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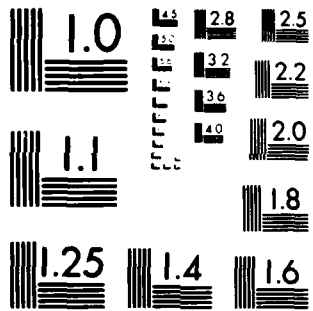
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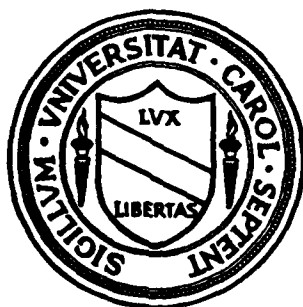
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AN ITERATED LOGARITHM LAW RESULT
FOR EXTREME VALUES FROM GAUSSIAN SEQUENCES

William P. McCormick

TECHNICAL REPORT #29

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ITEM #20, CONTINUED:

Let $\{X_n, n \geq 1\}$ be a stationary Gaussian sequence with $EX_1=0$, $EX_1^2=1$ and

$r_n = EX_1 X_{n+1}$. Let $Z_n^{(i)}$ denote the i th maximum of X_1, \dots, X_n and $a_n = (\ln \ln n)(2 \ln n)^{-1/2}$,

$b_n = (2 \ln n)^{1/2} - (\ln(4\pi \ln n))/(2(2 \ln n)^{1/2})$. Then assuming $r_n (\ln n)^2 = o(1)$ the set of

almost sure limit points of the vectors $((Z_n^{(1)} - b_n) a_n^{-1}, (Z_n^{(2)} - b_n) a_n^{-1}, \dots, (Z_n^{(\ell)} - b_n) a_n^{-1})$

is determined. The number of components $\ell = \ell(n) \rightarrow \infty$ as $n \rightarrow \infty$. This extends a result of

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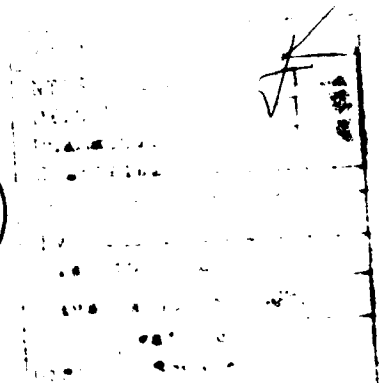
AN ITERATED LOGARITHM LAW RESULT
FOR EXTREME VALUES FROM GAUSSIAN SEQUENCES

William P. McCormick

Abstract

Let $\{X_n, n \geq 1\}$ be a stationary Gaussian sequence with $EX_1=0$, $EX_1^2=1$ and $r_n = EX_1 X_{n+1}$. Let $Z_n^{(i)}$ denote the i th maximum of X_1, \dots, X_n and $a_n = (\ln \ln n)(2 \ln n)^{-1/2}$, $b_n = (2 \ln n)^{1/2} - (\ln(4\pi \ln n))/(2(2 \ln n)^{1/2})$. Then assuming $r_n (\ln n)^2 = o(1)$ the set of almost sure limit points of the vectors $((Z_n^{(1)} - b_n) a_n^{-1}, (Z_n^{(2)} - b_n) a_n^{-1}, \dots, (Z_n^{(l)} - b_n) a_n^{-1})$ is determined. The number of components $l = l(n) \rightarrow \infty$ as $n \rightarrow \infty$. This extends a result of Hebbar.

Keywords: Iterated logarithm, Gaussian sequence, almost sure limit set.



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1. Introduction

Let $\{X_n, n \geq 1\}$ be a stationary Gaussian sequence with $EX_1=0$, $EX_1^2=1$ and $r_n = EX_1 X_{n+1}$. Let $Z_n^{(i)}$ denote the i th maximum of X_1, \dots, X_n that is $Z_n^{(i)}$ equals the $n-i+1$ order statistic. Set $a_n = \ln \ln n / \sqrt{2 \ln n}$ and $b_n = \sqrt{2 \ln n} - \ln(4\pi \ln n) / 2\sqrt{2 \ln n}$. In [1]

Hebbar considers the set of almost sure limit points of the sequence of vectors

$$\left\{ \left(\frac{Z_n^{(1)} - b_n}{a_n}, \frac{Z_n^{(2)} - b_n}{a_n}, \dots, \frac{Z_n^{(\ell)} - b_n}{a_n} \right), n \geq 1 \right\}.$$

He shows that under the assumption $r_n (\ln n)^{2+\epsilon} = o(1)$ for some $\epsilon > 0$ the above sequence has almost sure limit set equal to $\{(x_1, x_2, \dots, x_\ell) : 0 \leq x_\ell \leq \dots \leq x_1 \text{ and } \sum_1^\ell x_i \leq 1\}$. In the present paper we strengthen this result in two directions. We relax the condition on r_n to $r_n (\ln n)^2 = o(1)$ and further we allow the number ℓ of extreme values considered to grow to infinity with

n . Let $v_n^{(i)} = \frac{Z_n^{(i)} - b_n}{a_n}$. Then we consider the points in \mathbb{R}^∞ given by $(v_n^{(1)}, \dots,$

$v_n^{(\ell)}, 0, 0, \dots)$ where $\ell = \ell(n) \rightarrow \infty$ as $n \rightarrow \infty$. In \mathbb{R}^∞ we consider two modes of convergence--pointwise convergence and ℓ_1 -convergence. With $\ell(n)$ suitably bounded we show that the almost sure limit set in \mathbb{R}^∞ is given by

$$A = \{(x_1, x_2, \dots) : 0 \leq x_{i+1} \leq x_i, i=1, 2, \dots, \text{ and } \sum_1^\infty x_i \leq 1\}$$

2. Almost sure limit set

We consider two modes of convergence in \mathbb{R}^∞ , pointwise which is metrized by

$$d(\underline{x}, \underline{y}) = \sum_{n=1}^{\infty} \left(\frac{|x_n - y_n|}{1 + |x_n - y_n|} \right) 2^{-n} \text{ and } \ell_1.$$

Let us observe that a point \underline{x} is a limit point of a sequence \underline{x}_n with respect to pointwise convergence if and only if for each fixed ℓ , (x_1, \dots, x_ℓ) is a limit point of $(x_n^{(1)}, \dots, x_n^{(\ell)})$. Therefore with regard to pointwise convergence our extension of Hebbar's result is precisely to

weaken the mixing condition on r_n since finite dimensional results suffice to prove this case. Furthermore in this case we consider the almost sure limit points of the sequence $\{(v_n^{(1)}, v_n^{(2)}, \dots, v_n^{(n)}, 0, 0, \dots), n \geq 1\}$ that is we take $\ell(n) = n$.

However when we consider the random element $(v_n^{(1)}, \dots, v_n^{(\ell)}, 0, 0, \dots)$ as a point in ℓ_1 then we must take into account the rate at which $\ell(n)$ grows with n . In this case we prove an iterated logarithm law result with $\ell(n) = [\ell n_3 n]$. In the following we consider the ℓ_1 case only since the pointwise convergence case immediately follows.

The proof closely follows the method in [1] although additional detail is required to accommodate the infinite dimensional setting. However Lemma 6 in [1] receives an entirely different proof here that depends on an extension of a result of Mittal [2].

Remark: Let $\underline{x} = (x_1, x_2, \dots) \in A$ and assume $x_1 > 0$. Define the following sequences

$\lambda_k = [\ell n(\frac{1}{x_1} \ell n k)]$, $s_k = \sum_1^{\lambda_k} x_i$, $s = \sum_1^{\infty} x_i$ (assume $s < 1$) and $\alpha_k = [\exp(k \frac{1}{s^k})]$. Our program will be to show that the sequence $(v_{\alpha_k}^{(1)}, v_{\alpha_k}^{(2)}, \dots, v_{\alpha_k}^{(\lambda_k)}, 0, 0, \dots)$, $k \geq 1$ has \underline{x}

as a limit point almost surely. Then since $\ell_{\alpha_k} \leq \lambda_k$ and $\ell_{\alpha_k} \rightarrow \infty$ as $k \rightarrow \infty$ it fol-

lows easily that \underline{x} is a limit point of $(v_{\alpha_k}^{(1)}, v_{\alpha_k}^{(2)}, \dots, v_{\alpha_k}^{(\ell_{\alpha_k})}, 0, 0, \dots)$. In the lem-

mas which follow it will be assumed that $r_n(\ell n n)^2 = o(1)$ and that $s = \sum_1^{\infty} x_i < 1$.

Lemma 1. For any $\epsilon > 0$ we have

$$(2.1) \quad P\left\{ \sum_{i=1}^{\lambda_k} (v_{\alpha_k}^{(i)} - x_i) > \epsilon \quad \text{and} \quad v_{\alpha_k}^{(i)} > x_i, i=1, \dots, \lambda_k, \text{ i.o.} \right\} = 0.$$

Proof: To establish (2.1) it suffices to prove

$$(2.2) \quad P\left\{ \max_{1 \leq i \leq \lambda_k} (v_{\alpha_k}^{(i)} - x_i) > \epsilon / \lambda_k, \quad \min_{1 \leq i \leq \lambda_k} (v_{\alpha_k}^{(i)} - x_i) > 0, \text{ i.o.} \right\} = 0.$$

Further by Borel Cantelli to establish (2.2) it suffices to show

$$(2.3) \sum_k [\lambda_k \max_{1 \leq j \leq \lambda_k} P\{v_{\alpha_k}^{(j)} > x_j + \varepsilon/\lambda_k, v_{\alpha_k}^{(i)} > x_i, 1 \leq i \leq \lambda_k\}] < \infty.$$

Let $\{(y_{\alpha_k}^{(1)}, y_{\alpha_k}^{(2)}, \dots, y_{\alpha_k}^{(\lambda_k)}), k \geq 1\}$ be any triangular array with $y_{\alpha_k}^{(i)} \geq x_i, 1 \leq i \leq \lambda_k$

and $\max_{1 \leq i \leq \lambda_k} (y_{\alpha_k}^{(i)} - x_i) > \varepsilon/\lambda_k$. Let $\eta_{\alpha_k}^{(i)} = b_{\alpha_k} + y_{\alpha_k}^{(i)} a_{\alpha_k}$. Then we establish (2.3) by

showing that

$$(2.4) \sum_{k=1}^{\infty} \lambda_k P\{Z_{\alpha_k}^{(i)} \geq \eta_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k\} < \infty.$$

Let $Z_n^{*(i)}$ be the i th maximum of a sample of size n of i.i.d. standard normal random variables. Then in order to show (2.4) it suffices to show

$$(2.5) \sum_1^{\infty} \lambda_k P\{Z_{\alpha_k}^{*(i)} \geq \eta_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k\} < \infty \quad \text{and}$$

$$(2.6) \sum_1^{\infty} \lambda_k |P\{Z_{\alpha_k}^{*(i)} \geq \eta_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k\} - P\{Z_{\alpha_k}^{(i)} \geq \eta_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k\}| < \infty.$$

In considering (2.5) observe that

$$(2.7) \begin{aligned} & P\{Z_{\alpha_k}^{*(i)} \geq \eta_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k\} \\ &= P\{Z_{\alpha_k}^{*(1)} \geq \eta_{\alpha_k}^{(1)}\} \\ &\quad - \sum_{i=2}^{\lambda_k} P\{Z_{\alpha_k}^{*(1)} > \eta_{\alpha_k}^{(1)}, \dots, Z_{\alpha_k}^{*(i-1)} > \eta_{\alpha_k}^{(i-1)}, Z_{\alpha_k}^{*(i)} \leq \eta_{\alpha_k}^{(i)}\} \end{aligned}$$

Further it can be easily checked that

$$(2.8) P\{Z_{\alpha_k}^{*(1)} > \eta_{\alpha_k}^{(1)}\} = k \frac{y_{\alpha_k}^{(1)}}{s_k} + o\left(\frac{1}{\alpha_k}\right) \quad \text{and}$$

$$(2.9) \quad P\{Z_{\alpha_k}^{*(1)} > \eta_{\alpha_k}^{(1)}, \dots, Z_{\alpha_k}^{*(i-1)} > \eta_{\alpha_k}^{(i-1)}, Z_{\alpha_k}^{*(i)} \leq \eta_{\alpha_k}^{(i)}\}$$

$$= k \left[-\left(\frac{1}{s_k} \sum_1^{i-1} y_{\alpha_k}^{(t)}\right) - k \left[-\left(\frac{1}{s_k} \sum_1^i y_{\alpha_k}^{(t)}\right) + O\left(\frac{1}{\alpha_k}\right) - \frac{1}{s_k} \sum_1^{\lambda_k} y_{\alpha_k}^{(t)} + O\left(\frac{\lambda_k}{\alpha_k}\right) \right] \right]$$

Thus by (2.8) and (2.9) we obtain that (2.7) equals k

Thus since $\sum_1^{\lambda_k} y_{\alpha_k}^{(t)} > s_k + \epsilon \lambda_k^{-1}$ and $s_k \cdot s = \sum_1^{\infty} x_i \leq 1$, (2.5) is established.

Next we consider (2.6). First observe that

$$(2.10) \quad \begin{aligned} & |P\{Z_{\alpha_k}^{*(i)} \geq \eta_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k\} - P\{Z_{\alpha_k}^{(i)} \geq \eta_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k\}| \\ & \leq |P\{\eta_{\alpha_k}^{(1)} \leq Z_{\alpha_k}^{*(1)} \leq z_{\alpha_k}\} - P\{\eta_{\alpha_k}^{(1)} \leq Z_{\alpha_k}^{(1)} \leq z_{\alpha_k}\}| \\ & + \sum_{i=2}^{\lambda_k} |P\{\eta_{\alpha_k}^{(1)} \leq Z_{\alpha_k}^{*(1)} \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq Z_{\alpha_k}^{*(i-1)} \leq z_{\alpha_k}, Z_{\alpha_k}^{*(i)} \leq \eta_{\alpha_k}^{(i)}\} \\ & - P\{\eta_{\alpha_k}^{(1)} \leq Z_{\alpha_k}^{(1)} \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq Z_{\alpha_k}^{(i-1)} \leq z_{\alpha_k}, Z_{\alpha_k}^{(i)} \leq \eta_{\alpha_k}^{(i)}\}| \\ & + P\{Z_{\alpha_k}^{*(1)} > z_{\alpha_k}\} + P\{Z_{\alpha_k}^{(1)} > z_{\alpha_k}\} \end{aligned}$$

where $z_{\alpha_k} = 2\sqrt{\ell n \alpha_k}$.

It can be checked that

$$(2.11) \quad \begin{aligned} & P\{Z_{\alpha_k}^{*(1)} > z_{\alpha_k}\} = O\left(\frac{1}{\alpha_k}\right) \quad \text{and} \\ & |P\{Z_{\alpha_k}^{*(1)} > z_{\alpha_k}\} - P\{Z_{\alpha_k}^{(1)} > z_{\alpha_k}\}| \leq \alpha_k^{-2} \frac{1-\bar{r}_1}{1+\bar{r}_1} \end{aligned}$$

where $\bar{r}_x = \sup_{i>x} |r_i|$. Thus by (2.11) $\lambda_k (P\{Z_{\alpha_k}^{*(1)} > z_{\alpha_k}\} + P\{Z_{\alpha_k}^{(1)} > z_{\alpha_k}\})$ is sum-

mable on k.

Similarly it is easily checked that

$$(2.12) \quad |P\{Z_{\alpha_k}^{*(1)} \leq \eta_{\alpha_k}^{(1)}\} - P\{Z_{\alpha_k}^{(1)} \leq \eta_{\alpha_k}^{(1)}\}| \lambda_k \leq (\text{CONST.}) k^{-\frac{1+2y_{\alpha_k}^{(1)}}{s_k}} \lambda_k.$$

Since $s_k < 1$ and $y_{\alpha_k}^{(1)} \geq x_1 > 0$ for all k , the series in (2.12) is summable.

Now consider a term of the form

$$(2.13) \quad |P\{\eta_{\alpha_k}^{(1)} \leq Z_{\alpha_k}^{*(1)} \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq Z_{\alpha_k}^{*(i-1)} \leq z_{\alpha_k}, Z_{\alpha_k}^{*(i)} \leq \eta_{\alpha_k}^{(i)}\} \\ - P\{\eta_{\alpha_k}^{(1)} \leq Z_{\alpha_k}^{(1)} \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq Z_{\alpha_k}^{(i-1)} \leq z_{\alpha_k}, Z_{\alpha_k}^{(i)} \leq \eta_{\alpha_k}^{(i)}\}| \\ \leq S = \sum_{t_1, \dots, t_{i-1}} |P\{\eta_{\alpha_k}^{(1)} \leq X_{t_1}^* \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq X_{t_{i-1}}^* \leq z_{\alpha_k}, X_t^* \leq \eta_{\alpha_k}^{(i)}\}$$

for all $t \neq t_1, \dots, t_{i-1}, 1 \leq t \leq \alpha_k\}$

where $\{X_1^*, X_2^*, \dots\}$ denotes an i.i.d. sequence of standard normal random variables and where the summation is over all $1 \leq t_1, \dots, t_{i-1} \leq \alpha_k$ and $t_u \neq t_v$ of $u \neq v$.

Let $0 < \theta < 1$ be fixed and to be specified later. We write

$$S = S_0 + S_1 + \dots + S_{i-2}$$

where S_u denotes the sum over all t_1, \dots, t_{i-1} such that when the t 's are ordered $t_{(1)} < \dots < t_{(i-1)}$ there are exactly u indices h where $t_{(h+1)} - t_{(h)} < \alpha_k^\theta$.

Consider S_0 . We have

$$|P\{\eta_{\alpha_k}^{(1)} < X_{t_1}^* < z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} < X_{t_{i-1}}^* < z_{\alpha_k}, X_t^* \leq \eta_{\alpha_k}^{(i)}, t \neq t_1, \dots, t_{i-1} \text{ and } 1 \leq t \leq \alpha_k\} \\ - P\{\eta_{\alpha_k}^{(1)} < X_{t_1} < z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} < X_{t_{i-1}} < z_{\alpha_k}, X_t \leq \eta_{\alpha_k}^{(i)}, t \neq t_1, \dots, t_{i-1} \text{ and } 1 \leq t \leq \alpha_k\}| \\ \leq (\text{Const.})(T_0 + \sum_{0 \leq u \neq v \leq i-1} T_{u,v})$$

where $T_0 = \sum_{s,t}^{(0)} |r| \phi(\eta_{\alpha_k}^{(i)}, \eta_{\alpha_k}^{(i)}, |r|)$

$$\cdot P\{\eta_{\alpha_k}^{(1)} < X_{t_1} < z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} < X_{t_{i-1}} < z_{\alpha_k} | X_s = \eta_{\alpha_k}^{(i)}, X_t = \eta_{\alpha_k}^{(i)}\}$$

where $r = r_{s,t}$ and $\sum^{(0)}$ is summation over all $s \neq t$ and $s, t \neq t_1, \dots, t_{i-1}$, $1 \leq s, t \leq \alpha_k$ and where $\phi(\cdot, \cdot, r)$ denotes the bivariate normal density with zero means, unit variances and correlation r . Further for $v > 0$,

$$T_{0,v} = \sum_s^{(0,v)} |r| \phi(\eta_{\alpha_k}^{(i)}, \eta_{\alpha_k}^{(v)}, |r|)$$

$$\cdot P\{\eta_{\alpha_k}^{(1)} < X_{t_1} < z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(v-1)} < X_{t_{v-1}} < z_{\alpha_k}, \eta_{\alpha_k}^{(v+1)} < X_{t_{v+1}} < z_{\alpha_k},$$

$$\dots \eta_{\alpha_k}^{(i-1)} < X_{t_{i-1}} < z_{\alpha_k} | X_{t_v} = \eta_{\alpha_k}^{(v)}, X_s = \eta_{\alpha_k}^{(i)}\}$$

where the sum is over $s \neq t_1, \dots, t_{i-1}$ and $1 \leq s \leq \alpha_k$ and $r = r_{st_v}$.

$T_{u,0}$ is defined in exactly the same way and finally for $u, v > 0$

$$T_{u,v} = |r| \phi(\eta_{\alpha_k}^{(u)}, \eta_{\alpha_k}^{(v)}, |r|)$$

$$P\{\eta_{\alpha_k}^{(1)} < X_{t_1} < z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(u-1)} < X_{t_{u-1}} < z_{\alpha_k}, \eta_{\alpha_k}^{(u+1)} < X_{t_{u+1}} < z_{\alpha_k},$$

$$\dots \eta_{\alpha_k}^{(v-1)} < X_{t_{v-1}} < z_{\alpha_k}, \eta_{\alpha_k}^{(v+1)} < X_{t_{v+1}} < z_{\alpha_k}, \dots,$$

$$\eta_{\alpha_k}^{(i-1)} \leq X_{t_{i-1}} \leq z_{\alpha_k} | X_{t_u} = \eta_{\alpha_k}^{(u)}, X_{t_v} = \eta_{\alpha_k}^{(v)}\}.$$

We will give details only for the sum T_0 since the other sums are handled in the same way. For T_0 first consider the case when

$$(2.14) \quad \min\{|s-t_u|, |t-t_u|, u=1, \dots, i-1\} > \alpha_k^\theta$$

In evaluating T_0 we need to evaluate

$$(2.15) \quad P\{\eta_{\alpha_k}^{(1)} \leq X_{t_1} \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq X_{t_{i-1}} \leq z_{\alpha_k} | X_s = \eta_{\alpha_k}^{(i)}, X_t = \eta_{\alpha_k}^{(i)}\}$$

when

$$(2.16) \quad t_{(h+1)} - t_{(h)} > \alpha_k^\theta, \quad h=1, \dots, i-2$$

and (2.14) hold. Now subject to (2.14) we have that

$$(2.17) \quad E(X_{t_i} | X_s = \eta_{\alpha_k}^{(i)}, X_t = \eta_{\alpha_k}^{(i)}) = O((\ell n \alpha_k)^{-3/2}) \quad \text{and}$$

$$\text{CORR}(X_{t_u}, X_{t_v} | X_s, X_t) = r_{t_u, t_v} + O((\ell n \alpha_k)^{-4})$$

Therefore by (2.17) the probability in (2.15) is at most

$$(2.18) \quad P\{\eta_{\alpha_k}^{(u)} - c(\ell n \alpha_k)^{-3/2} \leq X_{t_u} \leq z_{\alpha_k} + c(\ell n \alpha_k)^{-3/2}, u=1, \dots, i-1\}$$

for some constant c not depending on k . Conditioning on X_{t_1} yields that (2.18) is at most

$$(1 - \Phi(b_{\alpha_k} (1 - c\bar{r}))) P\{b_{\alpha_k} (1 - c\bar{r})^2 \leq X_{t_u} \leq z_{\alpha_k} (1 - c\bar{r})^2, u=2, \dots, i-1\}$$

where $\bar{r} = \bar{r}_{\alpha_k}^\theta$.

Iterating the procedure yields that (2.18) is at most

$$(2.19) \quad \prod_{u=1}^{i-1} [1 - \Phi(b_{\alpha_k} (1 - c\bar{r})^u)]$$

Finally since $i \leq \lambda_k = \lfloor \ell n (\frac{\ell n k}{x_1}) \rfloor$ and $(1 - c\bar{r})^u \geq 1 - 2\lambda_k c\bar{r}$, (2.19) is at most

$$(2.20) \quad [1 - \Phi(b_{\alpha_k} - c \frac{\ell n k}{(\ell n \alpha_k)^{3/2}})]^{(i-1)}. \quad c \text{ is some constant.}$$

In the same way it can be checked that if for some u_0 , $|s - t_{u_0}| < \alpha_k^\theta$ but $|t - t_u| > \alpha_k^\theta$, $u=1, \dots, i-1$ or the same case with s and t interchanged then (2.15) is at most

$$(2.21) \quad (1 - \Phi(\gamma b_{\alpha_k})) [1 - \Phi(b_{\alpha_k} - c \frac{\ell n k}{(\ell n \alpha_k)^{3/2}})]^{(i-2)}$$

where $\gamma > 0$.

And finally if both $|s-t_{u_0}| < \alpha_k^\theta$ and $|t-t_{v_0}| < \alpha_k^\theta$ for some u_0 and v_0 then (2.15)

is at most

$$(2.22) \quad (1-\Phi(\gamma b_{\alpha_k}))^2 [1-\Phi(b_{\alpha_k} - c \frac{\ln k}{(\ln \alpha_k)^{3/2}})]^{(i-3)}$$

Thus we have that provided (2.16) holds

$$(2.23) \quad T_0 \leq \frac{(\text{CONST.})}{(\ln \alpha_k)^{(2y_{\alpha_k}^{(i)}+1)}} [1-\Phi(b_{\alpha_k} - c \frac{\ln k}{(\ln \alpha_k)^{3/2}})]^{(i-1)} = \frac{(\text{CONST.})}{(\ln \alpha_k)^{(2y_{\alpha_k}^{(i)}+1)}} \frac{1}{\alpha_k^{(i-1)}}$$

Similarly if (2.16) holds we find

$$(i) \quad T_{0,v} \leq \frac{(\text{CONST.})}{(y_{\alpha_k}^{(i)} + y_{\alpha_k}^{(v)} + 1)} \frac{1}{\alpha_k^{(i-1)}}$$

$$(2.24) \quad (ii) \quad T_{u,0} \leq \frac{(\text{CONST.})}{(y_{\alpha_k}^{(i)} + y_{\alpha_k}^{(u)} + 1)} \frac{1}{\alpha_k^{(i-1)}}$$

$$(iii) \quad T_{u,v} \leq \frac{(\text{CONST.})}{(y_{\alpha_k}^{(u)} + y_{\alpha_k}^{(v)} + 1)} \frac{1}{\alpha_k^{(i-1)}}$$

Thus by (2.23) and (2.24)

$$(2.25) \quad S_0 \leq \left[\frac{(\text{CONST.})}{\alpha_k^{(i-1)}} \frac{(\ln_2 k)^2}{(1+2x_i)/s_k} \right] (\alpha_k^{(i-1)}) \leq \frac{1}{k^{1+e}}$$

for some $e > 0$.

Next consider S_h . For simplicity let us consider a summand in (2.13) when $t_1 < t_2 < \dots < t_{i-1}$ and

$$(2.26) \quad 0 < t_2 - t_1, t_3 - t_2, \dots, t_{n+1} - t_n \leq \alpha_k^\theta \quad \text{and} \quad t_{u+1} - t_u > \alpha_k^\theta \quad u=h+1, \dots, i-1.$$

Then

$$|P\{\eta_{\alpha_k}^{(1)} \leq X_{t_1}^* \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq X_{t_{i-1}}^* \leq z_{\alpha_k}, X_t \leq \eta_{\alpha_k}^{(i)}\}$$

for all $t \neq t_1, \dots, t_{i-1}, 1 \leq t \leq \alpha_k\}$

$$- P\{\eta_{\alpha_k}^{(1)} \leq X_{t_1} \leq z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} \leq X_{t_{i-1}} \leq z_{\alpha_k}, X_t \leq \eta_{\alpha_k}^{(i)}\}$$

for all $t \neq t_1, \dots, t_{i-1}, 1 \leq t \leq \alpha_k\}$

$$\leq T_0 + \sum_{0 \leq u \neq v \leq i-1} T_{u,v}$$

where the $T_{u,v}$ have the same meaning as before except now condition (2.26) holds.

Consider T_0 . We need to evaluate

$$P\{\eta_{\alpha_k}^{(1)} < X_{t_1} < z_{\alpha_k}, \dots, \eta_{\alpha_k}^{(i-1)} < X_{t_{i-1}} < z_{\alpha_k} \mid X_s = \eta_{\alpha_k}^{(i)}, X_t = \eta_{\alpha_k}^{(i)}\}$$

Suppose that (2.14) holds. Then the above conditional probability is at most the expression given at (2.18). Therefore let us consider

$$(2.27) \quad P\{\eta_{\alpha_k}^{(u)} \leq X_{t_u} \leq z_{\alpha_k}, u=1, \dots, i-1\}$$

subject to condition (2.26).

Let $K = K(\alpha_k) = [\exp(\sqrt{2n\alpha_k})]$. Suppose that in addition to (2.26) we have that

$$(2.28) \quad t_2 - t_1, \dots, t_{m+1} - t_m \leq K, \quad \text{and} \quad K < t_{m+2} - t_{m+1}, \dots, t_{i-1} - t_{i-2}.$$

Then given (2.26) and (2.28), (2.27) equals at most

$$\prod_{u=0}^m (1 - \Phi(b_{\alpha_k} (1 - c\bar{r}_1)^u)) \cdot \prod_{u=m+1}^h (1 - \Phi(b_{\alpha_k} (1 - c\bar{r}_K)^u)) \cdot \prod_{u=h+1}^{i-2} (1 - \Phi(b_{\alpha_k} (1 - c\bar{r}_\theta)^u))$$

$$\leq (\text{CONST.}) \alpha_k^{-Q}$$

where $Q = 1 + \delta_1 - \delta_1^{m+1} + (h-m)\delta_k^{m+1} + (i-2-h)$ and $\delta_1 = 1 - c\bar{r}_1$, $\delta_k = 1 - c\bar{r}_k$
 where c is some constant and without loss of generality we can assume $c\bar{r}_1 < 1$
 because if necessary we can work with the sequence $\{X_{mn}, n \geq 1\}$ where m is some
 fixed integer.

Then as before we find that

$$(i) \quad T_0 \leq \frac{(\text{CONST.})}{(\ln \alpha_k)} \frac{1}{\alpha_k^{(2y_{\alpha_k}^{(i)} + 1)Q}}$$

$$(2.29) \quad (ii) \quad T_{u,0} \leq \frac{(\text{CONST.})}{(\ln \alpha_k)} \frac{1}{\alpha_k^{(y_{\alpha_k}^{(u)} + y_{\alpha_k}^{(i)} + 1)Q}}$$

and

$$(iii) \quad T_{u,v} \leq \frac{(\text{CONST.})}{(\ln \alpha_k)} \frac{1}{\alpha_k^{(y_{\alpha_k}^{(u)} + y_{\alpha_k}^{(v)} + 1)Q}}$$

Thus by the inequalities in (2.29) we have that

$$S_h \leq \sum_{m=0}^h \left(\frac{(\text{CONST.})}{\alpha_k^Q} \right) \left(\frac{(\ln_2 k)^2}{1 + 2x_i} \right) \left(\alpha_k^{[i-h-1+\theta(h-m)]_k^m} \right) \frac{1}{k^{S_k}}$$

from which it can be easily checked that

$$(2.30) \quad S_n \leq \alpha_k^{-f} \quad \text{for some } f > 0 \text{ not depending on } h.$$

Finally from (2.25) and (2.30) we have that

$$S \leq \frac{(\text{CONST.})}{k^{1+e}} \quad k \geq 1 \text{ for some } e > 0.$$

Hence (2.6) holds completing the proof of lemma 1.

Lemma 2. Let $c_n^{(i)} = b_n + x_i a_n$. Then

$$(2.31) \quad P\{Z_{\alpha_k}^{(i)} > c_{\alpha_k}^{(i)}, i=1, \dots, \lambda_k \text{ i.o.}\} = 1.$$

Proof: Let $\beta_k = [1/2 \alpha_k]$ and let $\tilde{Z}_{\alpha_k}^{(i)}$ be the i th maximum of the random variables $X_{\alpha_k - \beta_k + 1}, \dots, X_{\alpha_k}$. Let z_{α_k} be as in Lemma 1 and define $I_k = \prod_{i=1}^{\lambda_k} I_{[c_{\alpha_k}^{(i)} < \tilde{Z}_{\alpha_k}^{(i)} < z_{\alpha_k}]}$.

Let $J_n = \sum_{k=[n^a]}^n I_k$ where $0 < a < 1$ is a fixed real number. Then to show (2.31) it suffices to show

$$(i) \quad EJ_n \rightarrow \infty \text{ as } n \rightarrow \infty \text{ and}$$

(2.32)

$$(ii) \quad J_n/EJ_n \xrightarrow{P} 1 \text{ as } n \rightarrow \infty.$$

The proof of (2.32) follows the method of proof of Lemma 3 in [1] with changes similar to those in our Lemma 1. Therefore the details of this proof will be omitted.

Remark: The sequence $(v_{\alpha_k}^{(1)}, \dots, v_{\alpha_k}^{(\lambda_k)}, 0, 0, \dots)$ has \underline{x} as an almost sure limit point.

To see this let $N_n = \{\omega: \sum_1^{\lambda_k} (v_{\alpha_k}^{(i)} - x_i) > 1/n, v_{\alpha_k}^{(i)} > x_i, i=1, \dots, \lambda_k \text{ i.o.}\}$ and

$N = \bigcup_{n=1}^{\infty} N_n$. Let $A = \{\omega: v_{\alpha_k}^{(i)} > x_i, i=1, \dots, \lambda_k \text{ i.o.}\}$ and $\wedge = A \cap N^c$. It is easy

to check that if $\omega \in \wedge$, then $(v_{\alpha_k}^{(1)}(\omega), \dots, v_{\alpha_k}^{(\lambda_k)}(\omega), 0, 0, \dots)$ has \underline{x} as a limit point

and by Lemmas 1 and 2 $P(\wedge) = 1$. Therefore by the Remark preceding Lemma 1, we have that \underline{x} is an almost sure limit point of $(v_{\alpha_k}^{(1)}, \dots, v_{\alpha_k}^{(\lambda_k)}, 0, 0, \dots)$.

Lemma 3. Let $\underline{x} = (x_1, x_2, \dots)$ be any point in \mathbb{R}^{∞} with $0 \leq x_{i+1} \leq x_i, i=1, 2, \dots$ and $\sum_1^{\infty} x_i > 1$. Then \underline{x} cannot be an a.s. limit point of the sequence $(v_n^{(1)}, \dots, v_n^{(\ell_n)}, 0, 0, \dots)$

Proof: Let m be such that $s_m = \sum_1^m x_i > 1$, and $x_m > 0$. Let $z_i = (\frac{s_m+1}{2s_m})x_i$, $i=1, \dots, m$. Then since $\sum_1^m z_i > 1$, it follows as in [1] that $P\{v_n^{(i)} > z_i, i=1, \dots, m \text{ i.o.}\} = 0$. Let $N = \{\omega: v_n^{(1)} > z_1, \dots, v_n^{(m)} > z_m, \text{ i.o.}\}$. Then if $\omega \in N^c$, \underline{x} cannot be a limit point of $(v_n^{(1)}, \dots, v_n^{(\ell_n)}, 0, 0, \dots)$ because for all n sufficiently large

$$\sum_1^{\ell(n)} |v_n^{(i)} - x_i| \geq \min_{1 \leq i \leq m} (x_i - z_i) = (\frac{s_m+1}{2s_m})x_m.$$

A useful uniform bound on the tail probabilities of the normalized maxima for a Gaussian sequence is provided by Lemma 1 in [2]. We state a version of this result which is suited to our problem.

Lemma 4. Let $c_n = \sqrt{2\ell n}$. Let $\{X_{k,n}\}$, $k=1, \dots, n$, $n=1, 2, \dots$ be a triangular array of standard normal random variables. Then setting $r_n(i,j) = EX_{i,n}X_{j,n}$,

$M_n = \max_{1 \leq k \leq n} X_{k,n}$ and $\delta_n(x) = \sup_{|i-j| \geq x} |r_n(i,j)|$ we have

$$e^{tA^2} P\{c_n(M_n - b_n) \leq -A\} = o(1) \text{ as } A \rightarrow \infty$$

uniformly in n for all t in a neighborhood of zero provided

$$(i) \quad \overline{\lim}_{n \rightarrow \infty} \delta_n(1) < 1$$

$$(ii) \quad \delta_n(n^\alpha) \ell n = o(1) \text{ for some fixed } 0 < \alpha < 1.$$

Lemma 5. For any fixed positive integer ℓ and $\epsilon > 0$, $P\{Z_n^{(\ell)} < b_n - \epsilon a_n, \text{ i.o.}\} = 0$.

It is easily checked that it is sufficient to show $P\{Z_{n_k}^{(\ell)} < b_{n_{k+1}} - \epsilon a_{n_{k+1}}, \text{ i.o.}\} = 0$.

Also since for k sufficiently large $b_{n_{k+1}} - \epsilon a_{n_{k+1}} < b_{n_k} - \epsilon/2 a_{n_k}$ it is enough to show

$$(2.33) \quad P\{Z_{n_k}^{(\ell)} < b_{n_k} - \epsilon a_{n_k}, \text{ i.o.}\} = 0.$$

Observe that

$$(2.34) \quad \begin{aligned} & P\{Z_n^{(\ell)} < b_n - \varepsilon a_n\} \leq P\{Z_n^{(1)} > b_n + 2a_n\} \\ & + \sum_{i=2}^{\ell} P\{Z_n^{(i)} < b_n - \varepsilon a_n < Z_n^{(i-1)} < Z_n^{(1)} < b_n + 2a_n\} \end{aligned}$$

Now

$$(2.35) \quad P\{Z_n^{(1)} > b_n + 2a_n\} \leq n(1 - \Phi(b_n + 2a_n)) = \frac{1}{(\ell n n)^2}$$

Further we have that

$$(2.36) \quad \begin{aligned} & P\{Z_n^{(i)} < b_n - \varepsilon a_n < Z_n^{(i-1)} < Z_n^{(1)} < b_n + 2a_n\} \\ & = \sum_{t_1, \dots, t_{i-1}} P\{X_j \leq b_n - \varepsilon a_n, j \neq t_1, \dots, t_{i-1}, 1 \leq j \leq n\} \end{aligned}$$

$$\text{and } b_n - \varepsilon a_n < X_{t_u} < b_n + 2a_n, u=1, \dots, i-1\}$$

For a fixed $0 < \theta < 1$ and n sufficiently large

$$P\{X_j \leq b_n - \varepsilon a_n, j \neq t_1, \dots, t_{i-1}, 1 \leq j \leq n\}$$

and

$$(2.37) \quad \begin{aligned} & b_n - \varepsilon a_n < X_{t_u} < b_n + 2a_n, u=1, \dots, i-1\} \\ & \leq \int_{b_n - \varepsilon a_n}^{b_n + 2a_n} \dots \int_{b_n - \varepsilon a_n}^{b_n + 2a_n} P\{X_j \leq b_n - \varepsilon a_n, |j - t_u| > n^\theta, \\ & u = 1, \dots, i-1, 1 \leq j \leq n | X_{t_u} = x_u, u = 1, \dots, i-1\} \end{aligned}$$

$$\begin{aligned} & dP\{X_{t_1} \in dx_1, \dots, X_{t_{i-1}} \in dx_{i-1}\} \\ & \leq P\{\tilde{X}_j \leq b_n - \varepsilon/2 a_n, |j - t_u| > n^\theta, u=1, \dots, i-1, 1 \leq j \leq n\} \end{aligned}$$

$$P\{b_n - \varepsilon a_n \leq X_{t_u} \leq b_n + 2a_n, u=1, \dots, i-1\}$$

where $\text{CORR}(\tilde{X}_j, \tilde{X}_k) = r_{jk} + O(\frac{1}{n^2})$.

Let $1 \leq t_{1,n}, \dots, t_{i-1,n} \leq n$ be chosen to maximize

$$P\{\tilde{X}_j \leq b_n - \epsilon/2 a_n, |j-t_u| > n^\theta, u=1, \dots, i-1, 1 \leq j \leq n\}$$

Let $Y_{1,m}, Y_{2,m}, \dots, Y_{m,m}$ represent the $\tilde{X}_j, |j-t_u| > n^\theta, u=1, \dots, i-1, 1 \leq j \leq n$ in their natural order and let $M_m = \max_{1 \leq k \leq m} Y_k$. Note $n - 2(i-1)n^\theta \leq m \leq n$. Then the sum in

(2.36) is at most

$$(2.38) \quad P\{M_m \leq b_m - \epsilon/4 a_m\} \cdot \sum_{t_1, \dots, t_{i-1}} P\{b_n - \epsilon a_n \leq X_{t_u} \leq b_n + 2a_n, u=1, \dots, i-1\}$$

It is easily checked that the $Y_{k,m}$ satisfy the hypothesis of Lemma 4. Hence

$$(2.39) \quad P\{M_m \leq b_m - \epsilon/4 a_m\} \leq e^{-c(\ln \ln n)^2}$$

for some constant $c > 0$ not depending on n .

Further if $\min\{|t_u - t_v| : 1 \leq u < v \leq i-1\} \geq n^\theta$, then following the approach in Lemma 1 one can check that

$$(2.40) \quad P\{b_n - \epsilon a_n \leq X_{t_u} \leq b_n + 2a_n, u=1, \dots, i-1\} \leq (1 - \Phi(b_n - \epsilon a_n))^{i-1} \\ = (1/n e^{i \epsilon \ln \ln n})^{i-1}$$

While if there are exactly h indices say u_1, \dots, u_h such that when the t 's are ordered $t_{(u_1+1)} - t_{(u_1)} < n^\theta, \dots, t_{(u_h+1)} - t_{(u_h)} < n^\theta$ then

$$P\{b_n - \epsilon a_n \leq X_{t_u} \leq b_n + 2a_n, u=1, \dots, i-1\} \\ \leq (1 - \Phi(b_n \delta))^h (1 - \Phi(b_n - \epsilon a_n))^{i-h-1}$$

where $0 < \delta < 1$ is some constant not depending on n

$$(2.41) \quad \leq (1/n)^{h\delta^2 + i - h - 1} e^{\epsilon i^2 (\ln \ln n)}$$

Therefore by choosing $\theta < \delta^2$ we have by (2.39), (2.40) and (2.41) that (2.38) is at most $e^{-c(\ln \ln n)^2}$ for some $c > 0$. Therefore by (2.34), (2.35) and the above

we see that

$$P\{Z_n^{(\ell)} < b_n - \varepsilon a_n\} \leq \frac{1}{(\ell n)^2} .$$

Since this series evaluated at the n_k is summable on k , (2.33) holds completing the proof of Lemma 5.

Theorem 1. Under the assumptions of Lemma 1 the almost sure limit points of the sequence $(v_n^{(1)}, \dots, v_n^{(\ell n)}, 0, 0, \dots)$ in ℓ_1 coincide with the set

$$A = \{(x_1, x_2, \dots) : 0 \leq x_{i+1} \leq x_i, i=1, 2, \dots, \sum_1^{\infty} x_i \leq 1\}$$

Proof: Lemmas 1 and 2 establish that each point of A is an almost sure limit point while Lemmas 3 and 5 establish that no point in A^c can be an almost sure limit point.

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