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**Abstract** We derive the asymptotic normality of the quantile regression estimators for a simple random sample from a finite population.

**1. Quantile regression** Quantile regression is a method for estimating models of conditional quantile functions introduced by Koenker and Bassett (1978). Koenker and Bassett (1978) also derive the asymptotic normality of these estimators under some regularity conditions. We extend their result to the finite population case via the method of Kato et al. (2009).

Notations:

$\{\mathcal{P}_k, k = 1, 2, \dots\}$ : a sequence of finite populations of size  $N_k$

$(y_j, 1, x_{1j}, \dots, x_{pj}) = (y_j, \mathbf{x}_j), j = 1, \dots, N_k$ : characteristics of  $\mathcal{P}_k$

$(y_i, \mathbf{x}_i), i = 1, \dots, n_k$ : simple random sample without replacement from  $\mathcal{P}_k$

Model:

$$\begin{aligned} y_j &= \beta_0 + \beta_1 x_{1j} + \dots + \beta_p x_{pj} + u_j \\ &= \boldsymbol{\beta}^T \mathbf{x}_j + u_j, \quad j = 1, \dots, N_k, \end{aligned}$$

$\boldsymbol{\beta} = (\beta_0, \dots, \beta_p)^T$ : an unknown  $(p+1) \times 1$  parameter vector

$u_j$ : the unexplained residual terms

$F_k(\cdot) = \sum_{j=1}^{N_k} I(u_j \leq \cdot) / N_k$  with  $F_k(0) = \tau$

: the (population) distribution function of  $u_i$

Quantile regression estimator of  $\boldsymbol{\beta}$ :

$$(\hat{\beta}_0, \hat{\beta}_1, \dots, \hat{\beta}_p) = \operatorname{argmin}_{\boldsymbol{\beta}} \sum_{i=1}^{n_k} \rho_\tau(y_i - \boldsymbol{\beta}^T \mathbf{x}_i),$$

over all  $\boldsymbol{\phi} = (\phi_0, \dots, \phi_p)^T$ , where

$$\rho_\tau(u) = u(\tau - I(u \leq 0))$$

I: indicator function

$$I(u \leq 0) = \begin{cases} 1 & u \leq 0 \\ 0 & u > 0 \end{cases}$$

## 2. Asymptotic Normality

### 2.1 Assumptions

$$(A0) \lim_{k \rightarrow \infty} \min n_k, N_k - n_k = \infty$$

(A1) There is a sequence of functions  $\{f_k\}$  such that

$$\lim_{k \rightarrow \infty} \left[ \frac{F_{N_k}(0 + \delta_k) - F_{N_k}(0)}{\delta_k} - f_k(0) \right] = 0,$$

for any sequence  $\{\delta_k\}$  of order  $\sim O(n_k^{-1/2})$  and

$$0 < \inf_k f_k(0) \leq \sup_k f_k(0) < \infty.$$

(A2) (i) There exists a positive definite matrix  $C_k$  such that

$$\lim_{k \rightarrow \infty} \left\| \frac{1}{N_k} \frac{N_k - n_k}{N_k - 1} \mathbf{X}_{n_k}^T \mathbf{X}_{n_k} - C_k \right\|_{\max} = 0,$$

where  $\|A\|_{\max} = \max |a_{ij}|$ .

(ii)  $\lim_{k \rightarrow \infty} \left( \frac{\sum_{S_{k\tau}^1} (x_{lj} - \bar{x}_l)^2}{\sum_{j=1}^{N_k} (x_{1j} - \bar{x}_1)^2} \right) = 0$  where  $S_{k\tau}^l$ : the set of  $|x_{lj} - \bar{x}_l| > \tau \sqrt{\frac{n_k(N_k - n_k)}{N_k} \sum_{j=1}^{N_k} (x_{lj} - \bar{x}_l)^2}$ ,  $l = 1, \dots, p$ .

### 2.2 Main Result

**Theorem 1.** Under the assumptions (A0), (A1) and (A2), we have

$$\frac{\sqrt{n_k}}{f_k(0)} C_k^{-1/2} (\hat{\boldsymbol{\beta}}_k - \boldsymbol{\beta}) \xrightarrow{d} N(\mathbf{0}, \tau(1 - \tau) \mathbf{I}_{p+1}),$$

as  $k \rightarrow \infty$ .

### References

- Koenker, R. and G. Bassett (1978). *Econometrica* **46**, 33-50.
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