

**A Bahadur - Type Representation for Empirical Quantiles
of a Large Class of Stationary, Possibly Infinite-Variance,
Linear Processes**

By

C.H. Hesse
Department of Statistics
University of California, Berkeley

Technical Report No. 132
December 1987
(revised June 1989)

Department of Statistics
University of California
Berkeley, California 94720

**A Bahadur - Type Representation for Empirical Quantiles
of a Large Class of Stationary, Possibly Infinite-Variance,
Linear Processes**

By

C.H. Hesse
Department of Statistics
University of California, Berkeley

Abstract

Bahadur (1966) has obtained an asymptotic almost sure representation for empirical quantiles of independent and identically distributed random variables. In this paper we present an analogous result for a large class of stationary linear processes.

Acknowledgement: The author would like to thank Professor E.J. Hannan for many stimulating discussions.

Key Words: Stationary linear processes, Quantiles, Bahadur representation, Almost sure convergence.

AMS subject classifications: Primary 62G30;
Secondary 60F15, 60G10

0. Introduction

Bahadur (1966) has initiated the asymptotic representation theory of sample quantiles via the empirical distribution function. In particular, he demonstrated that under certain fairly mild regularity conditions on the distribution F and the density f of the iid sequence $X(1), X(2), \dots$ the following is true with probability one:

$$X_{p,n} = \xi_p + \frac{p - F_n(\xi_p)}{f(\xi_p)} + R_n$$

and

$$R_n = O(n^{-3/4}(\log n)^{1/2}(\log \log n)^{1/4}).$$

Here, for $0 < p < 1$, ξ_p is the unique p -quantile of F , i.e. $F(\xi_p) = p$, $X_{p,n}$ is the p -th sample quantile based on $X(1), X(2), \dots, X(n)$, and F_n is the empirical distribution function based on the same sample.

Bahadur's theorem and proof give great insight into the relation between empirical quantiles and the empirical distribution function. It has triggered a number of refined studies in the iid case and subsequent extension to non-independent sequences: Analysis by Eicker (1966) has revealed that the remainder R_n is $o_p(n^{-3/4}g(n))$ if and only if $g(n) \rightarrow \infty$ as $n \rightarrow \infty$, and Kiefer (1967) gave the following definite answer,

$$\text{w.p.1} \quad \limsup_{n \rightarrow \infty} \pm \frac{n^{3/4} R_n}{(\log \log n)^{3/4}} = \frac{2^{5/4} (p(1-p))^{1/4}}{3^{3/4}}$$

for either choice of sign.

Other references in the iid case include Duttweiler (1973) and Ghosh (1971), who obtained a simpler proof of Bahadur's representation but for a weaker result.

There are some extensions to sequences of random variables with certain dependency

structures, e.g. m -dependence, ϕ -mixingness, strong mixingness, compare Sen (1968, 1972) and Babu and Singh (1978). In this paper we obtain an analogous strong representation for a very broad class of stationary linear processes with parameters decreasing at a polynomial rate. In particular, the following sequences are considered:

$$(0.1) \quad X(n) = \sum_{i=0}^{\infty} \delta(i) \varepsilon(n-i)$$

where $\varepsilon(n)$ are iid innovations with $E(|\varepsilon(n)|^\alpha) < \infty$ for some $\alpha > 0$ and $|\delta(i)| \leq c \cdot i^{-q}$ for some $c, q > 0$ and $i \geq 1$.

The class of linear processes in (0.1) is very broad. It includes both finite and infinite variance linear processes and also incorporates processes based on both continuous and certain (due to restrictions on the stationary distribution function that will be imposed later) discrete innovation series $\varepsilon(n)$. It covers m -dependent sequences, all autoregressive - moving average processes and certain sequences which are neither ϕ -mixing nor strong mixing. Examples of sequences within the class (0.1) which are not strong mixing are easily obtained: The first order autoregressive process

$$(0.2) \quad X(n) - \frac{1}{2} X(n-1) = \varepsilon(n)$$

is strongly mixing iff $\varepsilon(n)$ has a distribution with absolutely integrable characteristic function, such as the normal distribution; see Chanda (1974). If for example the $\varepsilon(n)$ are iid symmetric Bernoulli, $X(n)$ is not strongly mixing, compare also Andrews (1983). However, (0.2) is in the class (0.1), (even has an absolutely continuous stationary distribution function), which is easily demonstrated by obtaining the stationary solution of the difference equation, namely

$$X(n) = \sum_{i=0}^{\infty} 2^{-i} \varepsilon(n-i).$$

1. STATEMENT OF RESULT

Theorem 1 in this section is the main contribution of the paper. It gives a Bahadur-type result for empirical quantiles of the broad class of stationary processes introduced in (0.1) with parameters $\delta(i)$ decreasing to zero in absolute value at a polynomial rate. For $0 < p < 1$, what is here and below meant by p -th sample quantile $X_{p,n}$ of a sequence of random variables $X(1), \dots, X(n)$ is the $[n \cdot p]$ -th order statistic, where $[x]$ denotes the smallest integer larger or equal to x .

As before F_n, F denote the empirical distribution and the stationary distribution of X , respectively, ξ_p is such that $F(\xi_p) = p$, and c denotes a generic positive constant, not always the same one. Other notation will be introduced as needed.

Theorem 1: Let

$$(1.1) \quad X(n) = \sum_{i=0}^{\infty} \delta(i) \varepsilon(n-i)$$

where the innovations $\varepsilon(n)$ are iid with $E(|\varepsilon(n)|^\alpha) < \infty$ for some $\alpha > 0$. Assume also that for $i \geq 1$, $|\delta(i)| \leq c \cdot i^{-q}$ with $q > 1 + 2/\alpha$, and that the density f of the stationary distribution of X is bounded away from 0 and ∞ in a neighborhood B_p of ξ_p . Then

$$X_{p,n} = \xi_p + \frac{p - F_n(\xi_p)}{f(\xi_p)} + R_n \quad \text{a.s.}$$

where $R_n = O(n^{-3/4+\gamma})$ for all $\gamma > \frac{\alpha^2(8q-5) + 2\alpha(10q-9) - 13}{4(2\alpha q - \alpha - 1)^2}$.

Remark 1: The lower bound for γ in Theorem 1 is decreasing both for increasing α and for increasing q .

2. PROOF OF MAIN RESULT

The proof is based on extensions of Lemmas 1-3 in Bahadur (1966) to the present context. Two of these extensions are straightforward while one (our Theorem 2) is somewhat involved.

Theorem 2: Under the conditions of Theorem 1 let, for given n , $0 < \beta < 1$ and $i = 1, \dots, n$

$$X_{\beta}(i) = \sum_{j=0}^{\lceil n^{\beta} \rceil - 1} \delta(j) \varepsilon(i-j).$$

Then it holds true for the empirical p -th quantile $(X_{\beta})_{p,n}$ of $X_{\beta}(i)$ that for all n sufficiently large

$$(X_{\beta})_{p,n} \in I_n = (\xi_p - a_n, \xi_p + a_n) \text{ a.s.}$$

$$\text{with } a_n = O\left(\max\left\{n^{1/2(\beta-1)}(\log n)^{1/2}, n^{+\frac{\alpha}{\alpha+1}(\beta(1+1/\alpha-q)+1/\alpha)}\right\}\right).$$

Remark 2: The parameter β determines the order of truncation of the infinite linear combination of innovations in (1.1). To obtain the strongest result in Theorem 1, an optimal choice for β will have to be made later.

Proof of Theorem 2: For latter use we first evaluate the difference between $X(i)$ and $X_{\beta}(i)$. Clearly,

$$\begin{aligned} |X(i) - X_{\beta}(i)| &\leq c \cdot \sum_{j=\lceil n^{\beta} \rceil}^{\infty} j^{-q} |\varepsilon(i-j)| \\ &\leq c \cdot \varepsilon^* n^{1/\alpha} \sum_{j=\lceil n^{\beta} \rceil}^{\infty} j^{-q+1/\alpha} \end{aligned}$$

where $\varepsilon^* = \sup\left\{|\varepsilon(0)|, \sup_{|k| \geq 1} \frac{|\varepsilon(k)|}{|k|^{1/\alpha}}\right\}$ is almost sure finite due to the conditions on $\varepsilon(k)$

and moreover $P(\varepsilon^* \geq \xi) \leq c \cdot \xi^{-\alpha}$ for all $\xi > 0$, as is easily proved. Hence, uniformly in i (up to n),

$$(2.1) \quad |X(i) - X_\beta(i)| \leq c \cdot \varepsilon^* n^{\beta(1 + 1/\alpha - q) + 1/\alpha}.$$

We write $\eta = -\beta(1 + 1/\alpha - q) - 1/\alpha$. Equation (2.1) implies that the difference between the p -th sample quantiles based on $X_\beta(i)$, $X(i)$, $i = 1, \dots, n$, respectively, is

$$(2.2) \quad |(X_\beta)_{p,n} - X_{p,n}| = O(n^{-\eta})$$

Here and below order relations are to be interpreted to hold almost surely.

Then, using Lemma 1 from the Appendix, we get

$$(2.3) \quad \sup_{y \in B_p} |F(y) - F_\beta(y)| = O(n^{-\frac{\alpha}{\alpha+1}\eta})$$

and

$$(2.3a) \quad \sup_{y \in B_p} |F(y) - F_\beta(y-)| = O(n^{-\frac{\alpha}{\alpha+1}\eta})$$

where F_β is the distribution of X_β and $F_\beta(y-)$ denotes the limit from the left, i.e.

$$F_\beta(y-) = \lim_{y_0 \uparrow y} F_\beta(y_0).$$

We now exploit the independence of the truncated series after $\log \lceil n^\beta \rceil$ by defining

$$S_{n,k}^\beta = \{X_\beta(k), X_\beta(k + \lceil n^\beta \rceil), \dots, X_\beta(k + (n_k - 1) \cdot \lceil n^\beta \rceil)\}, \quad k = 1, 2, \dots, \lceil n^\beta \rceil.$$

where n_k is either $\lceil n^{1-\beta} \rceil$ or $\lceil n^{1-\beta} \rceil - 1$, its dependence on k being of no concern.

If for given n , X_{p,n_k}^k denotes the p -th sample quantile of the k -th set $S_{n,k}^\beta$ containing n_k random variables, then by Lemma 2

$$(2.4) \quad \min_{1 \leq k \leq \lceil n^\beta \rceil} X_{p,n_k}^k \leq (X_\beta)_{p,n} \leq \max_{1 \leq k \leq \lceil n^\beta \rceil} X_{p,n_k}^k.$$

At this point it is necessary to point out one of the defects of the distribution F_β : its possible discontinuity. We remedy this by introducing the slightly perturbed but continuous random variables

$$G_{\beta}(X_{\beta}(i)) = U(i) \cdot F_{\beta}(X_{\beta}(i)-) + (1 - U(i)) F_{\beta}(X_{\beta}(i))$$

where $U(i)$ has the uniform distribution over $(0,1)$ and is independent of $X_{\beta}(i)$. If we also define $G_{\beta}(y)$ as

$$G_{\beta}(y) = U(i) \cdot F_{\beta}(y-) + (1 - U(i)) F_{\beta}(y)$$

then, using (2.3) and (2.3a), we get

$$\sup_{y \in B_p} |F_{\beta}(y) - F_{\beta}(y-)| \leq \sup_{y \in B_p} |F_{\beta}(y) - F(y)| + \sup_{y \in B_p} |F(y) - F_{\beta}(y-)| = O(n^{-\frac{\alpha}{\alpha+1}\eta})$$

and hence

$$(2.5) \quad \sup_{y \in B_p} |F_{\beta}(y) - G_{\beta}(y)| = O(n^{-\frac{\alpha}{\alpha+1}\eta})$$

Since, in particular, for all n sufficiently large $(X_{\beta})_{p,n} \in B_p$ a.s., by Lemma 5, Equation

(2.5) therefore implies that

$$|F_{\beta}((X_{\beta})_{p,n}) - G_{\beta}((X_{\beta})_{p,n})| = O(n^{-\frac{\alpha}{\alpha+1}\eta})$$

from which we deduce

$$(2.6) \quad (F_{\beta}(X_{\beta}))_{p,n} = (G_{\beta}(X_{\beta}))_{p,n} + O(n^{-\frac{\alpha}{\alpha+1}\eta})$$

by monotonicity of F_{β} and G_{β} . In Equation (2.6) $(F_{\beta}(X_{\beta}))_{p,n}$ is the $[n \cdot p]$ -th order statistic of $F_{\beta}(X_{\beta}(i))$, $i = 1, \dots, n$ and $(G_{\beta}(X_{\beta}))_{p,n}$ is defined similarly. Equation (2.6) demonstrates that the effect introduced by the perturbation with respect to the corresponding p -th quantiles may be ignored. Keeping in mind (2.6) we are in the sequel concerned with $G_{\beta}(X_{\beta}(i))$, $i = 1, \dots, n$ only.

We will first determined how close $(-\log(G_{\beta}(X_{\beta})))_{p,n}$, the p -th quantile of $-\log G_{\beta}(X_{\beta}(i))$, $i = 1, \dots, n$ is to $\log p^{-1}$. Since $X_{\beta}(i)$ has distribution F_{β} , $-\log G_{\beta}(X_{\beta}(i))$ has an exponential distribution with mean 1. On each set $S_{n,k}^{\beta}$, we may therefore apply the Renyi representation (see e.g., Shorack and Wellner (1986) p.723)

to these transformed random variables. In particular, for the $\lceil n_k \cdot p \rceil$ -th order statistic $(-\log G_\beta(X_\beta))_{p,n_k}^k$ of the transformed random variables in $S_{n,k}^\beta$ we obtain

$$(2.7) \quad (-\log G_\beta(X_\beta))_{p,n_k}^k = \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v}}{v}$$

where for each n,k the $E_{n,k,v}$ for different v are independent random variables with exponential distribution (with mean 1). This representation is used to establish that

$$(2.8) \quad \limsup_{n \rightarrow \infty} \max_{1 \leq k \leq \lceil n^\beta \rceil} \left(\frac{n^{1-\beta}}{\log n} \right)^{1/2} |(-\log G_\beta(X_\beta))_{p,n_k}^k - \log p^{-1}| < \infty.$$

In view of (2.7) and the rectangular rule of quadrature it suffices to show, in place of (2.8), that

$$(2.9) \quad \max_{1 \leq k \leq \lceil n^\beta \rceil} \left| \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v} - 1}{v} \right| = O(n^{1/2(\beta-1)} (\log n)^{1/2}).$$

Lemma 3 in the Appendix proves the statement in (2.9).

Equations (2.9), (2.7) together with (2.4) imply that

$$(-\log G_\beta(X_\beta))_{p,n} = \log p^{-1} + O(n^{1/2(\beta-1)} (\log n)^{1/2})$$

and hence

$$(2.10) \quad (G_\beta(X_\beta))_{p,n} = p + O(n^{1/2(\beta-1)} (\log n)^{1/2}).$$

Since

$$|G_\beta((X_\beta)_{p,n}) - F(X_\beta)_{p,n}| = |U(i) [F_\beta(X_\beta)_{p,n} - F(X_\beta)_{p,n}] + (1 - U(i)) [F_\beta(X_\beta)_{p,n} - F(X_\beta)_{p,n}]|$$

and because of (2.3), (2.3a) and Lemma 5 we get

$$(2.11) \quad (G_\beta(X_\beta))_{p,n} = (F(X_\beta))_{p,n} + O\left(n^{-\frac{\alpha}{\alpha+1}\eta}\right)$$

for n large enough. Then, combining (2.11) and (2.10) establishes

$$(F(X_\beta))_{p,n} = p + O\left(\max\left\{n^{1/2(\beta-1)} (\log n)^{1/2}, n^{-\frac{\alpha}{\alpha+1}\eta}\right\}\right).$$

Since, over the neighborhood B_p , the derivative of F is bounded away from 0 which implies that F^{-1} has bounded derivatives, we may transform from $(F(X_\beta))_{p,n}$ to $(X_\beta)_{p,n}$

and obtain

$(X_\beta)_{p,n} \in I_n$ with probability one for all sufficiently large n .

This completes the proof of Theorem 2.

Remark 3: Because of (2.2) and since $\alpha/(\alpha+1) < 1$ for $\alpha > 0$ the statement of Theorem 2 holds with $(X_\beta)_{p,n}$ replaced by $X_{p,n}$.

Proposition 1: If $F_n^{\beta^*}$ is the empirical distribution function of $X_{\beta^*}(i)$, $i = 1, \dots, n$ where β^* is not necessarily equal to β above, and

$$V_n(y) = |F_n^{\beta^*}(y) - F_n^{\beta^*}(\xi_p) - (F_{\beta^*}(y) - F_{\beta^*}(\xi_p))|.$$

Then

$$\limsup_{n \rightarrow \infty} (\gamma(n))^{-1} \sup_{y \in I_n} V_n(y) < \infty \text{ a.s.}$$

for any $\gamma(n)$ with

$$\gamma(n) = O\left(\max\left\{n^{-\frac{1}{2} \frac{\alpha}{\alpha+1} \eta - \frac{1}{2}(1-\beta^*) + \varepsilon}, n^{-\frac{3}{4} + \frac{\beta}{4} + \frac{\beta^*}{2} + \varepsilon}\right\}\right), \varepsilon > 0$$

and

$$I_n = (\xi_p - a_n, \xi_p + a_n)$$

where

$$(2.12) \quad a_n = O\left(\max\left\{n^{1/2(\beta-1)}(\log n)^{1/2}, n^{-\frac{\alpha}{\alpha+1} \eta}\right\}\right)$$

Proof of Proposition 1: Without loss of generality we assume that F_{β^*} is continuous.

If not we use the method introduced after (2.4) and consider G_{β^*} instead of F_{β^*} . Write $F_{n_k}^{\beta^*}$ for the empirical distribution function of $X_{\beta^*}(i)$ based on the subset $S_{n,k}^{\beta^*}$, and $V_{n_k}^k(y)$ accordingly, then

$$\sup_{y \in I_n} V_n(y) \leq \sup_{y \in I_n} \max_{1 \leq k \leq \lceil n^{\beta^*} \rceil} V_{n_k}^k(y).$$

Let

$$(2.13) \quad b_n = \lceil cn^s \rceil$$

be an integer sequence with an optimal exponent s to be selected later. Also, write

$$w_{n,v} = \xi_p + a_n \cdot b_n^{-1} \cdot v, \quad I_n^v \quad \text{for the interval } [w_{n,v}, w_{n,v+1}] \quad \text{and}$$

$$u_{n,v} = F_{\beta^*}(w_{n,v+1}) - F_{\beta^*}(w_{n,v}) \quad \text{for all } n \text{ and integers } v \text{ with } -b_n \leq v \leq b_n - 1. \quad \text{Then for}$$

$$\text{all } y \in I_n^v$$

$$V_{n_k}^k(y) \leq V_{n_k}^k(w_{n,v+1}) + u_{n,v}$$

$$V_{n_k}^k(y) \geq V_{n_k}^k(w_{n,v+1}) - u_{n,v}$$

and hence

$$(2.14) \quad \sup_{y \in I_n} V_n(y) \leq \max_{1 \leq k \leq \lceil n^{\beta} \rceil} \max_{-b_n \leq v \leq b_n} V_{n_k}^k(w_{n,v}) + \max_{-b_n \leq v \leq b_n - 1} u_{n,v} \\ \leq T_1(n) + T_2(n), \quad \text{say.}$$

Since $w_{n,v+1} - w_{n,v} \leq a_n b_n^{-1}$ for each v , and since F_{β^*} (or better G_{β^*}) is sufficiently well-behaved in a fixed neighborhood of ξ_p , it follows that $T_2(n) = O(a_n b_n^{-1})$.

As far as $T_1(n)$ is concerned, it suffices, in view of the Borell-Cantelli lemma and Bonferroni's inequality, to show that

$$(2.15) \quad \sum_{n=N_0}^{\infty} \sum_{k=1}^{\lceil n^{\beta} \rceil} \sum_v P(V_{n_k}^k(w_{n,v}) \geq \gamma(n)) < \infty$$

for N_0 sufficiently large so that here and below degeneracies are avoided. To demonstrate this, we exploit the fact that the distribution of $V_{n_k}^k(w_{n,v})$ is the same as that of $n_k^{-1} |B(n_k, \delta(n, v)) - n_k \cdot \delta(n, v)|$, where $B(n_k, \delta(n, v))$ is a binomial random variable with parameters n_k and $\delta(n, v) = |F_{\beta^*}(w_{n,v}) - F_{\beta^*}(\xi_p)|$. Using Bernstein's inequality

$$(2.16) \quad P(|B(n_k, \delta(n, v)) - n_k \cdot \delta(n, v)| \geq \gamma(n)) \leq 2 \exp(-h)$$

$$\text{with } h = \gamma(n)^2 / \{2[n_k \delta(n, v)(1 - \delta(n, v)) + (\gamma(n)/3) \max\{\delta(n, v), 1 - \delta(n, v)\}\} \}.$$

N_0 in (2.15) has to be chosen so large that $F_{\beta^*}(\xi_p + a_n) - F_{\beta^*}(\xi_p) < c_1 a_n$ and

$F_{\beta^*}(\xi_p) - F_{\beta^*}(\xi_p - a_n) < c_1 a_n$ for all $n > N_0$ and some constant c_1 . Using $h = h(n_k, \delta(n, v), \gamma(n)) \geq \gamma(n)^2 / 2 [n_k \cdot \delta(n, v) + \gamma(n)]$ and since $|v| \leq b_n$ implies $\delta(n, v) \leq c_1 \cdot a_n$ for $n > N_0$ it follows

$$P(V_{n_k}^k(w_{n,v}) \geq \gamma(n)) \leq 2 \exp(-h_1)$$

where $h_1 = h_1(n_k, \gamma(n)) = n_k^2 \gamma(n)^2 / 2 [c_1 n_k \cdot a_n + n_k \cdot \gamma(n)]$ which depends on k only through n_k and is independent of v . Hence

$$\sum_{k=1}^{\lceil n^{\beta^*} \rceil} \sum_v P(V_{n_k}^k(w_{n,v}) \geq \gamma(n)) \leq 4 b_n \cdot \lceil n^{\beta^*} \rceil \cdot \exp(-h_1(\lceil n^{1-\beta^*} \rceil - 1, \gamma(n)))$$

where $\gamma(n)$ and the exponent of n in $b_n = \lceil cn^s \rceil$ have to be chosen so that for all $n \geq N_0$ the expression $\beta^* + s - h_1(\lceil n^{1-\beta^*} \rceil - 1, \gamma(n)) / \log n$ is less than -1 . Hence, given β^* , we choose s so that the exponent of $a_n b_n^{-1} \cdot n^{1-\beta^*}$ is larger than s and since

$$O(a_n b_n^{-1}) = O\left(\max\left\{n^{-\frac{\alpha}{\alpha+1}\eta-s}, n^{1/2(\beta-1)-s}(\log n)^{1/2}\right\}\right)$$

this requires

$$-\frac{\alpha}{\alpha+1}\eta - s + 1 - \beta^* > s \quad \text{or} \quad 1 - \beta^* - \frac{1}{2}(1 - \beta) - s > s$$

which leads to

$$s < \frac{1}{2}\left(1 - \frac{\alpha}{\alpha+1}\eta - \beta^*\right) \quad \text{or} \quad s < \frac{1}{4} + \frac{\beta}{4} - \frac{\beta^*}{2}$$

and $\gamma(n) = O\left(\max\left\{n^{-\frac{1}{2}\frac{\alpha}{\alpha+1}\eta - \frac{1}{2}(1-\beta^*)+\varepsilon}, n^{-\frac{3}{4} + \frac{\beta}{4} + \frac{\beta^*}{2} + \varepsilon}\right\}\right)$, $\varepsilon > 0$. This completes the proof of Proposition 1.

Remark 4: The rate $\gamma(n)$ essentially determines the rate of convergence in Theorem 1.

An optimal choice for β in Theorem 2 is $\beta_0 = \frac{\alpha+3}{2\alpha q - \alpha - 1}$ so that the optimal a_n is

$$a_n^0 = O\left(n^{-1/2+\lambda}\right) \text{ for all } \lambda > \frac{\alpha+3}{2\alpha(2q-1)-2}.$$

Since $\sup_{y \in J_n} |F_{\beta^*}(y) - F(y)| = O\left(n^{-\frac{\alpha}{\alpha+1}\eta^*}\right)$ with $J_n = (\xi_p - a_n^0, \xi_p + a_n^0)$, and

$\eta^* = -\beta^*(1 + 1/\alpha - q) - 1/\alpha$, the optimal β^* is (in view of Lemma 4)

$$\beta_0^* = \beta_0 + \frac{\beta_0 - 1}{2 - \frac{4\alpha q}{\alpha + 1}}. \quad (\text{Note also that } \beta_0^* \geq \beta_0). \quad \text{This implies the optimal}$$

$$(2.17) \quad R_n = O(n^{-3/4+\gamma})$$

for all

$$(2.18) \quad \gamma > \frac{\alpha^2(8q-5) + 2\alpha(10q-9) - 13}{4(2\alpha q - \alpha - 1)^2}.$$

End of Remark.

Proof of Theorem 1: Theorem 2 and Proposition 1 provide us with the technology to establish the main result. Due to Theorem 2 and Remark 3, we may select N_0 such that for all $n \geq N_0$, $X_{p,n} \in I_n^0 = J_n$. Then, also, $F_n(X_{p,n}) = \lceil n \cdot p \rceil / n$. Since

$$\sup_{y \in J_n} |F_{\beta_0^*}(y) - F(y)| = O\left(n^{-\frac{\alpha}{\alpha+1}\eta_0^*}\right) \quad \text{and} \quad \sup_{y \in J_n} |F_n(y) - F_n^{\beta_0^*}(y)| = O\left(n^{-\frac{\alpha}{\alpha+1}\eta_0^*}\right) \quad \text{by}$$

Lemma 4, using Proposition 1 (with $\beta^* = \beta_0^*$) applied to $y = X_{p,n}$ gives

$$(2.19) \quad \frac{\lceil n \cdot p \rceil}{n} = F_n(\xi_p) + F(X_{p,n}) - F(\xi_p) + O(n^{-3/4+\gamma})$$

for all γ satisfying (2.18). Since, by assumption, F is sufficiently smooth within the neighborhood B_p of ξ_p we may use Taylor's theorem (in Young's form) to assert that

$$(2.20) \quad F(X_{p,n}) = F(\xi_p) + (X_{p,n} - \xi_p)f(\xi_p) + O((a_n^0)^2).$$

Consequently, combining (2.19) and (2.20)

$$X_{p,n} = \xi_p + \frac{\lceil n \cdot p \rceil / n - F_n(\xi_p)}{f(\xi_p)} + O(\max\{(a_n^0)^2, n^{-3/4+\gamma}\}).$$

Comparison of the rates $(a_n^0)^2$ and $n^{-3/4+\gamma}$, and observing that $\lceil n \cdot p \rceil / n = p + O(n^{-1})$

gives the desired result. This completes the proof of Theorem 1.

Appendix

The appendix contains the Lemmas used in the proof of Theorem 1.

Lemma 1: Let $X(i)$, $X_\beta(i)$, $i = 1, \dots, n$ be two sequences of random variables with stationary distribution functions F and F_β , respectively. F_β may depend on n . Assume that F has bounded derivative in some neighborhood B_p of ξ_p with $F(\xi_p) = p$. Assume also that

$$(A.1) \quad \max_{1 \leq i \leq n} |X(i) - X_\beta(i)| \leq c \cdot \epsilon^* \cdot n^{-\rho} \quad \text{a.s.}$$

where ρ is a positive constant and ϵ^* is a random variable independent of n and such that

$$(A.2) \quad P(\epsilon^* \geq \xi) \leq c \cdot \xi^{-\alpha}$$

for some $\alpha > 0$ and any $\xi > 0$. Then

$$\limsup_{n \rightarrow \infty} n^{\alpha\rho/(1+\alpha)} \sup_{y \in B_p} |F(y) - F_\beta(y)| < \infty.$$

Proof: We may conclude from (A.1) that for any λ with $0 < \lambda < \rho$ and all $y \in B_p$

$$P(X(i) \leq y - n^{-\lambda}) - P(\epsilon^* \geq \frac{1}{c} n^{\rho-\lambda}) \leq P(X_\beta(i) \leq y) \leq P(X(i) \leq y + n^{-\lambda}) + P(\epsilon^* \geq \frac{1}{c} n^{\rho-\lambda})$$

Using (A.2), it is clear that

$$F(y - n^{-\lambda}) - O(n^{-\alpha(\rho-\lambda)}) \leq F_\beta(y) \leq F(y + n^{-\lambda}) + O(n^{-\alpha(\rho-\lambda)})$$

and consequently, since F has bounded derivative over B_p ,

$$|F(y) - F_\beta(y)| = O(n^{-\lambda} + n^{-\alpha(\rho-\lambda)}).$$

Selecting $\lambda = \alpha\rho/(1 + \alpha)$ we obtain the best possible rate

$$\sup_{y \in B_p} |F(y) - F_\beta(y)| = O(n^{-\alpha\rho/(1 + \alpha)})$$

This completes the proof.

Theorem 2 utilizes this Lemma with $\rho = -\beta(1 + 1/\alpha - q) - 1/\alpha$.

Lemma 2: Let $J = \{X(i) : i \in \{1, 2, \dots, n\}\}$ be a set of random variables and $S_{n,k}$

$k = 1, \dots, r$ be r nonempty disjoint subsets of cardinality n_k of the set J with $\bigcup_{k=1}^r S_{n,k} = J$.

Then, for any $0 < p < 1$, the p -th sample quantile $X_{p,n}$ of J and the p -th sample quantiles X_{p,n_k}^k of $S_{n,k}$ satisfy the inequalities

$$\min_{1 \leq k \leq r} X_{p,n_k}^k \leq X_{p,n} \leq \max_{1 \leq k \leq r} X_{p,n_k}^k.$$

Proof: Since $\lceil n_k \cdot p \rceil \geq n_k \cdot p > \lceil n_k \cdot p \rceil - 1$ for all $k = 1, \dots, r$ it is true that

$$\sum_{k=1}^r \lceil n_k \cdot p \rceil \geq \lceil n \cdot p \rceil > \sum_{k=1}^r \lceil n_k \cdot p \rceil - r$$

and hence

$$(A.3) \quad \# \{X(i) : X(i) < \min_{1 \leq k \leq r} X_{p,n_k}^k\} \leq \sum_{k=1}^r (\lceil n_k \cdot p \rceil - 1) < \lceil n \cdot p \rceil$$

and

$$(A.4) \quad \# \{X(i) : X(i) \leq \max_{1 \leq k \leq r} X_{p,n_k}^k\} \geq \sum_{k=1}^r \lceil n_k \cdot p \rceil \geq \lceil n \cdot p \rceil.$$

(A.3) implies that $\min_{1 \leq k \leq r} X_{p,n_k}^k \leq X_{p,n}$ and (A.4) implies that $\max_{1 \leq k \leq r} X_{p,n_k}^k \geq X_{p,n}$.

Remark 5: Both r and n_k may be functions of n .

Lemma 3: For integers n, k, v let $E_{n,k,v}$ be random variables having an exponential distribution with mean 1. Also, for all n, k let E_{n,k,v_1} and E_{n,k,v_2} be independent whenever $v_1 \neq v_2$. Then, for any $0 < \beta < 1$,

$$\limsup_{n \rightarrow \infty} \left[\frac{n^{1-\beta}}{\log n} \right]^{1/2} \max_{1 \leq k \leq \lceil n^\beta \rceil} \left| \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v} - 1}{v} \right| < \infty \quad \text{a.s.,}$$

where the n_k are defined after (2.3a) in Section 2.

Proof: We start by deriving sharp bounds for

$$(A.5) \quad P \left[\left[\frac{n^{1-\beta}}{\log n} \right]^{1/2} \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v} - 1}{v} \geq M \right]$$

Using Chernoff's bound (Chernoff (1952)) we obtain that this probability is bounded by

$$\left[\prod_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{1}{1-t/v} \exp(-t/v) \right] \cdot \exp(-tMn^{1/2(\beta-1)} \cdot (\log n)^{1/2})$$

for all $0 \leq t \leq \lceil p \cdot n_k \rceil$. We choose $t = c_3 n^{1/2(1-\beta)} \cdot (\log n)^{1/2}$ with some positive constant c_3 to be determined later. Then

$$\begin{aligned} \log \left[\prod_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{1}{1-t/v} \exp(-t/v) \right] &= \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} -\log \left(1 - \frac{t}{v} \right) - \frac{t}{v} \\ &= \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{t^2}{2v} + o(n^{-1/2+\epsilon}), \quad \epsilon > 0 \\ &\leq \frac{c_3^2 n^{1-\beta} \log n}{2p \cdot n_k} + o(n^{-1/2+\epsilon}) \\ &\leq \frac{c_3^2}{2p} \log n \quad \text{for } n \text{ large enough.} \end{aligned}$$

Hence

$$(A.6) \quad P \left[\left(\frac{n^{1-\beta}}{\log n} \right)^{1/2} \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v} - 1}{v} \geq M \right] \leq n^{+c_3^2/2p - c_3 M}$$

Taking $c_3 = p \cdot M$ we may make, for given p , the exponent of n on the RHS of (A.6) as small as we want by increasing M . Hence for n and M large enough, the probability in (A.6) is bounded by $n^{-p \cdot M/2} = \psi(n, p, M)$ say.

The same argument can be applied to

$$(-1) \left(\frac{n^{1-\beta}}{\log n} \right)^{1/2} \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v} - 1}{v}.$$

Combining both, it is obvious that

$$P \left[\left(\frac{n^{1-\beta}}{\log n} \right)^{1/2} \left| \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v} - 1}{v} \right| \geq M \right] \leq 2\psi(n, p, M)$$

and the bound is independent of k . Exploiting this uniformity and choosing M large enough we get

$$(A.7) \quad \sum_{n=N_0}^{\infty} P \left[\max_{1 \leq k \leq \lceil n^\beta \rceil} \left[\frac{n^{1-\beta}}{\log n} \right]^{1/2} \left| \sum_{v=\lceil p \cdot n_k \rceil}^{n_k} \frac{E_{n,k,v} - 1}{v} \right| \geq M \right] < \infty$$

and the Borel-Cantelli lemma produces the desired result.

Lemma 4: With the notation of Section 2

$$\sup_{y \in J_n} |F_n(y) - F_n^{\beta_0^*}(y)| = O(n^{-\frac{\alpha}{\alpha+1}\eta_0^*})$$

where $J_n = (\xi_p - a_n^0, \xi_p + a_n^0)$, and $a_n^0, \beta_0, \beta_0^*$ as in Remark 4.

Proof: Choose r such that $\frac{\alpha}{\alpha+1}(\eta_0^* - \eta_0) < r < \eta_0^* - \frac{\alpha}{\alpha+1}\eta_0$, $d_n = \lceil cn^r \rceil$ and

define $y_{n,v} = \xi_p + a_n^0 d_n^{-1} \cdot v$.

For n sufficiently large and all v with $|v| \leq d_n$, the empirical distribution function evaluated at $y_{n,v}$,

$$F_n(y_{n,v}) = \frac{1}{n} \cdot \sum_{i=1}^n \chi(X(i) \leq y_{n,v})$$

is upperbounded a.s. by

$$\frac{1}{n} \sum_{i=1}^n \chi(X_{\beta_0^*}(i) \leq y_{n,v} + a_n^0 d_n^{-1}) + \frac{1}{n} \sum_{i=1}^n \chi(X_{\beta_0^*}(i) - X(i) > a_n^0 d_n^{-1})$$

Similarly,

$$F_n(y_{n,v}) \geq \frac{1}{n} \sum_{i=1}^n \chi(X_{\beta_0^*}(i) \leq y_{n,v} - a_n^0 d_n^{-1}) - \frac{1}{n} \sum_{i=1}^n \chi(X(i) - X_{\beta_0^*}(i) > a_n^0 d_n^{-1})$$

so that

$$(A.8) \quad |F_n(y_{n,v}) - F_n^{\beta_0^*}(y_{n,v})| \leq \max\{|F_n^{\beta_0^*}(y_{n,v} \pm a_n^0 d_n^{-1}) - F_n^{\beta_0^*}(y_{n,v})|\} \\ + \frac{1}{n} \sum_{i=1}^n \chi(|X(i) - X_{\beta_0^*}(i)| > a_n^0 d_n^{-1})$$

where here the max is to be taken over the choice of signs in the argument of

$$F_n^{\beta_0^*}(y_{n,v} \pm a_n^0 d_n^{-1}).$$

It is also easy to show that

$$\begin{aligned} \sup_{y \in J_n} |F_n(y) - F_n^{\beta_0^*}(y)| &\leq \max_{|v| \leq d_n} |F_n(y_{n,v}) - F_n^{\beta_0^*}(y_{n,v})| \\ &\quad + \max_{-d_n \leq v < d_n} |F_n^{\beta_0^*}(y_{n,v+1}) - F_n^{\beta_0^*}(y_{n,v})| \end{aligned}$$

Combining (A.8) and the previous inequality we see that

$$\begin{aligned} \sup_{y \in J_n} |F_n(y) - F_n^{\beta_0^*}(y)| &\leq 2 \cdot \max_{|v| \leq d_n} |F_n^{\beta_0^*}(y_{n,v+1}) - F_n^{\beta_0^*}(y_{n,v})| \\ &\quad + \frac{1}{n} \sum_{i=1}^n \chi(|X(i) - X_{\beta_0^*}(i)| > d_n^{-1}) \\ &= 2S(1) + S(2), \text{ say.} \end{aligned}$$

To show the required rate for $S(1)$ one makes use of the fact that

$n(F_n^{\beta_0^*}(y_{n,v+1}) - F_n^{\beta_0^*}(y_{n,v}))$ has the same distribution as

$$\sum_{i=1}^n \chi(y_{n,v} < X_{\beta_0^*}(i) \leq y_{n,v+1})$$

and $\chi(y_{n,v} < X_{\beta_0^*}(i) \leq y_{n,v+1})$, $i = 1, \dots, n$ is a sequence of $\lceil n^{\beta_0^*} \rceil$ -dependent Bernoulli random variables with parameter equal to

$$F_{\beta_0^*}(y_{n,v+1}) - F_{\beta_0^*}(y_{n,v}) = O(a_n^0 d_n^{-1}).$$

The inequalities in Hoeffding (1963), Section 5d admit a straightforward extension to $\lceil n^{\beta_0^*} \rceil$ -dependent random variables and, using these inequalities together with the Borel Cantelli lemma proves the required rate for $S(1)$.

To show that

$$\frac{1}{n} \sum_{i=1}^n \chi(|X(i) - X_{\beta_0^*}(i)| > a_n^0 d_n^{-1}) = O(n^{-\frac{\alpha}{\alpha+1} \eta_0^*})$$

it is sufficient to realize that

$$\max_{1 \leq i \leq n} |X(i) - X_{\beta_0^*}(i)| \leq c \cdot \epsilon^* n^{-\eta_0^*} \text{ a.s.}$$

by (2.1), that $a_n^0 d_n^{-1} > c \epsilon^* n^{-\eta_0^*}$ for all sufficiently large n , and that ϵ^* is a.s. finite.

Lemma 5: With the notation of Section 2 it is true for all β with $0 < \beta < 1$ that

$$(X_\beta)_{p,n} \in B_p \text{ a.s. for all } n \text{ large enough.}$$

Here B_p is the fixed neighborhood of ξ_p over which the density of $X(i)$ is bounded away from 0 and infinity.

Proof: We will show that $(X_\beta)_{p,n} \rightarrow \xi_p$ a.s. from which the statement follows.

For $\delta > 0$ it is clear that

$$F(\xi_p - \delta) < p < F(\xi_p + \delta)$$

Now, if we can also show that

$$(A.9) \quad \begin{aligned} F_n^\beta(\xi_p - \delta) &\rightarrow F(\xi_p - \delta) \text{ a.s.} \\ F_n^\beta(\xi_p + \delta) &\rightarrow F(\xi_p + \delta) \text{ a.s.} \end{aligned}$$

then it follows that

$$F_n^\beta(\xi_p - \delta) < p < F_n^\beta(\xi_p + \delta) \text{ a.s. for all } n \text{ large enough}$$

and therefore

$$\begin{aligned} \xi_p - \delta &\leq (X_\beta)_{p,n} \leq \xi_p + \delta \text{ a.s. for all } n \text{ large enough,} \\ \text{because clearly } F_n^\beta(\xi_p + \delta) &\geq p \text{ iff } \xi_p + \delta \geq (F_n^\beta)^{-1}(p) = (X_\beta)_{p,n}. \end{aligned}$$

So we only need to show (A.9). We prove only that

$$F_n^\beta(\xi_p - \delta) \rightarrow F(\xi_p - \delta) \text{ a.s.}$$

i.e. for all $\epsilon > 0$

$$(A.10) \quad \sum_n P(|n^{-1} \sum_{i=1}^n \chi(X_\beta(i) \leq \xi_p - \delta) - F(\xi_p - \delta)| > \epsilon) < \infty$$

The LHS of (A.10) is upperbounded by

$$(A.11) \quad \sum_n P(|n^{-1} \sum_{i=1}^n \chi(X_\beta(i) \leq \xi_p - \delta) - F_\beta(\xi_p - \delta)| > \epsilon - c n^{-\frac{\alpha}{\alpha+1}\eta})$$

by (2.3). By construction $\chi(X_{\beta}(i) \leq \xi_p - \delta)$, $i = 1, \dots, n$ is an $\lceil n^{\beta} \rceil$ -dependent sequence of Bernoulli random variables and again we may use the results in (Hoeffding 1963 Section 5d) to prove the finiteness of (A.11).

References:

Andrews, D.W.K. (1983). Nonstrong mixing autoregressive processes. Preprint, Cowles Foundation for Research in Economics, Yale University.

Babu, G.J. and Singh, K. (1978). On deviations between empirical and quantile processes for mixing random variables. *J. Multivariate Anal.* **8**, 532-549.

Bahadur, R.R. (1966). A note on quantiles in large samples. *Ann. Math. Statist.* **37**, 577-580.

Chanda, K.C. (1974). Strong mixing properties of linear stochastic processes. *J. Appl. Prob.* **11**, 401-408.

Chernoff, H. (1952). A measure of asymptotic efficiency for tests of a hypothesis based on the sum of observations. *Ann. Math. Statist.* **23**, 493-507.

Duttweiler, D.L. (1973). The mean-square error of Bahadur's order-statistic approximation. *Ann. Statist.* **1**, 446-453.

Eicker, F. (1966). On the asymptotic representation of sample quantiles. (Abstract). *Ann. Math. Statist.* **27**, 1424.

Gosh, J.K. (1971). A new proof of the Bahadur representation of quantiles and an application. *Ann. Math. Statist.* **42**, 1957-1961.

Hoeffding, W. (1963). Probability inequalities for sums of bounded random variables.

J. Amer. Statist. Assoc. **58**, 13-30.

Kiefer, J. (1967). On Bahadur's representation of sample quantiles. *Ann. Math. Statist.* **38**, 1323-1342.

Kiefer, J. (1970a). Deviations between the sample quantile process and the sample df. Proc. Conference on Nonparametric Techniques in Statistical Inference, Bloomington (ed. by M.L. Puri), Cambridge University Press, 299-319.

Kiefer, J. (1970b). Old and new methods for studying order statistics and sample quantiles. *ibid.*, 349-357.

Sen, P.K. (1968). Asymptotic normality of sample quantiles for m-dependent processes. *Ann. Math. Statist.* **39**, 1724-1730.

Sen, P.K. (1972). On the Bahadur representation of sample quantiles for sequences of ϕ -mixing random variables. *J. Multivariate Anal.* **2**, 77-95.

Shorack, G.R. and Wellner, J.A. (1986). Empirical processes with applications to statistics. Wiley, New York.

Department of Statistics

University of California

Berkeley, CA 94720

TECHNICAL REPORTS

Statistics Department

University of California, Berkeley

1. BREIMAN, L. and FREEDMAN, D. (Nov. 1981, revised Feb. 1982). How many variables should be entered in a regression equation? Jour. Amer. Statist. Assoc., March 1983, 78, No. 381, 131-136.
2. BRILLINGER, D. R. (Jan. 1982). Some contrasting examples of the time and frequency domain approaches to time series analysis. Time Series Methods in Hydrosciences, (A. H. El-Shaarawi and S. R. Esterby, eds.) Elsevier Scientific Publishing Co., Amsterdam, 1982, pp. 1-15.
3. DOKSUM, K. A. (Jan. 1982). On the performance of estimates in proportional hazard and log-linear models. Survival Analysis, (John Crowley and Richard A. Johnson, eds.) IMS Lecture Notes - Monograph Series, (Shanti S. Gupta, series ed.) 1982, 74-84.
4. BICKEL, P. J. and BREIMAN, L. (Feb. 1982). Sums of functions of nearest neighbor distances, moment bounds, limit theorems and a goodness of fit test. Ann. Prob., Feb. 1982, 11, No. 1, 185-214.
5. BRILLINGER, D. R. and TUKEY, J. W. (March 1982). Spectrum estimation and system identification relying on a Fourier transform. The Collected Works of J. W. Tukey, vol. 2, Wadsworth, 1985, 1001-1141.
6. BERAN, R. (May 1982). Jackknife approximation to bootstrap estimates. Ann. Statist., March 1984, 12 No. 1, 101-118.
7. BICKEL, P. J. and FREEDMAN, D. A. (June 1982). Bootstrapping regression models with many parameters. Lehmann Festschrift, (P. J. Bickel, K. Doksum and J. L. Hodges, Jr., eds.) Wadsworth Press, Belmont, 1983, 28-48.
8. BICKEL, P. J. and COLLINS, J. (March 1982). Minimizing Fisher information over mixtures of distributions. Sankhyā, 1983, 45, Series A, Pt. 1, 1-19.
9. BREIMAN, L. and FRIEDMAN, J. (July 1982). Estimating optimal transformations for multiple regression and correlation.
10. FREEDMAN, D. A. and PETERS, S. (July 1982, revised Aug. 1983). Bootstrapping a regression equation: some empirical results. JASA, 1984, 79, 97-106.
11. EATON, M. L. and FREEDMAN, D. A. (Sept. 1982). A remark on adjusting for covariates in multiple regression.
12. BICKEL, P. J. (April 1982). Minimax estimation of the mean of a mean of a normal distribution subject to doing well at a point. Recent Advances in Statistics, Academic Press, 1983.
14. FREEDMAN, D. A., ROTHENBERG, T. and SUTCH, R. (Oct. 1982). A review of a residential energy end use model.
15. BRILLINGER, D. and PREISLER, H. (Nov. 1982). Maximum likelihood estimation in a latent variable problem. Studies in Econometrics, Time Series, and Multivariate Statistics, (eds. S. Karlin, T. Amemiya, L. A. Goodman). Academic Press, New York, 1983, pp. 31-65.
16. BICKEL, P. J. (Nov. 1982). Robust regression based on infinitesimal neighborhoods. Ann. Statist., Dec. 1984, 12, 1349-1368.
17. DRAPER, D. C. (Feb. 1983). Rank-based robust analysis of linear models. I. Exposition and review. Statistical Science, 1988, Vol.3 No. 2 239-271.
18. DRAPER, D. C. (Feb 1983). Rank-based robust inference in regression models with several observations per cell.
19. FREEDMAN, D. A. and FIENBERG, S. (Feb. 1983, revised April 1983). Statistics and the scientific method, Comments on and reactions to Freedman, A rejoinder to Fienberg's comments. Springer New York 1985 Cohort Analysis in Social Research, (W. M. Mason and S. E. Fienberg, eds.).
20. FREEDMAN, D. A. and PETERS, S. C. (March 1983, revised Jan. 1984). Using the bootstrap to evaluate forecasting equations. J. of Forecasting, 1985, Vol. 4, 251-262.
21. FREEDMAN, D. A. and PETERS, S. C. (March 1983, revised Aug. 1983). Bootstrapping an econometric model: some empirical results. JBES, 1985, 2, 150-158.
22. FREEDMAN, D. A. (March 1983). Structural-equation models: a case study.
23. DAGGETT, R. S. and FREEDMAN, D. (April 1983, revised Sept. 1983). Econometrics and the law: a case study in the proof of antitrust damages. Proc. of the Berkeley Conference, in honor of Jerzy Neyman and Jack Kiefer. Vol I pp. 123-172. (L. Le Cam, R. Olshen eds.) Wadsworth, 1985.

24. DOKSUM, K. and YANDELL, B. (April 1983). Tests for exponentiality. Handbook of Statistics, (P. R. Krishnaiah and P. K. Sen, eds.) 4, 1984, 579-611.
25. FREEDMAN, D. A. (May 1983). Comments on a paper by Markus.
26. FREEDMAN, D. (Oct. 1983, revised March 1984). On bootstrapping two-stage least-squares estimates in stationary linear models. Ann. Statist., 1984, 12, 827-842.
27. DOKSUM, K. A. (Dec. 1983). An extension of partial likelihood methods for proportional hazard models to general transformation models. Ann. Statist., 1987, 15, 325-345.
28. BICKEL, P. J., GOETZE, F. and VAN ZWET, W. R. (Jan. 1984). A simple analysis of third order efficiency of estimate Proc. of the Neyman-Kiefer Conference, (L. Le Cam, ed.) Wadsworth, 1985.
29. BICKEL, P. J. and FREEDMAN, D. A. Asymptotic normality and the bootstrap in stratified sampling. Ann. Statist. 12 470-482.
30. FREEDMAN, D. A. (Jan. 1984). The mean vs. the median: a case study in 4-R Act litigation. JBES, 1985 Vol 3 pp. 1-13.
31. STONE, C. J. (Feb. 1984). An asymptotically optimal window selection rule for kernel density estimates. Ann. Statist., Dec. 1984, 12, 1285-1297.
32. BREIMAN, L. (May 1984). Nail finders, edifices, and Oz.
33. STONE, C. J. (Oct. 1984). Additive regression and other nonparametric models. Ann. Statist., 1985, 13, 689-705.
34. STONE, C. J. (June 1984). An asymptotically optimal histogram selection rule. Proc. of the Berkeley Conf. in Honor of Jerzy Neyman and Jack Kiefer (L. Le Cam and R. A. Olshen, eds.), II, 513-520.
35. FREEDMAN, D. A. and NAVIDI, W. C. (Sept. 1984, revised Jan. 1985). Regression models for adjusting the 1980 Census. Statistical Science, Feb 1986, Vol. 1, No. 1, 3-39.
36. FREEDMAN, D. A. (Sept. 1984, revised Nov. 1984). De Finetti's theorem in continuous time.
37. DIACONIS, P. and FREEDMAN, D. (Oct. 1984). An elementary proof of Stirling's formula. Amer. Math Monthly, Feb 1986, Vol. 93, No. 2, 123-125.
38. LE CAM, L. (Nov. 1984). Sur l'approximation de familles de mesures par des familles Gaussiennes. Ann. Inst. Henri Poincaré, 1985, 21, 225-287.
39. DIACONIS, P. and FREEDMAN, D. A. (Nov. 1984). A note on weak star uniformities.
40. BREIMAN, L. and IHAKA, R. (Dec. 1984). Nonlinear discriminant analysis via SCALING and ACE.
41. STONE, C. J. (Jan. 1985). The dimensionality reduction principle for generalized additive models.
42. LE CAM, L. (Jan. 1985). On the normal approximation for sums of independent variables.
43. BICKEL, P. J. and YAHAV, J. A. (1985). On estimating the number of unseen species: how many executions were there?
44. BRILLINGER, D. R. (1985). The natural variability of vital rates and associated statistics. Biometrics, to appear.
45. BRILLINGER, D. R. (1985). Fourier inference: some methods for the analysis of array and nonGaussian series data. Water Resources Bulletin, 1985, 21, 743-756.
46. BREIMAN, L. and STONE, C. J. (1985). Broad spectrum estimates and confidence intervals for tail quantiles.
47. DABROWSKA, D. M. and DOKSUM, K. A. (1985, revised March 1987). Partial likelihood in transformation models with censored data. Scandinavian J. Statist., 1988, 15, 1-23.
48. HAYCOCK, K. A. and BRILLINGER, D. R. (November 1985). LIBDRB: A subroutine library for elementary time series analysis.
49. BRILLINGER, D. R. (October 1985). Fitting cosines: some procedures and some physical examples. Joshi Festschrift, 1986. D. Reidel.
50. BRILLINGER, D. R. (November 1985). What do seismology and neurophysiology have in common? - Statistics! Comptes Rendus Math. Rep. Acad. Sci. Canada, January, 1986.
51. COX, D. D. and O'SULLIVAN, F. (October 1985). Analysis of penalized likelihood-type estimators with application to generalized smoothing in Sobolev Spaces.

52. O'SULLIVAN, F. (November 1985). A practical perspective on ill-posed inverse problems: A review with some new developments. To appear in Journal of Statistical Science.
53. LE CAM, L. and YANG, G. L. (November 1985, revised March 1987). On the preservation of local asymptotic normality under information loss.
54. BLACKWELL, D. (November 1985). Approximate normality of large products.
55. FREEDMAN, D. A. (June 1987). As others see us: A case study in path analysis. Journal of Educational Statistics, 12, 101-128.
56. LE CAM, L. and YANG, G. L. (January 1986). Replaced by No. 68.
57. LE CAM, L. (February 1986). On the Bernstein - von Mises theorem.
58. O'SULLIVAN, F. (January 1986). Estimation of Densities and Hazards by the Method of Penalized likelihood.
59. ALDOUS, D. and DIACONIS, P. (February 1986). Strong Uniform Times and Finite Random Walks.
60. ALDOUS, D. (March 1986). On the Markov Chain simulation Method for Uniform Combinatorial Distributions and Simulated Annealing.
61. CHENG, C-S. (April 1986). An Optimization Problem with Applications to Optimal Design Theory.
62. CHENG, C-S., MAJUMDAR, D., STUFKEN, J. & TURE, T. E. (May 1986, revised Jan 1987). Optimal step type design for comparing test treatments with a control.
63. CHENG, C-S. (May 1986, revised Jan. 1987). An Application of the Kiefer-Wolfowitz Equivalence Theorem.
64. O'SULLIVAN, F. (May 1986). Nonparametric Estimation in the Cox Proportional Hazards Model.
65. ALDOUS, D. (JUNE 1986). Finite-Time Implications of Relaxation Times for Stochastically Monotone Processes.
66. PITMAN, J. (JULY 1986, revised November 1986). Stationary Excursions.
67. DABROWSKA, D. and DOKSUM, K. (July 1986, revised November 1986). Estimates and confidence intervals for median and mean life in the proportional hazard model with censored data. Biometrika, 1987, 74, 799-808.
68. LE CAM, L. and YANG, G.L. (July 1986). Distinguished Statistics, Loss of information and a theorem of Robert B. Davies (Fourth edition).
69. STONE, C.J. (July 1986). Asymptotic properties of logspline density estimation.
71. BICKEL, P.J. and YAHAV, J.A. (July 1986). Richardson Extrapolation and the Bootstrap.
72. LEHMANN, E.L. (July 1986). Statistics - an overview.
73. STONE, C.J. (August 1986). A nonparametric framework for statistical modelling.
74. BIANE, PH. and YOR, M. (August 1986). A relation between Lévy's stochastic area formula, Legendre polynomial, and some continued fractions of Gauss.
75. LEHMANN, E.L. (August 1986, revised July 1987). Comparing Location Experiments.
76. O'SULLIVAN, F. (September 1986). Relative risk estimation.
77. O'SULLIVAN, F. (September 1986). Deconvolution of episodic hormone data.
78. PITMAN, J. & YOR, M. (September 1987). Further asymptotic laws of planar Brownian motion.
79. FREEDMAN, D.A. & ZEISEL, H. (November 1986). From mouse to man: The quantitative assessment of cancer risks. To appear in Statistical Science.
80. BRILLINGER, D.R. (October 1986). Maximum likelihood analysis of spike trains of interacting nerve cells.
81. DABROWSKA, D.M. (November 1986). Nonparametric regression with censored survival time data.
82. DOKSUM, K.J. and LO, A.Y. (Nov 1986, revised Aug 1988). Consistent and robust Bayes Procedures for Location based on Partial Information.
83. DABROWSKA, D.M., DOKSUM, K.A. and MIURA, R. (November 1986). Rank estimates in a class of semiparametric two-sample models.

84. BRILLINGER, D. (December 1986). Some statistical methods for random process data from seismology and neurophysiology.
85. DIACONIS, P. and FREEDMAN, D. (December 1986). A dozen de Finetti-style results in search of a theory. Ann. Inst. Henri Poincaré, 1987, 23, 397-423.
86. DABROWSKA, D.M. (January 1987). Uniform consistency of nearest neighbour and kernel conditional Kaplan - Meier estimates.
87. FREEDMAN, D.A., NAVIDI, W. and PETERS, S.C. (February 1987). On the impact of variable selection in fitting regression equations.
88. ALDOUS, D. (February 1987, revised April 1987). Hashing with linear probing, under non-uniform probabilities.
89. DABROWSKA, D.M. and DOKSUM, K.A. (March 1987, revised January 1988). Estimating and testing in a two sample generalized odds rate model. J. Amer. Statist. Assoc., 1988, 83, 744-749.
90. DABROWSKA, D.M. (March 1987). Rank tests for matched pair experiments with censored data.
91. DIACONIS, P and FREEDMAN, D.A. (April 1988). Conditional limit theorems for exponential families and finite versions of de Finetti's theorem. To appear in the Journal of Applied Probability.
92. DABROWSKA, D.M. (April 1987, revised September 1987). Kaplan-Meier estimate on the plane.
- 92a. ALDOUS, D. (April 1987). The Harmonic mean formula for probabilities of Unions: Applications to sparse random graphs.
93. DABROWSKA, D.M. (June 1987, revised Feb 1988). Nonparametric quantile regression with censored data.
94. DONOHO, D.L. & STARK, P.B. (June 1987). Uncertainty principles and signal recovery.
95. CANCELLED
96. BRILLINGER, D.R. (June 1987). Some examples of the statistical analysis of seismological data. To appear in Proceedings, Centennial Anniversary Symposium, Seismographic Stations, University of California, Berkeley.
97. FREEDMAN, D.A. and NAVIDI, W. (June 1987). On the multi-stage model for carcinogenesis. To appear in Environmental Health Perspectives.
98. O'SULLIVAN, F. and WONG, T. (June 1987). Determining a function diffusion coefficient in the heat equation.
99. O'SULLIVAN, F. (June 1987). Constrained non-linear regularization with application to some system identification problems.
100. LE CAM, L. (July 1987, revised Nov 1987). On the standard asymptotic confidence ellipsoids of Wald.
101. DONOHO, D.L. and LIU, R.C. (July 1987). Pathologies of some minimum distance estimators. Annals of Statistics, June, 1988.
102. BRILLINGER, D.R., DOWNING, K.H. and GLAESER, R.M. (July 1987). Some statistical aspects of low-dose electron imaging of crystals.
103. LE CAM, L. (August 1987). Harald Cramér and sums of independent random variables.
104. DONOHO, A.W., DONOHO, D.L. and GASKO, M. (August 1987). Macspin: Dynamic graphics on a desktop computer. IEEE Computer Graphics and applications, June, 1988.
105. DONOHO, D.L. and LIU, R.C. (August 1987). On minimax estimation of linear functionals.
106. DABROWSKA, D.M. (August 1987). Kaplan-Meier estimate on the plane: weak convergence, LIL and the bootstrap.
107. CHENG, C-S. (Aug 1987, revised Oct 1988). Some orthogonal main-effect plans for asymmetrical factorials.
108. CHENG, C-S. and JACROUX, M. (August 1987). On the construction of trend-free run orders of two-level factorial designs.
109. KLASS, M.J. (August 1987). Maximizing $E \max_{1 \leq k \leq n} S_k^+ / ES_n^+$: A prophet inequality for sums of I.I.D. mean zero variates.
110. DONOHO, D.L. and LIU, R.C. (August 1987). The "automatic" robustness of minimum distance functionals. Annals of Statistics, June, 1988.
111. BICKEL, P.J. and GHOSH, J.K. (August 1987, revised June 1988). A decomposition for the likelihood ratio statistic and the Bartlett correction — a Bayesian argument.

112. BURDZY, K., PITMAN, J.W. and YOR, M. (September 1987). Some asymptotic laws for crossings and excursions.
113. ADHIKARI, A. and PITMAN, J. (September 1987). The shortest planar arc of width 1.
114. RITOV, Y. (September 1987). Estimation in a linear regression model with censored data.
115. BICKEL, P.J. and RITOV, Y. (Sept. 1987, revised Aug 1988). Large sample theory of estimation in biased sampling regression models I.
116. RITOV, Y. and BICKEL, P.J. (Sept.1987, revised Aug. 1988). Achieving information bounds in non and semiparametric models.
117. RITOV, Y. (October 1987). On the convergence of a maximal correlation algorithm with alternating projections.
118. ALDOUS, D.J. (October 1987). Meeting times for independent Markov chains.
119. HESSE, C.H. (October 1987). An asymptotic expansion for the mean of the passage-time distribution of integrated Brownian Motion.
120. DONOHO, D. and LIU, R. (Oct. 1987, revised Mar. 1988, Oct. 1988). Geometrizing rates of convergence, II.
121. BRILLINGER, D.R. (October 1987). Estimating the chances of large earthquakes by radiocarbon dating and statistical modelling. Statistics a Guide to the Unknown, pp. 249-260 (Eds. J.M. Tanur et al.) Wadsworth, Pacific Grove.
122. ALDOUS, D., FLANNERY, B. and PALACIOS, J.L. (November 1987). Two applications of urn processes: The fring analysis of search trees and the simulation of quasi-stationary distributions of Markov chains.
123. DONOHO, D.L., MACGIBBON, B. and LIU, R.C. (Nov.1987, revised July 1988). Minimax risk for hyperrectangles.
124. ALDOUS, D. (November 1987). Stopping times and tightness II.
125. HESSE, C.H. (November 1987). The present state of a stochastic model for sedimentation.
126. DALANG, R.C. (December 1987, revised June 1988). Optimal stopping of two-parameter processes on nonstandard probability spaces.
127. Same as No. 133.
128. DONOHO, D. and GASKO, M. (December 1987). Multivariate generalizations of the median and trimmed mean II.
129. SMITH, D.L. (December 1987). Exponential bounds in Vapnik-Červonenkis classes of index 1.
130. STONE, C.J. (Nov.1987, revised Sept. 1988). Uniform error bounds involving logspline models.
131. Same as No. 140
132. HESSE, C.H. (Dec. 1987, revised June 1989). A Bahadur - Type representation for empirical quantiles of a large class of stationary, possibly infinite - variance, linear processes
133. DONOHO, D.L. and GASKO, M. (December 1987). Multivariate generalizations of the median and trimmed mean, I.
134. DUBINS, L.E. and SCHWARZ, G. (December 1987). A sharp inequality for martingales and stopping-times.
135. FREEDMAN, D.A. and NAVIDI, W. (December 1987). On the risk of lung cancer for ex-smokers.
136. LE CAM, L. (January 1988). On some stochastic models of the effects of radiation on cell survival.
137. DIACONIS, P. and FREEDMAN, D.A. (April 1988). On the uniform consistency of Bayes estimates for multinomial probabilities.
- 137a. DONOHO, D.L. and LIU, R.C. (1987). Geometrizing rates of convergence, I.
138. DONOHO, D.L. and LIU, R.C. (January 1988). Geometrizing rates of convergence, III.
139. BERAN, R. (January 1988). Refining simultaneous confidence sets.
140. HESSE, C.H. (December 1987). Numerical and statistical aspects of neural networks.
141. BRILLINGER, D.R. (Jan. 1988). Two reports on trend analysis: a) An elementary trend analysis of Rio negro levels a Manaus, 1903-1985. b) Consistent detection of a monotonic trend superposed on a stationary time series.
142. DONOHO, D.L. (Jan. 1985, revised Jan. 1988). One-sided inference about functionals of a density.

143. DALANG, R.C. (Feb. 1988, revised Nov. 1988). Randomization in the two-armed bandit problem.
144. DABROWSKA, D.M., DOKSUM, K.A. and SONG, J.K. (February 1988). Graphical comparisons of cumulative hazards for two populations.
145. ALDOUS, D.J. (February 1988). Lower bounds for covering times for reversible Markov Chains and random walks on graphs.
146. BICKEL, P.J. and RITOV, Y. (Feb.1988, revised August 1988). Estimating integrated squared density derivatives.
147. STARK, P.B. (March 1988). Strict bounds and applications.
148. DONOHO, D.L. and STARK, P.B. (March 1988). Rearrangements and smoothing.
149. NOLAN, D. (March 1988). Asymptotics for a multivariate location estimator.
150. SEILLIER, F. (March 1988). Sequential probability forecasts and the probability integral transform.
151. NOLAN, D. (Mar. 1988, revised May 1989). Asymptotics for multivariate trimming.
152. DIACONIS, P. and FREEDMAN, D.A. (April 1988). On a theorem of Kuchler and Lauritzen.
153. DIACONIS, P. and FREEDMAN, D.A. (April 1988). On the problem of types.
154. DOKSUM, K.A. (May 1988). On the correspondence between models in binary regression analysis and survival analysis.
155. LEHMANN, E.L. (May 1988). Jerzy Neyman, 1894-1981.
156. ALDOUS, D.J. (May 1988). Stein's method in a two-dimensional coverage problem.
157. FAN, J. (June 1988). On the optimal rates of convergence for nonparametric deconvolution problem.
158. DABROWSKA, D. (June 1988). Signed-rank tests for censored matched pairs.
159. BERAN, R.J. and MILLAR, P.W. (June 1988). Multivariate symmetry models.
160. BERAN, R.J. and MILLAR, P.W. (June 1988). Tests of fit for logistic models.
161. BREIMAN, L. and PETERS, S. (June 1988). Comparing automatic bivariate smoothers (A public service enterprise).
162. FAN, J. (June 1988). Optimal global rates of convergence for nonparametric deconvolution problem.
163. DIACONIS, P. and FREEDMAN, D.A. (June 1988). A singular measure which is locally uniform. (Revised by Tech Report No. 180).
164. BICKEL, P.J. and KRIEGER, A.M. (July 1988). Confidence bands for a distribution function using the bootstrap.
165. HESSE, C.H. (July 1988). New methods in the analysis of economic time series I.
166. FAN, JIANQING (July 1988). Nonparametric estimation of quadratic functionals in Gaussian white noise.
167. BREIMAN, L., STONE, C.J. and KOOPERBERG, C. (August 1988). Confidence bounds for extreme quantiles.
168. LE CAM, L. (August 1988). Maximum likelihood an introduction.
169. BREIMAN, L. (Aug.1988, revised Feb. 1989). Submodel selection and evaluation in regression I. The X-fixed case and little bootstrap.
170. LE CAM, L. (September 1988). On the Prokhorov distance between the empirical process and the associated Gaussian bridge.
171. STONE, C.J. (September 1988). Large-sample inference for logspline models.
172. ADLER, R.J. and EPSTEIN, R. (September 1988). Intersection local times for infinite systems of planar brownian motions and for the brownian density process.
173. MILLAR, P.W. (October 1988). Optimal estimation in the non-parametric multiplicative intensity model.
174. YOR, M. (October 1988). Interwinings of Bessel processes.
175. ROJO, J. (October 1988). On the concept of tail-heaviness.
176. ABRAHAM, D.M. and RIZZARDI, F. (September 1988). BLSS - The Berkeley interactive statistical system: An overview.

177. MILLAR, P.W. (October 1988). Gamma-funnels in the domain of a probability, with statistical implications.
178. DONOHO, D.L. and LIU, R.C. (October 1988). Hardest one-dimensional subfamilies.
179. DONOHO, D.L. and STARK, P.B. (October 1988). Recovery of sparse signals from data missing low frequencies.
180. FREEDMAN, D.A. and PITMAN, J.A. (Nov. 1988). A measure which is singular and uniformly locally uniform. (Revision of Tech Report No. 163).
181. DOKSUM, K.A. and HOYLAND, ARNLJOT (Nov. 1988, revised Jan. 1989). A model for step-stress accelerated life testing experiments based on Wiener processes and the inverse Gaussian distribution.
182. DALANG, R.C., MORTON, A. and WILLINGER, W. (November 1988). Equivalent martingale measures and no-arbitrage in stochastic securities market models.
183. BERAN, R. (November 1988). Calibrating prediction regions.
184. BARLOW, M.T., PITMAN, J. and YOR, M. (Feb. 1989). On Walsh's Brownian Motions.
185. DALANG, R.C. and WALSH, J.B. (Dec. 1988). Almost-equivalence of the germ-field Markov property and the sharp Markov property of the Brownian sheet.
186. HESSE, C.H. (Dec. 1988). Level-Crossing of integrated Ornstein-Uhlenbeck processes
187. NEVEU, J. and PITMAN, J.W. (Feb. 1989). Renewal property of the extrema and tree property of the excursion of a one-dimensional brownian motion.
188. NEVEU, J. and PITMAN, J.W. (Feb. 1989). The branching process in a brownian excursion.
189. PITMAN, J.W. and YOR, M. (Mar. 1989). Some extensions of the arcsine law.
190. STARK, P.B. (Dec. 1988). Duality and discretization in linear inverse problems.
191. LEHMANN, E.L. and SCHOLZ, F.W. (Jan. 1989). Ancillarity.
192. PEMANTLE, R. (Feb. 1989). A time-dependent version of Pólya's urn.
193. PEMANTLE, R. (Feb. 1989). Nonconvergence to unstable points in urn models and stochastic approximations.
194. PEMANTLE, R. (Feb. 1989, revised May 1989). When are touchpoints limits for generalized Pólya urns.
195. PEMANTLE, R. (Feb. 1989). Random walk in a random environment and first-passage percolation on trees.
196. BARLOW, M., PITMAN, J. and YOR, M. (Feb. 1989). Une extension multidimensionnelle de la loi de l'arc sinus.
197. BREIMAN, L. and SPECTOR, P. (Mar. 1989). Submodel selection and evaluation in regression — the X-random case.
198. BREIMAN, L., TSUR, Y. and ZEMEL, A. (Mar. 1989). A simple estimation procedure for censored regression models with known error distribution.
199. BRILLINGER, D.R. (Mar. 1989). Two papers on bilinear systems: a) A study of second- and third-order spectral procedures and maximum likelihood identification of a bilinear system. b) Some statistical aspects of NMR spectroscopy, *Actas del 2º congreso latinoamericano de probabilidad y estadística matemática*, Caracas, 1985.
200. BRILLINGER, D.R. (Mar. 1989). Two papers on higher-order spectra: a) Parameter estimation for nonGaussian processes via second and third order spectra with an application to some endocrine data. b) Some history of the study of higher-order moments and spectra.
201. DE LA PENA, V. and KLASS, M.J. (April 1989). L bounds for quadratic forms of independent random variables.
202. FREEDMAN, D.A. and NAVIDI, W.C. (April 1989). Testing the independence of competing risks.
203. TERDIK, G. (May 1989). Bilinear state space realization for polynomial stochastic systems.
204. DONOHO, D.L. and JOHNSTONE, I.M. (May 1989). Minimax risk over l_p -Balls.
205. PEMANTLE, R., PROPP, J. and ULLMAN, D. (May 1989). On tensor powers of integer programs.

206. MILASEVIC, P. and NOLAN, D. (May 1989). Estimation on the sphere: A geometric approach.

Copies of these Reports plus the most recent additions to the Technical Report series are available from the Statistics Department technical typist in room 379 Evans Hall or may be requested by mail from:

Department of Statistics
University of California
Berkeley, California 94720

Cost: \$1 per copy.