

Large Sample Theory

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Exercises, Section 13, Asymptotic Distribution of Sample Quantiles.

1. (a) Find the asymptotic joint distribution of $(X_{(np)}, X_{(n(1-p))})$ when sampling from a Cauchy distribution $\mathcal{C}(\mu, \sigma)$. You may assume $0 < p < 1/2$. See Example 13.2 and Exercise 13.3.

(b) Find the asymptotic distribution of $\hat{\mu}_n = (1/2)(X_{(np)} + X_{(n(1-p))})$.

(c) What value of p minimizes the asymptotic variance of $\hat{\mu}_n$? Compare this estimate to the sample median as an estimate of μ .

2. Let $0 < p_1 < \dots < p_k < 1$, and let $X_{(\lceil np_i \rceil)}$ be the corresponding sample quantiles for a sample of size n from a distribution with location parameter θ having distribution function $F(x - \theta)$ and density $f(x - \theta)$. Let u_i denote the p_i th quantile of F (i.e. $F(u_i) = p_i$).

(a) Let $Z_i = X_{(\lceil np_i \rceil)} - u_i$. Let \mathbf{Z} represent the vector $(Z_1, \dots, Z_k)^T$ and $\mathbf{1}$ represent the k -vector of all 1's. Show that $\sqrt{n}(\mathbf{Z} - \theta\mathbf{1}) \xrightarrow{\mathcal{L}} \mathcal{N}(\mathbf{0}, \mathbb{F})$, where \mathbb{F} is the symmetric matrix with components $\sigma_{ij} = \frac{p_i(1-p_j)}{f(u_i)f(u_j)}$ for $i \leq j$.

(b) Find the asymptotic best linear unbiased estimate of θ based on \mathbf{Z} . That is, for $\hat{\theta} = \mathbf{a}^T \mathbf{Z}$, find \mathbf{a} to minimize $\mathbf{a}^T \mathbb{F} \mathbf{a}$ subject to $\mathbf{1}^T \mathbf{a} = 1$ (in terms of \mathbb{F}^{-1}).

(c) In view of (b), it is comforting to know that the inverse of \mathbb{F} has a simple form. It is a tridiagonal matrix. Find it.

(d) Find $\hat{\theta}$ of (b) explicitly, for the uniform distribution, $F(x) = x$ for $0 \leq x \leq 1$.

3. Suppose we have a sample, X_1, \dots, X_n , from the family of distributions on the real line with density $f(x|\theta, \alpha) = c(\alpha)e^{-|x-\theta|^\alpha}$, $\alpha > 0$. We may use the sample mean, \bar{X}_n , or the sample median, m_n to estimate the location parameter θ .

(a) Find the constant $c(\alpha)$.

(b) What is the asymptotic distribution of \bar{X}_n ?

(c) What is the asymptotic distribution of m_n ?

(d) For what values of α is the asymptotic variance of m_n smaller than the asymptotic variance of \bar{X}_n ?

4. Let X_1, \dots, X_n be a sample from $\mathcal{N}(\theta, \sigma^2)$ with σ^2 known. It is desired to estimate the p th quantile, $x_p = \theta + \sigma z_p$, where z_p is the p th quantile of the standard normal distribution. The maximum likelihood estimate of x_p is clearly $\hat{x}_p = \bar{X}_n + \sigma z_p$. What is the asymptotic distribution of $\sqrt{n}(\hat{x}_p - x_p)$? What is the asymptotic efficiency of $X_{(\lceil np \rceil)}$ relative to \hat{x}_p ?

5. Let X_1, \dots, X_n be a sample from the Pareto distribution with density $f(x|\theta) = \theta/(x + \theta)^2$ for $x > 0$, and distribution function $F(x|\theta) = x/(x + \theta)$ for $x > 0$. Let $x_p(\theta)$ denote the p th quantile of the distribution and let $X_{(\lceil np \rceil)}$ denote the sample p th quantile.

(a) What is the asymptotic distribution of $X_{(\lceil np \rceil)}$ as $n \rightarrow \infty$?

(b) Find a constant $c(p)$ such that $\hat{\theta}_n = c(p)X_{(\lceil np \rceil)}$ is a consistent asymptotically unbiased estimate of θ . For what value of p is the asymptotic variance of $\hat{\theta}_n$ a minimum?