{d} N(0, \Sigma) \quad \text{CLT.}$

Delta method:

$$\sqrt{n} [g(\bar{x}^3, \bar{x}^2, \bar{x}) - g(E(x^3), E(x^2), E(x))] \xrightarrow{d} \nabla g(\cdot)^T N(0, \Sigma)$$

S_n is asymptotically normal.

2)

$$S_n = \frac{1}{\sqrt{n}} \sum_{i=1}^n \left[\left(\frac{X_i - \hat{\mu}}{\hat{\sigma}} \right)^3 - \overline{\left(\frac{X - \hat{\mu}}{\hat{\sigma}} \right)^3} \right]$$

$$\mu = E_{P_X}[X]$$

$$E_{\hat{P}_X}[X] = \hat{\mu} = \bar{X}$$

{ Replace P_X with \hat{P}_X ①
 use principle of substitution & similarity. ②

$$S_n^* = \frac{1}{\sqrt{n}} \sum_{i=1}^n \left[\underbrace{\left(\frac{X_i^* - \hat{\mu}^*}{\hat{\sigma}^*} \right)^3}_{\textcircled{2}} - \underbrace{\overline{\left(\frac{X - \hat{\mu}}{\hat{\sigma}} \right)^3}}_{\textcircled{1}} \right]$$

Then, Compute the quantiles of S_n^* $\hat{q}_{\alpha/2}$, $\hat{q}_{1-\alpha/2}$

Reject H_0 iff $S_n^* \notin [\hat{q}_{\alpha/2}, \hat{q}_{1-\alpha/2}]$

Bootstrap provides no refinements, but it avoids the computation of variance ψ , which is complex.

$$3) \quad S_n^* = \sqrt{n} \left[g(\overline{(X^*)^3}, \overline{(X^*)^2}, \overline{X^*}) - g(\overline{X^3}, \overline{X^2}, \overline{X}) \right]$$

By Pairwise Bootstrap CLT

$$\sqrt{n} \left[\begin{pmatrix} \overline{(X^*)^3} \\ \overline{(X^*)^2} \\ \overline{X^*} \end{pmatrix} - \begin{pmatrix} \overline{X^3} \\ \overline{X^2} \\ \overline{X} \end{pmatrix} \right] \xrightarrow{d} N(0, \Sigma)$$

By Bootstrap delta method

$$J_n [g(\bar{X}^*), (\bar{X}^*)^2, \bar{X}^*) - g(\bar{X}^3, \bar{X}^2, \bar{X})] \xrightarrow{d} \nabla g(\cdot)^T N(0, \Sigma)$$

Recall that $S_n \xrightarrow{d} \nabla g(\cdot)^T N(0, \Sigma)$

$$\text{So, } \|\hat{F}_{S_n^*} - F_{S_n}\|_\infty \leq \underbrace{\|\hat{F}_{S_n^*} - \Phi_\Psi\|_\infty}_{o(1)} + \underbrace{\|\Phi_\Psi - F_{S_n}\|_\infty}_{o(1)}$$

Then S_n^* estimates consistently the distribution of S_n

Ex 2.

$$S_{n,1} = J_n [(\bar{X})^2 - (E(X))^2] = J_n [g(\bar{X}) - g(E(X))]$$
$$\xrightarrow{d} g'(E(X)) \cdot N(0, V(X))$$

CLT + Δ method.

$$S_{n,1}^* = J_n [g(\bar{X}^*) - g(\bar{X})]$$

$$\xrightarrow{d} g'(E(X)) N(0, V(X)) \text{ in probability}$$

$S_{n,1}^*$ consistently the distribution of $S_{n,1}$

$$3). \quad S_{n,3}^* = \left[\sqrt{n} (\bar{X}^* - \bar{X}) \right]^2$$

$$\text{By Bootstrap CLT, } \sqrt{n} (\bar{X}^* - \bar{X}) \rightarrow N(0, V(X))$$

$$S_{n,3}^* \xrightarrow{d} \chi^2$$

Ex3. with proposition 4.2.1,

$$\text{we should have: } E(V) = 1,$$

$$\text{Var}(V) = 1$$