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LINEAR BOUNDS ON THE EMPIRICAL DISTRIBUTION FUNCTION

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Let Γ_n denote the empirical df of a sample from the uniform (0, 1) df I . Let ξ_{nk} denote the k th smallest observation. Let $\lambda_n > 1$. Let A_n denote the event that Γ_n intersects the line $\lambda_n I$ on $[0, 1]$ and let B_n denote the event that Γ_n intersects the line I/λ_n on $[\xi_{n1}, 1]$. Conditions on λ_n are given that determine whether $P(A_n \text{ i.o.})$ and $P(B_n \text{ i.o.})$ equal 0 or 1. Results for A_n (for B_n) are related to upper class sequences for $1/(n\xi_{n1})$ (for $n\xi_{n2}$).

Upper class sequences for $n\xi_{nk}$, with $k > 1$, are characterized.

In the case of nonidentically distributed random variables, we present the result analogous to $P(A_n \text{ i.o.}) = 0$.

1. Introduction and statement of the theorems. Let ξ_1, \dots, ξ_n be independent uniform (0, 1) random variables having empirical df Γ_n and whose ordered values are $0 \leq \xi_{n1} \leq \dots \leq \xi_{nn} \leq 1$. The true df is the identity function on $[0, 1]$, which we denote by I .

We let $\|f\|_a^b \equiv \sup_{a \leq t \leq b} |f(t)|$, and we simply write $\|f\|$ in case $a = 0$ and $b = 1$.

Note that Γ_n lies entirely below the line λI if and only if $\|\Gamma_n/I\| \geq \lambda$ a.s. for each n . We can not make Γ_n lie entirely above any line through the origin with positive slope since $\Gamma_n(t) = 0$ for $0 \leq t < \xi_{n1}$; however Γ_n lies entirely above the line I/λ on the interval $[\xi_{n1}, 1]$ if and only if $\|I/\Gamma_n\|_{\xi_{n1}} \leq \lambda$. Our main concern in this paper is bounding Γ_n between straight lines through the origin. More precisely, we will characterize upper and lower class sequences for the random variables $\|\Gamma_n/I\|$ and $\|I/\Gamma_n\|_{\xi_{n1}}$.

"In probability upper and lower linear bounds" are well known (see Robbins (1954), Chang (1955) and Renyi (1973)); and see Shorack (1972) for applications. It is known that "a.s. linear bounds" do not exist (see Wellner (1977a)); also see Wellner (1977a) and (1977b) for applications of "a.s. nearly linear bounds."

Discussion of our theorems will be facilitated by contrasting them with the behavior of ξ_{n1} and ξ_{n2} that is set forth in Theorem 1.

THEOREM 1. *Let $k \geq 1$ be a fixed integer.*

(i) (Kiefer). *If $c_n \searrow$ then*

$$\begin{aligned}
 P(n\xi_{nk} \leq c_n \text{ i.o.}) &= 0 && \text{according as} && \sum_{n=1}^{\infty} \frac{c_n^k}{n} < \infty \\
 &= 1 && && = \infty.
 \end{aligned}$$

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(ii) (*Robbins and Siegmund when $k = 1$*). Let $c_n/n \searrow$ and suppose either $c_n \nearrow$ or $\liminf_{n \rightarrow \infty} c_n/\log_2 n \geq 1$. Then

$$\begin{aligned}
 P(n\xi_{nk} > c_n \text{ i.o.}) &= 0 && \text{according as} && \sum_{n=1}^{\infty} \frac{c_n^k}{n} \exp(-c_n) < \infty \\
 &= 1 && && = \infty.
 \end{aligned}$$

THEOREM 2. Let $n\lambda_n \nearrow$. Then

$$\begin{aligned}
 P(\|\Gamma_n/I\| \geq \lambda_n \text{ i.o.}) &= 0 && \text{according as} && \sum_{n=1}^{\infty} \frac{1}{n\lambda_n} < \infty \\
 &= 1 && && = \infty.
 \end{aligned}$$

Note that $\|\Gamma_n/I\| = \max\{i/(n\xi_{ni}) : 1 \leq i \leq n\}$ is $\geq \lambda_n$ if $n\xi_{n1}$ is $\leq c_n \equiv 1/\lambda_n$. Comparing the series criteria of Theorem 1(i) with Theorem 2, it is seen that small values of ξ_{n1} control large values of $\|\Gamma_n/I\|$. Note however that $\|\Gamma_n/I\|$ and $(n\xi_{n1})^{-1}$ have different limiting distributions.

Theorem 2 yields the known result $\limsup_{n \rightarrow \infty} \log \|\Gamma_n/I\|/\log_2 n = 1$ a.s. In fact, $\log \lambda_n = \sum_{i=2}^{p-1} \log_i n + \tau \log_p n$, with $p \geq 2$, is upper class or lower class for $\log \|\Gamma_n/I\|$ according as $\tau > 1$ or $\tau \leq 1$.

THEOREM 3. Let $\lambda_n/n \searrow$ and suppose either $\lambda_n \nearrow$ or $\liminf_{n \rightarrow \infty} \lambda_n/\log_2 n \geq 1$. Then

$$\begin{aligned}
 P(\|I/\Gamma_n\|_{\xi_{n1}}^1 \geq \lambda_n \text{ i.o.}) &= 0 && \text{according as} && \sum_{n=1}^{\infty} \frac{\lambda_n^2}{n} \exp(-\lambda_n) < \infty \\
 &= 1 && && = \infty.
 \end{aligned}$$

Note that $\|I/\Gamma_n\|_{\xi_{n1}}^1 = \max\{n\xi_{n,i+1}/i : 1 \leq i \leq n\}$ is $\geq \lambda_n$ if $n\xi_{n2}$ is $\geq c_n \equiv \lambda_n$. (Here, and in the following, $\xi_{n,n+1} \equiv 1$ for all n .) Comparing the series criteria of Theorem 1(ii) with Theorem 3, it is seen that large values of ξ_{n2} control large values of $\|I/\Gamma_n\|_{\xi_{n1}}^1$. Note however (see Renyi (1973)) that $\|I/\Gamma_n\|_{\xi_{n1}}^1$ and $n\xi_{n2}$ have different limiting distributions.

Theorem 3 yields the known result $\limsup_{n \rightarrow \infty} \|I/\Gamma_n\|_{\xi_{n1}}^1/\log_2 n = 1$ a.s. In fact, $\lambda_n = \log_2 n + 3 \log_3 n + \sum_{i=4}^{p-1} \log_i n + \tau \log_p n$, with $p \geq 4$, is upper class or lower class for $\|I/\Gamma_n\|_{\xi_{n1}}^1$ according as $\tau > 1$ or $\tau \leq 1$.

2. Proofs. Robbins (1954) showed that for any $n \geq 1$

$$(1) \quad P(\|\Gamma_n/I\| \geq \lambda) = 1/\lambda \quad \text{for all } \lambda > 1.$$

PROOF OF THEOREM 2. Suppose $\sum (n\lambda_n)^{-1} < \infty$. Let $n_k \equiv \text{int}(\alpha^k)$ where $\alpha > 1$ is fixed, and where $\text{int}(\cdot)$ denotes that greatest integer function. Note that

$$\begin{aligned}
 (a) \quad \infty &> \sum_{k=2}^{\infty} \sum_{j=n_{k-1}+1}^{n_k} (n\lambda_n)^{-1} \geq \sum_{k=2}^{\infty} (n_k - n_{k-1})(n_k \lambda_{n_k})^{-1} \\
 &\geq \text{constant} \cdot \sum_{k=2}^{\infty} (\lambda_{n_k})^{-1}.
 \end{aligned}$$

Let $A_k \equiv [\max\{\|\Gamma_n/I\| : n_k < n \leq n_{k+1}\} \geq \lambda_{n_k}]$; and note that monotoneity of $n\Gamma_n$

since $d_n^2 \exp(-d_n) \leq \lambda_n^2 \exp(-\lambda_n) + (2 \log_2 n)^2 \exp(-2 \log_2 n)$. Since $B_j \subset D_j \equiv [M_{n_j} \geq (n_j/n_{j+1}) d_{n_{j+1}}]$, it suffices to show that $\sum_2^\infty P(D_j) < \infty$. Now

$$\begin{aligned} \sum_{j=2}^\infty P(D_j) &\leq \sum_{j=2}^\infty 16(n_j/n_{j+1}) d_{n_{j+1}} \exp(-d_{n_{j+1}}) \exp\left(\left(1 - \frac{n_j}{n_{j+1}}\right) d_{n_{j+1}}\right) \\ &\quad \text{by (2)} \\ &\leq (\text{Constant}_\alpha) \sum_{j=2}^\infty d_{n_{j+1}} \exp(-d_{n_{j+1}}) \quad \text{as in (2.45) of [6]} \\ &< \infty \quad \text{in complete analogy with Lemma 8 of [6] and using (b).} \end{aligned}$$

This completes the convergence half of the proof.

Suppose $\sum_1^\infty (\lambda_n^2/n) \exp(-\lambda_n) = \infty$. Note that $\|I/\Gamma_n\|_{\xi_{n1}}^1 \geq n \xi_{n2}$, and Theorem 1(ii) shows that $P(n \xi_{n2} \geq \lambda_n \text{ i.o.}) = 1$. \square

REMARK. Now $\{n \xi_{n,i+1}/i : 1 \leq i \leq n\}$ is a reverse submartingale. This yields

$$P(\|I/\Gamma_n\|_{\xi_{n1}}^1 \geq \lambda) \leq \inf_{r>0} E(\exp(rn \xi_{n2}))/\exp(r\lambda) \leq 14\lambda^2 \exp(-\lambda)$$

for all $\lambda > 1$. This will only yield $P(\|I/\Gamma_n\|_{\xi_{n1}}^1 \geq \lambda_n \text{ i.o.}) = 0$ in Theorem 3 in case $\sum_1^\infty (\lambda_n^3/n) \exp(-\lambda_n) < \infty$.

PROOF OF THEOREM 1. (i) See Kiefer (1972). (ii) See Robbins and Siegmund (1971) for the case $k = 1$. See Frankel (1976) for a statement of this result when $k > 1$ and $c_n \nearrow \infty$; Frankel gives references to his 1972 thesis for a proof. It would appear that Frankel's technique is similar to that of Wichura (1973); using diffusion processes and speed measure, Wichura establishes some results very closely related to the present ones.

The authors' original version of this manuscript included a very long proof of Theorem 1(ii); it is available upon request. It uses only elementary techniques, and is a straightforward generalization of the proof of Robbins and Siegmund; the details are quite heavy. \square

3. The case of arbitrary df's. Suppose X_{n1}, \dots, X_{nn} are independent with completely arbitrary df's F_{n1}, \dots, F_{nn} on $(-\infty, \infty)$. Let $\bar{F}_n = n^{-1} \sum_1^n F_{ni}$ denote the average df, and let F_n denote the empirical df of the observations.

THEOREM 4. Let $n\lambda_n \nearrow$. Then $\sum_{n=1}^\infty (n\lambda_n)^{-1} < \infty$ implies $P(\|F_n/\bar{F}_n\| \geq \lambda_n \text{ i.o.}) = 0$.

PROOF. By Theorem 1.1.1 and Corollary 1.3.1 of van Zuylen (1976) we have

$$(4) \quad P(\|F_n/\bar{F}_n\| \geq \lambda) \leq 2\pi^2/3\lambda \quad \text{for all } \lambda > 1.$$

We can now just recopy the proof of Theorem 2, except that an appeal to (4) replaces the appeal to (1). \square

We did not generalize Theorem 3 to the present case. It is possible to obtain an exponential bound in place of the bound in van Zuylen's equation (1.1.4) by applying a binomial exponential bound to the probability $P(\sum_{i=1}^n z_i > n - j + 1)$ of his proof. However, the resulting bound is not as strong as (2); and so we omit the resulting weak generalization of Theorem 3 that we can prove in the present case.

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