LAWS OF THE ITERATED LOGARITHM FOR ORDER STATISTICS OF UNIFORM SPACINGS

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Let X_1, X_2, \cdots be a sequence of independent uniformly distributed random variables on [0, 1], and let K_n be the kth largest spacing induced by the order statistics of X_1, \cdots, X_{n-1} . We show that

 $\lim \sup (nK_n - \log n)/2 \log_2 n = 1/k$ almost surely,

and

 $\lim \inf(nK_n - \log n + \log_3 n) = c$ almost surely,

where $-\log 2 \le c \le 0$, and \log_i is the j times iterated logarithm.

1. Introduction. Consider a sequence X_1, X_2, \cdots of independent identically distributed random variables with a uniform distribution on [0, 1]. If $X_{(1)} < X_{(2)} < \cdots < X_{(n-1)}$ are the order statistics corresponding to X_1, \cdots, X_{n-1} , then the maximal uniform spacing (or, the maximal gap) M_n is defined by

$$M_n = \max_{1 \le i \le n} S_i$$

where $S_1 = X_{(1)}$, $S_i = X_{(i)} - X_{(i-1)}$ for 1 < i < n, and $S_n = 1 - X_{(n-1)}$. The S_i 's are called the *spacings*; see Pyke (1965).

Slud (1978) showed that $nM_n - \log n = O(\log_2 n)$ a.s.; we will refine Slud's result and show that

$$(1.1) \qquad \qquad \lim \sup(nM_n - \log n)/2 \log_2 n = 1 \quad \text{a.s.}$$

and that

(1.2)
$$\lim \inf nM_n - \log n + \log_3 n = c \quad \text{a.s.}$$

where $-\log 2 \le c \le 0$. Along the way, we will obtain a few large deviation results for M_n . In Section 2, we state without proof a few known results about the distribution and the weak convergence of M_n . In Sections 4 and 5, we will establish (1.1) and (1.2) for K_n , the kth largest spacing among S_1, \dots, S_n , when the constant "1" in (1.1) is replaced by 1/k.

2. Auxiliary results. It is well-known that (S_1, \dots, S_n) is uniformly distributed on the simplex $\{(x_1, \dots, x_n) \mid x_i \geq 0; \sum x_i = 1\}$, and that, therefore

$$P(S_1 > a_1; \dots; S_n > a_n) = (1 - \sum_{i=1}^n a_i)^{n-1}, \qquad \sum_{i=1}^n a_i < 1$$

= 0, otherwise,

where a_1, \dots, a_n are nonnegative numbers. From this, one can get Whitworth's formula (Whitworth (1897); see also Kendall and Moran (1963)):

$$\begin{split} P(M_n > x) &= P(\bigcup_{i=1}^n \left[S_i > x \right]) = \sum_i P(S_i > x) - \sum_{i < j} P(S_i > x; S_j > x) + \cdots \\ &= \sum_{k \ge 1; k x < 1} (-1)^{k+1} (1 - kx)^{n-1} \binom{n}{k}, \quad \text{all} \quad x > 0. \end{split}$$

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A very useful property of uniform spacings is the following.

LEMMA 2.1. If Y_1, \dots, Y_n are independent identically distributed exponential random variables, and if $T_n = \sum Y_i$, then (S_1, \dots, S_n) is distributed as $(Y_1/T_n, \dots, Y_n/T_n)$. In particular, M_n is distributed as L_n/T_n where $L_n = \max(Y_i)$.

For a proof of Lemma 2.1, see Pyke (1965).

LEMMA 2.2. (Sukhatme, 1937). If Y_1, \dots, Y_n are independent identically distributed exponential random variables with corresponding order statistics $Y_{(1)} < Y_{(2)} < \dots < Y_{(n)}$, then the following random variables are also independent and exponentially distributed:

$$nY_{(1)}, (n-1)(Y_{(2)}-Y_{(1)}), \dots, 2(Y_{(n-1)}-Y_{(n-2)}), Y_{(n)}-Y_{(n-1)}.$$

An immediate consequence of Lemma 2.2 is the following.

Lemma 2.3. M_n is distributed as

$$\sum_{i=1}^{n} (Y_i/i)/\sum_{i=1}^{n} Y_i$$

where Y_1, \dots, Y_n are independent exponentially distributed random variables.

The limit distribution of M_n was found by Levy (1939) and was rederived later by Darling (1952, 1953) and others.

LEMMA 2.4. For all
$$x \in R$$
, $P(nM_n < \log n + x) \to \exp(-\exp(-x))$ as $n \to \infty$.

LEMMA 2.5. $nM_n/\log n \to 1$ in probability as $n \to \infty$.

Note. If G_n is the distribution function of $nM_n - \log n$ and $G(x) = \exp(-\exp(-x))$, and if $a_n \log n \to \infty$ as $n \to \infty$, then

$$P(|nM_n/\log n - 1| > a_n) = G_n(-a_n \log n) + 1 - G_n(a_n \log n)$$

$$\leq 2\sup_x |G_n(x) - G(x)| + G(-a_n \log n) + 1 - G(a_n \log n) \to 0.$$

The distribution function $G(x) = \exp(-\exp(-x))$ has mean $\gamma = 0.5772157...$ (the Euler constant) and variance $\pi^2/6$; see Gnedenko (1943), Gumbel (1958), Barndorff-Nielsen (1963) and David (1970) for a closer analysis of its properties. A careful application of Lemma 2.3 also gives

Lemma 2.6.
$$E(nM_n - \log n) \rightarrow \gamma \text{ as } n \rightarrow \infty, \text{ and } Var(nM_n) \rightarrow \pi^2/6 \text{ as } n \rightarrow \infty.$$

3. Large deviation results. We will first derive exponential estimates for the probability in the tail of the gamma density. We recall here that the sum T_n of n independent exponentially distributed random variables has the gamma density $g_n(x) = x^{n-1}e^{-x}/(n-1)!$, $x \ge 0$.

LEMMA 3.1. For all x > 0,

$$P(T_n/n - 1 > x) \le \exp(-nx^2(1 - x)/2)$$

and

$$P(T_n/n - 1 < -x) \le \exp(-nx^2/2)$$
.

PROOF. Here and throughout the paper we will use these analytic inequalities, valid for all $x \ge 0$:

(3.1)
$$e^{x-x^2/2} \le 1 + x \le e^x \le 1 + x + x^2 e^x/2$$

$$(3.2) 1 - x \le e^{-x - x^2/2 - x^3/3} \le e^{-x - x^2/2} \le e^{-x} \le 1 - x + x^2/2.$$

Lemma 3.1 is now easily proved by Chernoff's classical technique (Chernoff, 1952). For any 0 < s < 1, we have $P(T_n/n - 1 > x) \le e^{-snx}E(e^{s(T_n-n)}) = e^{-sn(1+x)}(1-s)^{-n}$. This expression is minimal when 1 - s = 1/(1+x)(s = x/(1+x)), so that the said probability is not greater than $(e^{-x}(1+x))^n \le ((1-x+x^2/2)(1+x))^n = (1-x^2/2+x^3/2)^n \le e^{-nx^2(1-x)/2}$. Similarly, for all s > 0, $P(T_n/n - 1 < -x) \le e^{-snx}E(e^{-s(T_n-n)}) = e^{sn(1-x)}(1+s)^{-n} = (e^x(1-x))^n \le (e^{x-x-x^2/2})^n = e^{-nx^2/2}$ where we let s = x/(1-x) whenever x < 1. For $x \ge 1$, the result is trivially true.

LEMMA 3.2. Let $k \ge 1$ be a fixed integer, and let $a_n \to 0$ and $a_n \log n \to \infty$. If K_n is the k-th largest spacing among S_1, \dots, S_n , then

$$P(nK_n/\log n - 1 > a_n) \sim n^{-ka_n}/k!$$

and

$$P(nK_n/\log n - 1 \le -a_n) \sim n^{(k-1)a_n} \exp(-n^{a_n})/(k-1)!$$

PROOF. We will use the following fact about the tail of the binomial distribution. If B is a binomial random variable with parameters n and p, then $np \to 0$ implies $P(B \ge k) \sim P(B = k)$, and $np \to \infty$ implies $P(B < k) \sim P(B = k - 1)$ (Feller, 1957, page 140).

 K_n is distributed as L'_n/T_n where L'_n is the kth largest of n independent identically distributed random variables with exponential density and whose sum is T_n (Lemma 2.1). For arbitrary a, b > 0 we have

$$P(L'_n < (1-a-b)\log n) - P(T_n < n(1-b)) \le P(nK_n/\log n < 1-a)$$

(3.3)

$$\leq P(L'_n < (1-a+b) \log n) + P(T_n \geq n(1+b))$$

and

$$P(L'_n > (1 + a + b)\log n) - P(T_n > n(1 + b)) \le P(nK_n/\log n > 1 + a)$$

(3.4)

$$\leq P(L'_n > (1 + a - b)\log n) + P(T_n \leq n(1 - b)).$$

Let us take $a = a_n$ and $b = n^{-1/4}$. Lemma 3.2 follows if we can show the following things:

- (i) $P(L'_n < (1-a)\log n) \sim \exp(-n^a)n^{(k-1)a}/(k-1)!$;
- (ii) $P(L'_n > (1+a)\log n) \sim n^{-ka}/k!$;
- (iii) $P(|T_n n| > bn)/\min(P(L'_n < (1 a)\log n), P(L'_n > (1 + a)\log n)) \to 0;$
- (iv) $P(L'_n < (1-a-b)\log n) \sim P(L'_n < (1-a+b)\log n);$
- (v) $P(L'_n > (1 + a + b)\log n) \sim P(L'_n > (1 + a b)\log n)$.

Clearly, $P(L'_n < (1-a)\log n) = P(B < k)$ where B is binomial with parameters n and $p = \exp(-(1-a)\log n) = n^a/n$. Since $np \to \infty$, we have $P(B < k) \sim P(B = k-1) = \binom{n}{k-1} p^{k-1} (1-p)^{n-k+1} \sim (np)^{k-1} \exp(-np)/(k-1)! = n^{(k-1)a} \exp(-n^a)/(k-1)!$. Similarly, $P(L'_n > (1+a)\log n) = P(B \ge k)$ where now B is binomial with parameters n and $p = \exp(-(1+a)\log n) = 1/n^{1+a}$. Since $np \to 0$, we have $P(B \ge k) \sim P(B = k) \sim 1/n^{ka}k!$. This proves (i) and (ii). The same asymptotic results are valid if in (i) and (ii) we replace a by (a+b) or (a-b) on both sides. The ratio of the two terms of (v) (left divided by right) is $n^{-2kb} \sim 1$. The ratio of the two terms of (iv) is $n^{2(k-1)b} \exp(n^{(a-b)} - n^{(a+b)}) \sim 1$.

To prove (iii) we first use Lemma 3.1: $P(|T_n - n| > bn) \le 2 \exp(-nb^2/4)$ for n large enough. It remains to check that $n^{ka} \exp(-nb^2/4) \to 0$ and that $n^{(k-1)a} \exp(n^a - nb^2/4) \to 0$. This follows from $a \to 0$.

- **4. Outer Bounds.** In 1961 Barndorff-Nielsen (and independently Robbins and Siegmund (1970) and Deheuvels (1974)) established laws of the iterated logarithm for $Z_n = \min(X_1, \dots, X_n)$ where X_1, \dots, X_n is a sequence of independent uniform [0, 1] random variables. These results can be summarized as follows. Let a_n be positive and nonincreasing. Then,
 - (i) $Z_n < a_n$ i.o. (f.o.) when $\sum a_n = \infty$ ($\sum a_n < \infty$). See Geffroy (1958) for the first proof.
 - (ii) $Z_n > a_n$ i.o. (f.o.) when $\sum a_n \exp(na_n) = \infty$ ($\sum a_n \exp(-na_n) < \infty$) under the assumption that na_n is ultimately non-decreasing (Robbins and Siegmund, 1970). Barndorff-Nielsen's result uses the series $\sum \log_2 n(1 a_n)^n/n$ instead of $\sum a_n \exp(-na_n)$. For related work, see Frankel (1972) and Wichura (1973). For a short proof of the first order result: $Z_n > (1 + \varepsilon)\log_2 n/n$ i.o. (f.o.) when $\varepsilon = 0$ ($\varepsilon > 0$), see Kiefer (1970). For a survey, with proofs, see Galambos (1978).

In this section we derive sufficient conditions (of the summability type) for $nK_n > (1 + a_n)\log n$ finitely often a.s. and $nK_n < (1 - a_n)\log n$ finitely often a.s.

LEMMA 4.1. Let A_1, A_2, \cdots be a sequence of events with $P(A_n) \to 0$ as $n \to \infty$. If either $\sum P(A_n^c \cap A_{n+1}) < \infty$ or $\sum P(A_n \cap A_{n+1}^c) < \infty$, then $P(A_n$ f.o.) = 1.

PROOF. See Barndorff-Nielsen (1961).

THEOREM 4.1. Let $a_n \to 0$ and $a_n \log n \to \infty$ as $n \to \infty$ such that $(1 + a_n) \log n/n$ is ultimately nonincreasing. Then, $P(nK_n > (1 + a_n) \log n$ i.o.) = 0 when

PROOF. Let A_n be the event $nK_n > (1 + a_n)\log n$. By (2.1), $P(A_n) \to 0$ as $n \to \infty$. Then, for n large enough,

$$P(A_n \cap A_{n+1}^c) \le P(nK_n > (1+a_n)\log n)2k(1+a_{n+1})(\log(n+1)/(n+1))$$

= $2k(1+o(1))n^{-ka_n}k!^{-1}\log n/n$,

from which Theorem 4.1 follows after applying Lemma 4.1.

THEOREM 4.2. Let $a_n \to 0$ and $a_n \log n \to \infty$ as $n \to \infty$ such that $(1 - a_n) \log n/n$ is ultimately nonincreasing. Then, $P(nK_n < (1 - a_n) \log n \text{ i.o.}) = 0$ when

$$\sum_{n=1}^{\infty} (\log n/n) n^{ka_n} \exp(-n^{a_n}) < \infty.$$

PROOF. Let A_n be the event $nK_n < (1-a_n)\log n$. Once again, we will use Lemma 4.1. Obviously, $P(A_n) \sim n^{(k-1)a_n} \exp(-n^{a_n})/(k-1)! \to 0$ as $n \to \infty$. Also, if K'_n is the (k+1)st largest spacing among S_1, \dots, S_n , then for n large,

$$P(A_n^c \cap A_{n+1}) = P(A_n^c \cap A_{n+1} \cap [K'_n < (1 - a_{n+1})\log(n+1)/(n+1)])$$

$$\leq P(K'_n < (1 - a_n)\log n/n)2k \log n/n$$

$$= 2k(1 + o(1))n^{ka_n} \exp(-n^{a_n})k!^{-1} \log n/n.$$

REMARK 4.1. It follows trivially from Theorems 4.1 and 4.2 that $nK_n/\log n \to 1$ a.s. as $n \to \infty$. Of course, we have done too much work by invoking Lemma 3.2. For a short proof of $nM_n/\log n \to 1$ a.s., see Slud (1978) or Devroye (1979).

Remark 4.2. Condition (4.1) is satisfied if for some $\delta > 0$, $J \ge 2$, we have

$$a_n = (k \log n)^{-1} (\log_2 n + \sum_{j=2}^{J} \log_j n + \delta \log_J n).$$

In particular, it is satisfied if we take $a_n = (2 + \delta)\log_2 n/(k \log n)$, $\delta > 0$. Hence,

$$(4.3) \qquad \lim \sup(nK_n - \log n)/2 \log_2 n \le 1/k \text{ a.s.}$$

Remark 4.3. Condition (4.2) is satisfied if for some $J \ge 3$, $\delta > 0$, we have

$$a_n = (\log n)^{-1} (\log(2 \log_2 n + k \log_3 n + \sum_{j=3}^J \log_j n + \delta \log_J n)),$$

or when for some $\delta > 0$, $a_n = \log((2 + \delta)\log_2 n)/\log n$. Hence,

$$(4.4) \qquad \lim \inf(nK_n - \log n + \log_3 n) \ge -\log 2 \quad \text{a.s.},$$

independent of k. The influence of k on the lower outer bound is only in the second order term of the sequence a_n . In other words, whenever M_n is small, it is very likely that the second and third largest spacings are very close in magnitude to M_n .

5. Inner Bounds. In this section we will prove the following theorems:

THEOREM 5.1. $\limsup (nK_n - \log n)/2 \log_2 n = 1/k$ a.s.

THEOREM 5.2. $\liminf (nK_n - \log n + \log_3 n) = c$ a.s. for some $c \in [-\log 2, 0]$.

We will use the notation [.] for the integer part of a number. Furthermore, we will need two lemmas.

LEMMA 5.1. If
$$b_j = \exp(a\sqrt{j} \log j)$$
, where $a > 0$, then

$$(b_{j+1}-b_j)/b_j \sim a \log j/2\sqrt{j}$$
 as $j \to \infty$.

The same is true for $c_j = [b_j]$.

PROOF. In view of $(\sqrt{j+1} - \sqrt{j}) \sim \frac{1}{2} \sqrt{j}$ and $\log(1 + 1/j) \sim 1/j$, we have $(b_{j+1} - b_j)/b_j \sim a(\sqrt{j+1}\log(j+1) - \sqrt{j}\log j) \sim a\log j/2\sqrt{j}$.

LEMMA 5.2. If $b_j = \exp(j \log j)$, then

$$b_j/b_{j+1} \sim 1/ej$$
 as $j \to \infty$.

The same is true for $c_i = [b_i]$.

PROOF. By (3.1) and (3.2) we have
$$b_{j-1}/b_j = (j-1)^{-1} \exp(j \log(1-1/j)) \le 1/(e(j-1))$$
, and $b_{j-1}/b_j \ge (j-1)^{-1} \exp(-1-1/j) \ge (j-1)^{-1}e^{-1}(1-1/j) = 1/ej$.

PROOF OF THEOREM 5.1. In view of (4.3) we need only show that $nK_n - \log n > (2/k - \delta)\log_2 n$ i.o. almost surely, for all $\delta > 0$. We define the following sequences:

$$n_{j} = [\exp(\sqrt{j} \log j)],$$

$$t_{j} = [n_{j}(2/k - \delta/2)\log_{2}n_{j}/\log n_{j}],$$

$$a_{j} = (2/k - \delta)\log_{2}j/\log j,$$

$$d_{j} = (1 + a_{j})\log j/j,$$

$$d'_{j} = (1 + (3/k)\log_{2}n_{j}/\log n_{j})\log n_{j}/n_{j},$$

$$d''_{j} = (1 - \log(3 \log_{2}n_{j})/\log n_{j})\log n_{j}/n_{j},$$

Let us define the following events: A_N is the event that $K_{n_j} \in (d''_j, d'_j)$ for all $j \geq N$; B_N is the event that for some $j \geq N$, none of the random variables $X_{n_j}, \dots, X_{n_j+t_{j-1}}$ belong to the set C_j , where C_j is the union of k intervals of length d'_j each, with the restriction that the leftmost point of each interval coincides with the leftmost point of one of the k largest spacings.

We will see that $t_j + n_j < n_{j+1}$ for all j large enough, and that $d_j'' > d_{n_j+t_j}$ for all j large enough. Thus, $A_N \cap B_N \subseteq [K_{n_j+t_j} > d_{n_j+t_j}$ for some $j \ge N$]. The theorem now follows if we can show that $P(A_N^c) + P(B_N^c) \to 0$ as $N \to \infty$. From Theorems 4.1 and 4.2 we deduce that $P(A_N^c) \to 0$ as $N \to \infty$. Furthermore,

$$P(B_N^c) \le \prod_{j=N}^{\infty} (1 - (1 - kd_j')^{t_j}) \le \exp(-\sum_{j=N}^{\infty} (1 - kd_j')^{t_j}) = 0$$

whenever

$$\sum_{i=1}^{\infty} (1 - kd_i')^{t_i} = \infty.$$

Because $(1 - kd_j')^{t_j} \ge \exp(-d_j'kt_j - k^2d_j'^2t_j/2)$ and $d_j'^2t_j \to 0$, it suffices to check whether $\sum \exp(-kd_j't_j) = \infty$. We have $\exp(-kd_j't_j) \sim \exp(-(2 - \delta k/2)\log_2 n_j)$. $(1 + (3/k)\log_2 n_j/\log n_j) \sim \exp(-(2 - \delta k/2)\log_2 n_j) \sim (\sqrt{j} \log j)^{2-\delta k/2}$, which is not summable with respect to j.

We will now show that $n_j + t_j < n_{j+1}$ for all j large enough. Indeed, $n_{j+1} - n_j \sim n_j \log j/2\sqrt{j}$ (Lemma 5.1), while $t_j \sim (1/k - \delta/4)n_j/\sqrt{j}$.

Finally, let us establish that $d''_j > d_{n_j+t_j}$ for all j large enough. Clearly,

$$\begin{split} d_{n_j+t_j} &= \log(n_j + t_j)/(n_j + t_j) + (2/k - \delta)\log_2(n_j + t_j)/(n_j + t_j) \\ &< \log n_j/(n_j + t_j) + t_j/n_j^2 + (2/k - \delta)\log_2 n_j/n_j \\ &< (\log n_j/n_j)(1 - (1 + o(1))t_j/n_j) + o(1)/n_j + (2/k - \delta)\log_2 n_j/n_j \\ &< \log n_j/n_j - ((2/k - \delta/2)(1 + o(1))\log_2 n_j - (2/k - \delta)\log_2 n_j)/n_j \\ &= \log n_j/n_j - (\delta/2)(1 + o(1))\log_2 n_j/n_j. \end{split}$$

Also, $d_j'' = \log n_j/n_j - \log(3\log_2 n_j)/n_j > d_{n_j+t_j}$ for all j large enough.

PROOF OF THEOREM 5.2. We will show that for all $\delta > 0$, the inequality $nK_n < \log n - \log_3 n + \delta$ is satisfied i.o. almost surely, that is, a.s. $\liminf (nK_n - \log n + \log_3 n) \le 0$. This result together with (4.4) imply the statement of Theorem 5.2.

For given $\delta > 0$, define $n_j = [\exp(2j\log j)]$, $d_j = (\log n_j - \log_3 n_j + \delta)/n_j$, $t_j = n_j - n_{j-1}$ and $a_j = (\log_3 n_j - \delta/2)/\log n_j$. Let further N_j be the kth largest gap defined by $X_{n_{j-1}}, \dots, X_{n_j-1}$ on [0, 1]. Obviously, $N_j < d_j$ i.o. implies that $K_{n_j} < d_j$ i.o. Since the N_j 's are independent, $N_j < d_j$ i.o. almost surely whenever $\sum P(N_j < d_j) = \infty$. By Lemma 3.2,

$$P(N_i < (\log t_i/t_i)(1-a_i)) \sim t_i^{(k-1)a_i} \exp(-t_i^{a_i})/(k-1)!$$

because $a_j \log t_j \to \infty$. Also, $\exp(-t_j^{a_j}) \ge \exp(-n_j^{a_j}) = \exp(-c' \log_2 n_j) \sim (2j \log j)^{-c'}$ for some c' < 1. Thus, $\sum P(N_j < d_j) = \infty$ if $d_j > (\log t_j/t_j)(1 - a_j)$ for all j large enough. Now,

$$d_j t_j / \log n_j \ge (t_j / n_j) (1 - (\log_3 n_j - \delta) / \log n_j) = (1 - O(j^{-2})) (1 - (\log_3 n_j - \delta) / \log n_j)$$

which is greater than $1 - a_i = 1 - (\log_3 n_i - \delta/2)/\log n_j$ for all j large enough.

6. Applications.

EXAMPLE 6.1. Random covers. Assume that we try to cover [0, 1] by intervals of length \mathcal{L}_n centered at X_1, \dots, X_{n-1} (where the X_i 's are independent and uniformly distributed on [0, 1]). Let A_n be the event [[0, 1] is entirely covered]. Then, if $n\mathcal{L}_n = \log n - \log_3 n + \delta$,

$$P(A_n \text{ i.o.}) = \begin{bmatrix} 1, & \text{if } \delta > 0 \\ 0, & \text{if } \delta + \log 2 < 0. \end{bmatrix}$$

If $n\ell_n = \log n + (2 + \delta)\log_2 n$, we have

$$P(A_n^c \text{ i.o.}) = \begin{bmatrix} 1, & \text{if } \delta < 0 \\ 0, & \text{if } \delta > 0. \end{bmatrix}$$

It is perhaps interesting to compare this result with Shepp's covering theorem (1972): let $\ell_1 \geq \ell_2 \geq \cdots \geq 0$ be the lengths of arcs thrown at random on the circle with unit circumference ($\ell_1 < 1$). Then the circle is covered almost surely if and only if

$$\sum_{n=1}^{\infty} n^{-2} \exp(\ell_1 + \cdots + \ell_n) = \infty.$$

If $\ell_n = (1/n)(1 - (1 + \delta)/\log n)$, then this condition is satisfied when $\delta \leq 0$ and is violated when $\delta > 0$.

EXAMPLE 6.2. Uniform convergence of nonparametric estimates. Assume that f is a uniformly continuous function on [0, 1], and that f is estimated by

$$f_n(x) = \sum_{i=1}^n f(X_i) K\left(\frac{X_i - x}{\ell_n}\right) / \sum_{i=1}^n K\left(\frac{X_i - x}{\ell_n}\right)$$

where X_1, \dots, X_n are independent identically distributed uniform [0, 1] random variables, and K(u) is a nonincreasing nonnegative function of u when u > 0, and a nondecreasing nonnegative function of u when u < 0. Let the support of K be a compact set [a, b] (clearly, $a \le 0 \le b$) with a < b.

It is clear that $\sup_x |f_n(x) - f(x)| \to 0$ a.s. for all uniformly continuous f if and only if $M_n > (b-a)\ell_n$ f.o. almost surely. Now, if we take $n(b-a)\ell_n = \log n + (2+\delta)\log_n n$, then

$$\sup_{x} |f_n(x) - f(x)| \to 0 \text{ a.s.}$$
 as $n \to \infty$

for all uniformly continuous f if $\delta > 0$; the statement is false if $\delta < 0$.

EXAMPLE 6.3. Estimating the minimum of a density. Let f be a uniformly continuous density on [0, 1], and let z be the unique point with the property that $f(z) = \min_x f(x)$. Assume that X_1, X_2, \cdots is an independent sample from f, and that z is estimated by Z_n , the midpoint of the largest interval created by X_1, \cdots, X_n . From $nM_n/\log n \to 1$ a.s. for uniform distributions, one can show that $Z_n \to z$ a.s. as $n \to \infty$. For the study of laws of the iterated logarithm of M_n in the non-uniform case, additional assumptions about the rate of increase of f near z seem necessary. Notice also that if the maximum of f were estimated by the midpoint of the smallest interval, then one would not obtain almost sure convergence as in the case of Z_n .

EXAMPLE 6.4. Rate of convergence of nearest neighbor estimates. Let f and X_1, X_2, \cdots be as in Example 6.2, but consider now the nearest neighbor estimate $f_n(x) = f(X_n^N(x))$ where $X_n^N(x)$ is the nearest neighbor to x among X_1, \dots, X_n . If f is Lipschitz with constant C, then $\sup_x |f_n(x) - f(x)| \le \max(CM_{n+1}/2; CX_{(1)}; C(1-X_{(n)}))$ where $X_{(1)} < \dots < X_{(n)}$ are the order statistics obtained from X_1, \dots, X_n . From the properties of $X_{(1)}$ and M_n (Theorem 4.1) we have the following rate of convergence result:

$$\sup_{x} |f_n(x) - f(x)| (2n/C \log n) > 1 + a_n$$
 f.o. a.s.

when $a_n \log n \to \infty$, $(1 + a_n) \log n/n$ is ultimately nonincreasing and $\sum_{n=1}^{\infty} \log n/n^{1+a_n} < \infty$. On the other hand, if f(x) = Cx, then the supremum is equal to the maximum of the three given terms, so that we may conclude, by Theorem 5.1, that there exists a Lipschitz function with constant C such that

 $\sup_{x} |f_n(x) - f(x)| 2n/(C \log n) > 1 + (2 - \delta) \log_2 n/\log n$ i.o. a.s. for all $\delta > 0$.

In other words, in the class Lip(C), we have

(6.1)
$$\lim \sup ((2n/C)\sup_x |f_n(x) - f(x)| - \log n)/\log_2 n \le 2 \quad \text{a.s.}$$

but there always exists an f in Lip(C) for which (6.1) is valid with equality.

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