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DISTRIBUTIONS ARISING IN THE MODELLING OF  
ENVIRONMENTAL PROCESSES

by

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## ABSTRACT

We study a probabilistic model of effect of environmental changes on crop production proposed by Todorovic and Gani (1987). We discuss the limit behavior of total and maximal crop yields, describe the limit distributions arising in certain special cases and estimate the rate of convergence to the limit.

## INTRODUCTION

The following probabilistic model of effect of environmental changes on crop production is considered. Let  $X_n$  denote the annual crop yield for the  $n$ th year for a given area. The sequence of crop yields can be modelled as

$$(1.1) \quad X_n = Y_n \prod_{i=1}^n Z_i, \quad n = 1, 2, \dots,$$

where  $Y$ 's are nonnegative r.v.'s and  $Z$ 's are i.i.d. independent of  $Y$ 's.  $\{Y_n\}$  is treated as a sequence of "ideal" annual yields (i.e. in the case when the environmental effect is not taken into account). In (1.1),  $Z_i$  is the proportion of crop yield maintained in year  $i$  as the result of the environmental effect from the previous year due to a "bad" year ( $Z_i < 1$ ) or "good" year ( $Z_i \geq 1$ ). In the case of  $Z_i < 1$  (with  $Z_i$  treated as the effect of soil erosion during the year  $i - 1$ ) Todorovic and Gani (1987) showed that the total crop yield over an  $n$ -year period,

$$(1.2) \quad S_n = \sum_{i=1}^n X_i = \sum_{i=1}^n Y_i \prod_{j=1}^i Z_j,$$

has a strong limit  $S$  under the regularity condition  $E Y_1 < \infty$ . They also found the distribution of  $S$  for some special cases of d.f.'s  $F_{Z_1}$  and  $F_{Y_1}$ . Puri (1987) obtained a criterion for a.s. convergence of  $S_n$ .

Similarly to (1.2), the maximum

$$(1.3) \quad M_n = \bigvee_{i=1}^n X_i = \bigvee_{i=1}^n Y_i \prod_{j=1}^i Z_j$$

is viewed as the maximum crop yield for a period of  $n$  years. For  $0 \leq Z \leq 1$  Todorovich (1987) showed the existence of a weak limit for  $M_n$ .

Suppose that in each year a disastrous event may occur with probability  $p \in (0, 1)$ . As a result, of the disastrous event (for example, drought) the whole production could be devastated. The “disastrous year”  $\tau = \tau(p)$  is assumed to be geometrically distributed:

$$(1.4) \quad P(\tau(p) = k) = (1 - p)p^{k-1}, k = 1, 2, \dots,$$

and independent of  $Y$ 's and  $Z$ 's. Assuming that  $Y_i$ 's are iid the above implies that the total crop yield (until the year  $\tau$ ) is

$$(1.5) \quad S_\tau = \sum_{i=1}^{\tau} X_i \stackrel{d}{=} \sum_{i=1}^{1+\delta\tau} X_i$$

( $\delta$  being Bernoulli with success probability  $(1 - p)$  and independent of the rest r.v.'s)

$$= X_1 + \delta \sum_{k=2}^{1+\tau} X_k \stackrel{d}{=} (Y + \delta S_\tau) Z,$$

where  $Y (\stackrel{d}{=} Y_1)$ ,  $Z (\stackrel{d}{=} Z_1)$ ,  $S_\tau$  and  $\delta$  are assumed to be independent. (See Rachev and Todorovich (1990) for some examples of distributions of  $S_\tau$ ; if  $Z \equiv 1$ ,  $S_\tau$  is said to be geometrically infinitely divisible r.v., see Klebanov, Maniya, Melamed (1989).)

The disastrous event could be caused by the cumulative effect of a large number of “bad” events with high success probability,  $p \approx 1$ . This leads to a negative binomial model for the disastrous year  $\tau = \tau_1 + \dots + \tau_r$  where  $\tau_i$ 's are i.i.d. geometric with mean  $1/p$ . For  $p \rightarrow 1$ ,  $r \rightarrow \infty$ ,  $r(1 - p) \rightarrow \lambda > 0$  we have the Poisson approximation  $P(\tau = n + r) \approx e^{-\lambda} \lambda^n / n!$  This, in turn, leads to a Poisson model for the disastrous year. Assuming that  $N_i$  are i.i.d.  $\text{Poiss}(\lambda_i)$  and  $T_i = N_1 + \dots + N_i$  is viewed as the  $i$ th disastrous year, of interest are the distributions of

$$(1.6) \quad S_{T_1} = \sum_{i=0}^{T_1} X_i, \quad S_{T_{k+1}} - S_{T_k} = \sum_{i=T_k+1}^{T_{k+1}} X_i.$$

For the sake of convenience we start the process in (1.6) at year 0, i.e. we consider the sequences  $\{Y_n\}_{n \geq 0}$ ,  $\{Z_n\}_{n \geq 0}$  and  $X_n = Y_n \prod_{i=0}^n Z_i$ ,  $n \geq 0$ .

Similarly to (1.5) and (1.6) we shall be interested in the laws of

$$(1.7) \quad M_\tau = \bigvee_{i=1}^{\tau} X_i$$

and

$$(1.8) \quad M_{T_1} = \prod_{i=0}^{T_1} X_i, \quad \prod_{i=T_k+1}^{T_k} X_i, \quad k=1, 2, \dots .$$

Our results regarding the functionals (1.2), (1.3), (1.5)-(1.8) fall into three main categories. First, we prove (in Section 2) limit theorems for the above functionals. Second (in Section 3) we provide a characterization of some classes of distributions arising as limits in (1.1) and (1.3). Third (in Section 4) we consider the multidimensional setting. Given a family of crop years  $\{A_t, t \in T\}$  we will be interested in the limit behaviors of the random fields

$$\{S_n(t) := \sum_{i=1}^n Y_i(t) \prod_{j=1}^i Z_j(t), t \in T\},$$

and  $M_n(\cdot)$  where  $n$  may be random. This approach is an example of how one can consider the problems discussed in sections 2 and 3 in a more general setting.

## 2. CENTRAL LIMIT THEOREMS

Let  $\{Y_n\}_{n \geq 0}$  be a sequence of nonnegative i.i.d. r.v.'s,  $P(Y_0 > 0) > 0$ , and  $\{Z_n\}_{n \geq 0}$  is an independent of it sequence of positive i.i.d. r.v.'s. Set  $X_n$ ,  $S_n$  and  $M_n$  as in (1.1) -(1.3). Denote  $\xi_j = \log Z_j$ ,  $\nu = E \xi_0$  (when exists).

Lemma 2.1 Suppose that  $E \log(1 + Z_0) < \infty$ .

(a) If  $\nu \geq 0$  then w.p.1,  $X_n$  does not converge to 0 and thus  $S_n \rightarrow \infty$ . Moreover,  $M_n \rightarrow \infty$  if  $P(Z_0 = 1) \neq 1$ .

(b) If  $-\infty < \nu < 0$  the following are equivalent as  $n \rightarrow \infty$ :

- (bi)  $X_n \rightarrow 0$  a.s.;
- (bii)  $S_n$  converges to a finite limit a.s.;
- (biii)  $M_n$  converges to a finite limit a.s.;
- (biv)  $0 \leq E \log(1 + Y_0) < \infty$ .

Moreover (biv) implies (bi)-(biii) even if  $\nu = -\infty$ .

Proof If  $\nu > -\infty$ , then

$$(2.1) \quad X_n = Y_n e^{n\nu} \exp\left\{\sum_{j=1}^n (\xi_j - \nu)\right\}.$$

(a) Let first  $\nu > 0$ . By SSLN, for  $n$  large,  $|\sum_{j=1}^n (\xi_j - \nu)| \leq n\nu/2$  and thus

$X_n \geq Y_n e^{n\nu/2}$ , proving (a) in this case. Let now  $\nu = 0$ . Using (2.1) again and the fact that  $\sum_{j=1}^n \xi_j$  does not converge to  $-\infty$  w.p.1, it is clear that  $X_n$  w.p.1 does not converge to 0 as is claimed in (a). If  $Z_j \neq 1$  then  $X_n = Y_n \exp\{\sum_{j=1}^n \xi_j\}$  has with probability 1, infinite maximum as  $n \rightarrow \infty$ .

(b) If  $\nu > -\infty$ , then our statement follows from a simple modification (accounting for  $P(Z_j > 1) > 0$ , in general) of Theorems 1 and 2 of Puri (1987).

For  $\nu = -\infty$  choose a finite negative  $K$  with  $-\infty < E(\xi_1 \vee K) < 0$ . Set  $W_j = Z_j \vee e^K$ . Thus, as before,  $\sum_{n=1}^{\infty} Y_n \prod_{j=1}^n W_j < \infty$  a.s., and since  $Z_j \leq W_j$ , (bii) holds.  $\square$

Our next theorem is the CLT for the "total crop yield"  $S_n$  in (1.2). We assume that  $\xi_j = \log Z_j$  belongs to the domain of attraction of an  $\alpha$ -stable r.v.  $\eta_\alpha$  ( $1 < \alpha \leq 2$ ), i.e. there exist  $a_n > 0$  and  $b_n \in \mathbb{R}$  such that

$$(2.2) \quad a_n \sum_{i=1}^n \xi_i + b_n \xrightarrow{\mathfrak{D}} \eta_\alpha, \quad a_n = n^{-1/\alpha} L(n),$$

$L(n)$  is a slowly varying function.

**Theorem 2.1** Suppose that  $E \log(1 + Z_0) < \infty$ .

(a) If  $\nu > 0$  then  $a_n \log S_n + b_n \xrightarrow{\mathfrak{D}} \eta_\alpha$ .

(b) If  $\nu < 0$  then  $a_n \log(S - S_n) + b_n \xrightarrow{\mathfrak{D}} \eta_\alpha$ .

(c) Let  $\nu = 0$ , and assume (without loss of generality), that  $b_n \equiv 0$ . Suppose also that

$$(2.3) \quad P(\log Y_1 > 1/a_n) = o(n^{-1}), \quad n \rightarrow \infty.$$

Then, as  $n \rightarrow \infty$ ,

$$(2.4) \quad a_n \log S_n \xrightarrow{\mathfrak{D}} \sup_{0 \leq t \leq 1} \mathcal{L}(t),$$

where  $\mathcal{L}$  is a Levy stable motion on  $[0,1]$  with  $\mathcal{L}(1) \stackrel{d}{=} \eta_\alpha$ .

**Proof** (a) From Lemma 2.1 (a),  $S_n \rightarrow \infty$  a.s. Let  $U_n$  stand for the first  $X_k$ ,  $k > n$ , different from 0, i.e.  $U_n = X_{R_n}$  where  $R_n = \min\{k > n, X_k \neq 0\}$ .

Now (2.2) implies the weak convergence

$$(2.5) \quad a_n \log U_n + b_n = a_n \log Y_{R_n} + a_n \sum_{j=n+1}^{R_n} \xi_j + a_n \sum_{j=1}^n \xi_j + b_n \xrightarrow{\mathfrak{D}} \eta_\alpha.$$

On the other hand for any  $\epsilon > 0$ ,

$$(2.6) \quad \begin{aligned} \lim_{n \rightarrow \infty} P(a_n \log (S_n/U_n) > \epsilon) &= \lim_{n \rightarrow \infty} P(a_n \log (S_{R_n}/X_{R_n}) > \epsilon) \\ &= \lim_{n \rightarrow \infty} P\left(\sum_{i=1}^{n-1} Y_i \prod_{j=i+1}^n Z_j^{-1} > (e^{\epsilon/a_n} - 1) Y_{R_n} \prod_{j=n+1}^{R_n} Z_j\right) = 0. \end{aligned}$$

The last equality follows from Lemma 2.1 (b), as  $\sum_{i=1}^n Y_i \prod_{j=1}^i Z_j^{-1}$  converges to a finite limit.

(b) The proof is similar to that of part (a).

(c) Let  $m_n = a_n \max_{m=1, \dots, n} \sum_{j=1}^m \xi_j$ ,  $n = 1, 2, \dots$ . Then for any  $\lambda > 0$

$$(2.7) \quad P(S_n > \lambda) \leq P((Y_1 + \dots + Y_n) \exp\{a_n^{-1} m_n\} > \lambda).$$

Define the processes  $\mathcal{L}_n(\cdot) = a_n \sum_{j=1}^{[n \cdot]} \xi_j$  in  $\mathfrak{D}[0, 1]$ . Then by (2.2),  $\mathcal{L}_n \xrightarrow{w} \mathcal{L}$  as  $n \rightarrow \infty$ , in the Skorokhod topology. Thus, as  $n \rightarrow \infty$ ,  $m_n \vee 0 = \sup_{0 \leq t \leq 1} \mathcal{L}_n(t) \xrightarrow{\mathfrak{D}} \sup_{0 \leq t \leq 1} \mathcal{L}(t)$ .

Observe that  $P(m_n \geq 0) \rightarrow 1$  and, therefore,

$$(2.8) \quad m_n \xrightarrow{\mathfrak{D}} \sup_{0 \leq t \leq 1} \mathcal{L}(t).$$

Invoking (2.7), we obtain

$$(2.9) \quad P(a_n \log S_n > u) \leq P(a_n \log (Y_1 + \dots + Y_n) + m_n > u).$$

Next let us show that  $a_n \log(Y_1 + \dots + Y_n) \xrightarrow{P} 0$  provided that (2.3) holds: for any  $\epsilon > 0$ ,  $\delta > 0$  and  $n$  large,

$$\begin{aligned} P(a_n \log (Y_1 + \dots + Y_n) > \epsilon) &\leq P\left(\max_{i=1, \dots, n} Y_i > \frac{1}{n} \exp\{\epsilon/a_n\}\right) \\ &\leq 1 - (1 - P(\log Y_1 > \frac{1}{2} \epsilon/a_n))^n \leq 1 - (1 - \delta n^{-1})^n \approx 1 - e^{-\delta}. \end{aligned}$$

Letting  $\delta \rightarrow 0$ , we obtain  $a_n \log(Y_1 + \dots + Y_n) \xrightarrow{P} 0$ . Thus

$$(2.10) \quad \overline{\lim}_{n \rightarrow \infty} \frac{P(a_n \log S_n > u)}{P\left(\sup_{0 \leq t \leq 1} \mathcal{L}(t) > u\right)} \leq 1.$$

The opposite inequality is trivial in the case  $P(Y_1 > 0) = 1$ . Indeed, in this case we can use

$$P(S_n > \lambda) \geq P(Y_1 \exp\{a_n^{-1} m_n\} > \lambda),$$

which gives the right estimate in the same way as above. The general case  $0 < P(Y_1 > 0) < 1$  is much more involved.

We start with the case  $\alpha = 2$  in (2.2). For  $1 \leq k \leq n$  let  $m_n^{(k)}$  be the  $k$ th biggest value among  $a_n \sum_{j=1}^m \xi_j$ ,  $m=1, \dots, n$  ( $m_n^{(1)} = m_n$ ). Then, if  $n > 2k$ ,

$$(2.11) \quad |m_n - m_n^{(k)}| \leq (k-1)a_n^{-1} \max_{i=1, \dots, n} \xi_i \xrightarrow{P} 0.$$

Therefore, for any  $k = 1, 2, \dots$ , and  $n > 2k$

$$\begin{aligned} P(S_n > \lambda) &\geq P\left(\bigcup_{i=1}^k \{Y_i \exp\{a_n^{-1} m_n^{(i)}\} > \lambda\}\right), \\ &\geq \sum_{i=1}^k p(1-p)^{i-1} P(\tilde{Y}_1 \exp\{a_n^{-1} m_n^{(i)}\} > \lambda), \end{aligned}$$

where  $p = P(Y_1 > 0)$  and where  $\tilde{Y}_1$  has the conditional distribution of  $Y_1$  given  $Y_1 > 0$ .

Therefore, by (2.8) and (2.11)

$$\begin{aligned}
P(a_n \log S_n > u) &\geq \sum_{i=1}^k p(1-p)^{i-1} P(\tilde{Y}_1 \exp\{a_n^{-1} m_n^{(i)}\} > \exp\{a_n^{-1} u\}) \\
&= \sum_{i=1}^k p(1-p)^{i-1} P(a_n \log \tilde{Y}_1 + m_n^{(i)} > u) \\
&\rightarrow (1 - (1-p)^k) P\left(\sup_{0 \leq t \leq 1} \mathcal{L}(t) > u\right).
\end{aligned}$$

Since this is true for every  $k = 1, 2, \dots$ , we obtain the required counterpart to (2.10).

The case  $1 < \alpha < 2$  requires a different argument, as (2.11) fails in this case. Let  $T_0 = 0$ ,  $T_m = \inf\{k > T_{m-1} : Y_k > 0\}$ ,  $m = 1, 2, \dots$ , and let  $W_m = \sum_{i=T_{m-1}+1}^{T_m} \xi_i$ ,  $m = 1, 2, \dots$ . Then  $W_1, W_2, \dots$  is a sequence of i.i.d. random variables such that

$$\lim_{\lambda \rightarrow \infty} \frac{P(W_1 > \lambda)}{P(\xi_1 > \lambda)} = \lim_{\lambda \rightarrow \infty} \frac{P(W_1 < -\lambda)}{P(\xi_1 < -\lambda)} = \frac{1}{p}.$$

Let now  $\tilde{\mathcal{L}}_n(t) = a_n \sum_{j=1}^{[nt]} W_j$ ,  $0 \leq t \leq 1$ . Then  $\tilde{\mathcal{L}}_n \xrightarrow{w} p^{-1/\alpha} \mathcal{L}$  as  $n \rightarrow \infty$  in the Skorokhod

topology of  $\mathfrak{D}[0,1]$ . For  $n = 1, 2, \dots$  define

$$N_n = \# \text{ of } T_i \text{'s such that } T_i \leq n.$$

By the SLLN,

$$(2.12) \quad \frac{N_n}{n} \xrightarrow{n \rightarrow \infty} p \text{ a.s.}$$

We have

$$P(S_n > \lambda) = P\left(\sum_{m=1}^{N_n} Y_{T_m} \exp\left\{\sum_{j=1}^m W_j\right\} > \lambda\right)$$

$$\geq P\left(\max_{i=1, \dots, n} Y_i > 0\right) P(Y_{T_1} e^{V_n} > \lambda) \approx (1 - (1-p)^n) P(Y_{T_1} e^{V_n} > \lambda),$$

where  $V_n = \max_{m=1, \dots, N_n} \sum_{j=1}^m W_j$ . Let  $\tilde{\mathcal{L}}_n(t) = a_{N_n} \sum_{j=1}^{[N_n t]} W_j$ ,  $0 \leq t \leq 1$ . It follows from (2.12) and Theorem 17.1 of Billingsley (1968) (easily modifiable to our case) that  $\tilde{\mathcal{L}}_n \xrightarrow{w} p^{-1/\alpha} \mathcal{L}$  as  $n \rightarrow \infty$  is  $\mathfrak{D}[0,1]$ . Therefore,  $a_{N_n} V_n \xrightarrow{w} p^{-1/\alpha} \sup_{0 \leq t \leq 1} \mathcal{L}(t)$ . Note also that by (2.12),  $a_{N_n}/a_n \rightarrow p^{1/\alpha}$  a.s. as  $n \rightarrow \infty$ . Therefore,



$$\begin{aligned} P(a_n \log S_n > u) &\geq (1 - (1-p)^n)P(Y_{T_1} e^{V_n} > \exp\{a_n^{-1} u\}) \\ &= (1 - (1-p)^n)P(a_n \log Y_{T_1} + a_n V_n > u) \xrightarrow{n \rightarrow \infty} P(\sup_{0 \leq t \leq 1} \mathcal{L}(t) > \lambda). \end{aligned}$$

This completes the proof of the theorem.  $\square$

As far as the maximal yield (1.3) for  $n$  years is concerned we have the following analogue of Theorem 2.1.

Theorem 2.2. Under the assumptions of Theorem 2.1 the following holds:

(a) If  $\nu \geq 0$  then  $a_n \log M_n + b_n \xrightarrow{\mathfrak{D}} \eta_\alpha$ ;

(b) If  $\nu < 0$  then  $a_n \log(\bigvee_{j>n} X_j) + b_n \xrightarrow{\mathfrak{D}} \eta_\alpha$ ;

(c) If  $\nu = 0$  and (2.4) holds then  $a_n \log M_n \xrightarrow{\mathfrak{D}} \sup_{0 \leq t \leq 1} \mathcal{L}(t)$ .

Proof (a) and (c) follow from the corresponding assertions in Theorem 2.1 observing that  $\frac{S_n}{n} \leq M_n \leq S_n$ , and the proof of (b) uses arguments similar to those in Theorem 2.1.  $\square$

Next we examine the geometric random sum  $S_\tau$  given by (1.5) and (1.4). We say that  $\xi_0 = \log Z_0$  belongs to the domain of attraction of geometric  $\alpha$ -stable r.v.  $G_\alpha$  if there exist functions  $a = a(p) > 0$  and  $b = b(p)$  on  $[0, 1]$  such that

$$(2.13) \quad a \sum_{i=1}^{\tau} (\xi_i + b) \xrightarrow{\mathfrak{D}} G_\alpha \text{ as } p \rightarrow 0.$$

Here  $a(p) = p^{1/\alpha} L(1/p)$ , where  $L$  is slowly varying function. The ch.f. of  $G_\alpha$  admits the representation  $f_{G_\alpha}(t) = 1/(1 - \log \phi_\alpha(t))$ , where  $\phi_\alpha$  is the ch.f. of an  $\alpha$ -stable r.v. (Klebanov, Maniya and Melamed (1984)). Similarly,  $f_{\xi_0} = 1/(1 - \log \psi)$ , where  $\psi$  is the ch.f. of a distribution in the domain of attraction of  $\alpha$ -stable r.v. with ch.f.  $\phi_\alpha$  (Mittnik and Rachev (1990)). Examples of geometric  $\alpha$ -stable distribution are the exponential law ( $\alpha = 1$ ) and the Laplace law ( $\alpha = 2$ ).

Theorem 2.3. Suppose that  $\mathbb{E} \log(1 + Z_0) < \infty$  and (2.13) holds.

(a) If  $\nu > 0$  then  $a(\log S_\tau - \tau b) \xrightarrow{\mathfrak{D}} G_\alpha$  as  $p \rightarrow 0$ ;

(b) If  $\nu < 0$  then  $a(\log \sum_{j \geq \tau+1} X_j - \tau b) \xrightarrow{\mathfrak{D}} G_\alpha$  as  $p \rightarrow 0$ ;

(c) If  $\nu = 0$  and

$$P(\log Y_1 > n^{1/\alpha} L(n)^{-1}) = o(n^{-1}), \quad n \rightarrow \infty,$$

then

$$(2.14) \quad a \log S_\tau \xrightarrow{\mathfrak{D}} \sup_{0 \leq t \leq 1} \mathfrak{G}(t) \quad \text{as } p \rightarrow 0.$$

Here  $\mathfrak{G}$  is a “geometric Levy stable motion”, i.e. is the weak limit in  $\mathfrak{D}[0,1]$  of

$$\mathfrak{G}_p(t) = a \sum_{j=1}^{[\tau t]} \xi_j, \quad 0 \leq t \leq 1.$$

The proof is similar to that of Theorem 2.1 and thus omitted.

Remark 2.1. Regarding the existence of the process  $\mathfrak{G}$  as the weak limit of  $\mathfrak{G}_p$  one can check the following:

(a) The finite dimensional distribution  $(\mathfrak{G}_p(t_1), \dots, \mathfrak{G}_p(t_d))$  ( $0 \leq t_1 < \dots < t_d \leq 1$ ) converges to a “geometric strictly stable distribution”  $(\mathfrak{G}(t_1), \dots, \mathfrak{G}(t_d))$  with ch.f.  $g(\theta)$  of the form  $1/(1 - \log \psi(\theta))$ , where  $\psi(\theta)$  is the ch.f. of a strictly  $\alpha$ -stable random vector on  $\mathbb{R}^d$ ;

(b) The set of laws of  $\mathfrak{G}_p(\cdot)$  ( $0 < p < 1$ ) is tight.

Similarly to Theorems 2.2 and 2.3 we obtain the following limit theorem for the distribution of the maximal crop yield.

Theorem 2.4 Under the assumptions of Theorem 2.3 the following holds:

(a) If  $\nu > 0$  then  $a(\log M_\tau - \tau b) \xrightarrow{\mathfrak{D}} G_\alpha$  as  $p \rightarrow 0$ ;

(b) If  $\nu < 0$  then  $a(\log \bigvee_{j=\tau+1} X_j - \tau b) \xrightarrow{\mathfrak{D}} G_\alpha$  as  $p \rightarrow 0$ ;

(c) Suppose that the conditions of Theorem 2.3 (c) holds. Then, as  $p \rightarrow 0$ ,

$$a \log M_\tau \xrightarrow{\mathfrak{D}} \sup_{0 \leq t \leq 1} \mathfrak{G}(t).$$

Finally, let us consider the total crop yield until a Poisson ( $\lambda$ ) random moment  $T = T(\lambda)$ . Suppose, as before, that  $Y_i$ 's are i.i.d.,  $Z_i$ 's are i.i.d. and  $T$  is independent of  $Y_i$ 's and  $Z_i$ 's. Suppose also that the ch.f.  $f_{\xi_0}$  of  $\xi_0 = \log Z_0$  satisfies

$$(2.15) \quad \lim_{u \rightarrow 0} |u|^{-\alpha} (1 - f_{\xi_0^{-a}}(u)) = \mu$$

for some  $\mu > 0$ , real  $a$  and  $0 < \alpha \leq 2$ .

Theorem 2.5. Suppose that (2.15) holds and let

$$S_T = \sum_{i=0}^T X_i = \sum_{i=0}^T Y_i \prod_{j=0}^i Z_j.$$

(a) If  $\nu = E\xi_0 > 0$  then, as  $\lambda \rightarrow \infty$ ,

$$(2.16) \quad \lambda^{-1/\alpha} (\log S_T - aT) \stackrel{\mathfrak{D}}{\Rightarrow} Y_{(\alpha)},$$

where  $Y_{(\alpha)}$  is symmetric stable r.v. with ch.f.  $\exp\{-\mu|\theta|^\alpha\}$ ;

(b) If  $\nu < 0$  then, as  $\lambda \rightarrow \infty$ ,

$$(2.17) \quad \lambda^{-1/\alpha} (\log \sum_{j=T}^{\infty} X_j - aT) \stackrel{\mathfrak{D}}{\Rightarrow} Y_{(\alpha)},$$

$$(2.18) \quad \lambda^{-1/\alpha} (\log \sum_{j=T_1}^{T_k} X_j - aT_1) \stackrel{\mathfrak{D}}{\Rightarrow} Y_{(\alpha)},$$

where  $T_1, T_k$  are as in (1.6);

(c) Let  $\nu = 0$  and  $P(\log Y_1 > u) = o(u^{-\alpha})$  as  $u \rightarrow \infty$ . Then, as  $\lambda \rightarrow \infty$ ,

$$\lambda^{-1/\alpha} \log S_T \stackrel{\mathfrak{D}}{\Rightarrow} \sup_{0 \leq t \leq 1} \mathcal{L}(t),$$

where  $\mathcal{L}(\cdot)$  is a Levy stable motion on  $[0, 1]$  with  $\mathcal{L}(1) \stackrel{d}{=} Y_{(\alpha)}$ .

Proof. For simplicity we assume that  $P(Y_1 > 0) = 1$ . The problems arising from  $P(Y_1 = 0) > 0$  can be handled as in the proof of Theorem 2.1.

(a) Since  $\nu > 0$  then by Lemma 2.1,  $S_T \rightarrow \infty$  as  $\lambda \rightarrow \infty$ .

Claim. Under (2.15),

$$(2.19) \quad \lambda^{-1/\alpha} \sum_{i=0}^T \eta_i \xrightarrow{\mathfrak{D}} Y_{(\alpha)},$$

where  $\eta_i = \xi_i - a$ .

In fact, for the ch.f. of the left hand side of (2.19) we have

$$f_{\lambda^{-1/\alpha} \sum_{i=1}^T Y_i}(\theta) = \exp\{-\lambda(1 - f_{\eta_0}(\lambda^{-1/\alpha} \theta))\}$$

(setting  $\lambda^{-1/\alpha} \theta = u$  and letting  $u \rightarrow 0$ )

$$= \exp\{-\mu |\theta|^\alpha \frac{1}{\mu} |u|^{-\alpha} (1 - f_{\eta_0}(u))\} \rightarrow \exp\{-\mu |\theta|^\alpha\}. \quad \square$$

Using the claim we arrive at

$$\lambda^{-1/\alpha} (\log X_T - aT) = \lambda^{-1/\alpha} \log Y_T + \lambda^{-1/\alpha} \sum_{i=0}^T (\xi_i - a) \xrightarrow{\mathfrak{D}} Y_{(\alpha)}.$$

Further, similarly to Theorem 2.1 (a), we conclude that for any  $\epsilon > 0$ ,

$$P(\lambda^{-1/\alpha} (\log S_T - \log X_T) > \epsilon) \rightarrow 0 \text{ as } \lambda \rightarrow \infty.$$

The proof of (b) and (c) is handled as in Theorem 2.1 (b) and (c), respectively.  $\square$

Analogous theorems can be established regarding the limit distributions of  $M_{T_1}^{\bigvee_{i=T_k+1}^{T_{k+1}}} X_i$  (c.f. (1.8)).

### 3. CHARACTERIZATIONS OF THE DISTRIBUTIONS OF THE TOTAL AND MAXIMAL CROP YIELD.

The subject of this section is the characterizations of the limit laws of  $S_n$  (cf.(1.2))

and  $M_n$  (cf. (1.3)).

Invoking Lemma 2.1 (b) we conclude that for  $Z_0 \in (0, 1)$ ,  $S_n$  (resp.  $M_n$ ) converges to a finite limit  $S$  (resp.  $M$ ) if (and only if, if  $E \log Z_0 > -\infty$ )

$$(3.1) \quad 0 \leq E \log (1 + Y_0) < \infty.$$

Given (3.1), the limits  $S$  and  $M$  satisfy the equations:

$$(3.2) \quad S \stackrel{d}{=} (S + Y) Z$$

and

$$(3.3) \quad M \stackrel{d}{=} (M \vee Y) Z,$$

where  $Y \stackrel{d}{=} Y_0$ ,  $Z \stackrel{d}{=} Z_0$  and all r.v.'s above are assumed to be mutually independent. Next we give a complete characterization of the class  $\mathfrak{F}_1$  (resp.  $\mathcal{M}_1$ ) of laws  $L(S)$  of  $S$  (resp.  $L(M)$ ) such that for any  $L(Z) \in \mathfrak{Z}_1$  there exist  $Y = Y(Z)$  such that (3.2) (resp. (3.3)) holds. The class  $\mathfrak{Z}_1$  of  $Z$ -laws  $L(Z)$  consists of distributions on  $(0, 1)$  with densities

$$(3.4) \quad f_\alpha(z) = (1 + \alpha)z^\alpha, \quad 0 < z < 1, \quad \alpha \geq 0.$$

The class  $\mathfrak{Z}_1$  was introduced by Todorovich and Gani (1987). Some particular examples of laws  $L(S) \in \mathfrak{F}_1$ ,  $L(M) \in \mathcal{M}_1$  were studied by Todorovich and Gani (1987), Todorovich (1987), and Rachev and Todorovich (1990).

In the sequel for any  $0 < \beta < 1$ , and a r.v.  $Y$ , denote

$$(3.5) \quad Y_\beta := \begin{cases} 0 & \text{with probability } 1 - \beta, \\ Y & \text{with probability } \beta. \end{cases}$$

**Lemma 3.1** Suppose  $S$  is a solution of (3.2) for a given  $Y$  with  $0 \leq E \log (1 + Y) < \infty$  and  $Z$  being uniform. Then  $S$  is also solution (3.2) with  $Z$  having density (3.4) and  $Y$  replaced by  $Y_{1/(1+\alpha)}$ .

**Proof** If  $Z$  is uniform then (3.2) implies the following expression for the Laplace transform  $\phi_Y$  of  $Y$ :

$$\phi_Y(\theta) = 1 + \theta \phi'_S(\theta) / \phi_S(\theta).$$

Using (3.5) we observe that

$$\phi_{Y_{1/(1+\alpha)}}(\theta) = 1 + (1 + \alpha)^{-1} \theta \phi'_S(\theta) / \phi_S(\theta).$$

Integrating the above equation gives

$$\phi_S(\theta) = \theta^{-\alpha} \int_0^\theta (1 + \alpha) x^\alpha \phi_S(x) \phi_Y(x) dx = \mathbb{E} e^{-\theta(Z+Y_{1/(1+\alpha)})Z},$$

where  $Z$  has density given by (3.4). □

The above lemma allows us to restrict our considerations only to the case of  $Z$  being uniform, which is what we do in the sequel.

**Lemma 3.2.** Any  $Y$  with

$$(3.6) \quad 0 \leq \mathbb{E} \log(1 + Y) < \infty$$

solves (3.2) for  $S$  whose Laplace transform is given by

$$(3.7) \quad \phi_S(\theta) = \exp\left\{-\int_0^\theta \frac{1}{x}(1 - \phi_Y(x)) dx\right\}.$$

Moreover, (3.6) is necessary for (3.7) to be a solution of (3.2).

Proof. Assuming,  $S \stackrel{d}{=} (S + Y)Z$  and  $Z$  being uniform, (3.7) holds (cf. Todorovich and Gani (1987)). From (3.7),  $\int_0^\theta \frac{1}{x}(1 - \phi_Y(x)) dx < \infty$ . Now,

$$(3.8) \quad \frac{1}{\theta}(1 - \phi_Y(\theta)) = \int_0^\infty e^{-\theta x} \bar{F}_Y(x) dx, \quad \bar{F}_Y := 1 - F_Y,$$

and so

$$\infty > \int_0^1 \int_0^\infty e^{-xy} \bar{F}_Y(y) dy dx = \int_0^\infty \bar{F}_Y(y) \frac{1-e^{-y}}{y} dy.$$

Thus

$$\infty > \int_1^\infty \frac{1}{y} \bar{F}_Y(y) dy = \int_1^\infty \log y F_Y(dy).$$

i.e. (3.6) holds.

Note that by (3.8),  $\frac{1}{\theta}(1 - \phi_Y(\theta))$  is a completely monotone function. Thus, by Feller (1971), Criterion XIII, 4.2, if (3.6) holds then  $\exp\{-\int_0^\infty \frac{1}{x}(1 - \phi_Y(x)) dx\}$  is also a completely monotone function. Therefore (3.7) defines a Laplace transform of a nonnegative r.v., which clearly satisfies (3.2).  $\square$

**Lemma 3.3.** Any r.v.  $S$  solving  $S \stackrel{d}{=} (S + Y)Z$  is infinite divisible. More specifically, let  $Y$  correspond to  $S$  in (3.2), then for any  $\beta \in (0, 1)$  the r.v.  $S^{(\beta)}$  with Laplace transform  $\phi_S^\beta$  satisfies  $S^{(\beta)} \stackrel{d}{=} (S^{(\beta)} + Y_\beta)Z$ , where  $Y_\beta$  is given by (3.5).

**Proof.** Using (3.7),

$$\phi_S^\beta(\theta) = \exp\left\{-\int_0^\infty \frac{1}{\theta}[1 - (\beta\phi_Y(\theta) + 1 - \beta)] d\theta\right\}.$$

Note that  $\beta\phi_Y + 1 - \beta$  is the Laplace transform of  $Y_\beta$  and, clearly,  $E \log(1 + Y_\beta) < \infty$ . Now use Lemma 3.2.  $\square$

We move next to give a detailed description of the class  $\mathcal{S}_1$ .

**Theorem 3.1.** The class  $\mathcal{S}_1$  of the laws  $L(S)$  solving  $S \stackrel{d}{=} (S + Y)Z$  consists of all nonnegative infinitely divisible r.v.'s, whose Levy measure  $M_S$  is of the following form:

$$(3.9) \quad M_S \ll \text{Leb} \text{ and } M_S(dx) = H(x)dx,$$

where  $H(0) \in [0,1]$ ,  $H$  is non-increasing on  $[0, \infty)$  and vanishing at  $\infty$ . The corresponding  $Y$  has  $1 - H$  as its distribution function.

**Proof.** By Lemma 2.3 and Theorem XIII, 7.2 of Feller (1971), there is a unique Levy measure  $M_S$  on  $(0, \infty)$  such that  $\int_0^\infty (1 \wedge x^{-1}) M_S(dx) < \infty$  and  $\phi_S(\theta) = \exp\left\{-\int_0^\infty \frac{1}{x}(1 - e^{-\theta x}) M_S(dx)\right\}$ .

On the other hand,  $\phi_S$  enjoys (3.7) and thus  $\int_0^\infty \frac{1}{x}(1 - e^{-\theta x})M_S(dx) = \int_0^\theta \frac{1}{x}(1 - \phi_Y(x))dx$ .  
 Differentiating for  $\theta > 0$  we obtain, by (3.6), that  $\int_0^\infty e^{-\theta x}M_S(dx) = \frac{1}{\theta}(1 - \phi_Y(\theta)) = \int_0^\infty e^{-\theta x}\bar{F}_Y(x)dx$ .  
 By the uniqueness of  $M_S$  we conclude that  $M_S(dx) = (1 - F_Y(x)) dx, x > 0$ .  $\square$

Our next task is the characterization of the class  $\mathcal{M}_1$  of laws  $L(M)$  satisfying (3.3) for any  $L(Z) \in \mathcal{Z}_1$ .

Theorem 3.2.  $\mathcal{M}_1$  consists all absolutely continuous laws  $L(M)$  with density  $f_M$  and c.d.f.  $F_M$  satisfying the following conditions:

- (i)  $f_M(x)$  is non-increasing on  $(0, \infty)$ ;
- (ii)  $xf_M(x)/F_M(x)$  is non-increasing on  $(0, \infty)$ .

Proof. Suppose first that  $L(M) \in \mathcal{M}_1$  and let  $Z^{(\alpha)}$  have a density  $f_{Z^{(\alpha)}}(z) = (1 + \alpha)z^\alpha, 0 < z < 1$ . Then  $M \stackrel{d}{=} (M \vee Y) Z^{(\alpha)}$  is equivalent to

$$(3.10) \quad \bar{F}_Y(x) = \frac{1}{1 + \alpha} \frac{xf_M(x)}{F_M(x)}, \quad x > 0.$$

The latter implies the necessity of (ii) and

$$(3.11) \quad \left\{ \begin{array}{l} \lim_{x \downarrow 0} xf_M(x)/F_M(x) \in (0, 1], \\ \lim_{x \rightarrow \infty} xf_M(x) = 0. \end{array} \right.$$

For  $\alpha = 0$ , by (3.10), we obtain  $F_M(x) = \exp\{-\int_x^\infty \frac{1}{t}\bar{F}_Y(t) dt\}$  so that  $f_M(x)$  is a non-increasing function. This proves the necessity part of the theorem.

Suppose now that (i) and (ii) hold. Then (3.11) holds as well. Then the right-hand side of (3.10) is the distribution tail of a nonnegative random variable which satisfies  $M \stackrel{d}{=} (M \vee Y) Z^{(\alpha)}$ .  $\square$

Remark 3.1. Note that by (3.10), for any  $L(M) \in \mathcal{M}_1$  and  $0 < \alpha < 1$ ,

$$M \stackrel{d}{=} (M \vee Y_\alpha)Z^{(\alpha)} \Leftrightarrow M \stackrel{d}{=} (M \vee Y_0)Z^{(0)},$$

where  $Y_\alpha$  is determined by (3.5). The last relation is parallel to the corresponding relation in the scheme of summation (cf. Lemma 3.1).



**Remark 3.2.** Note that Gamma  $(p, \lambda)$  distributions with  $0 < p \leq 1$  belong to  $\mathcal{M}_1$  while those with  $p > 1$  do not.

Next we consider the class  $\mathcal{F}_2$  (resp.  $\mathcal{M}_2$ ) of laws  $L(S)$  (resp.  $L(M)$ ) such that (3.2) (resp. (3.3)) holds for any  $L(Z) \in \mathcal{Z}_2 \equiv \{\delta_z, 0 < z < 1\}$  and suitably chosen  $Y = Y(Z)$ .

The class  $\mathcal{F}_2$  coincides with class  $L$  of Khinchine (cf. Feller (1971), Sect. 8, Chapter XVII) of nonnegative r.v.'s. We state here a more refined description of  $\mathcal{F}_2$  that in Feller (1971), Theorem XVII.8.

**Theorem 3.3.** The class  $\mathcal{F}_2$  coincides with the family of all nonnegative infinitely divisible r.v.'s with the Laplace transform

$$(3.11) \quad \phi_S(t) = \exp\left\{-at - \int_0^\infty \frac{1}{x} (1 - e^{-tx}) M_S(dx)\right\}$$

with  $a \geq 0$  and Levy measure  $M_S : M_S \ll \lambda$ , ( $\lambda$  is the Lebesgue measure on  $[0, \infty)$ ), and  $\frac{dM_S}{d\lambda}$  has a nonincreasing version.

For any  $S \in \mathcal{F}_2$  and  $z \in (0, 1)$ , the corresponding  $Y$  in the equation

$$(3.12) \quad S \stackrel{d}{=} (S + Y) z$$

is nonnegative infinitely divisible with Laplace transform

$$(3.13) \quad \phi_Y(t) = \exp\left\{-\frac{at(1-z)}{z} - \int_0^\infty \frac{1}{x} (1 - e^{-tx}) M_Y(dx)\right\},$$

where  $M_Y \ll \lambda$  and

$$(3.14) \quad \frac{dM_Y}{d\lambda}(x) = \frac{dM_S}{d\lambda}(zx) - \frac{dM_S}{d\lambda}(x).$$

**Proof.** Clearly,

$$(3.15) \quad \phi_Y(t) = \phi_S(t/z) / \phi_S(t).$$

It follows from Lemma XVII.8 of Feller (1971) that  $S$  and  $Y$  in (3.12) must be infinitely divisible. Moreover, for any Borel set  $A$  in  $(0, \infty)$ ,

$$(3.16) \quad \frac{1}{z} M_S(zA) = M_S(A) + M_Y(A) \geq M_S(A).$$

Let  $F(x) = M_{\mathcal{G}}((0, x])$ ,  $x > 0$ . From (3.16) it follows that  $M_{\mathcal{G}} \ll \lambda$  with a nonincreasing Radon-Nikodim derivative. The fact that  $a \geq 0$  in (3.13) follows from Section XVII.4.(b) of Feller(1971). The description of  $\phi_Y$  now follows from (3.15).

Conversely, if  $S$  has a distribution described by (3.11), then for any  $z \in (0, 1)$  choose  $Y$  as in (3.13). By (3.15),  $S$  and  $Y$  solve (3.12).  $\square$

Remark 3.3 It is interesting to note that  $\mathcal{F}_1 \subset \mathcal{F}_2$ .

The class  $\mathcal{M}_2$  coincides with the class of the laws of max self-decomposable r.v.'s  $M$ , that are solutions of  $M \stackrel{d}{=} \alpha M \vee W$  ( $0 < \alpha < 1$ ) ( see Balkema, de Haan and Karandikar (1990) and the references there). The next theorem is similar to the Mejlzer (1956) characterization of the class of max self-decomposable r.v.'s. The Mejlzer (1956) result is based on a characterization of the weak limits of the normalized maxima  $a_n\{\max(X_1, X_2, \dots, X_n) - b_n\}$  when  $X_i$ 's are independent nonidentically distributed.

Theorem 3.4.  $\mathcal{M}_2$  consists of the laws of positive absolutely continuous r.v.'s  $M$  such that  $x f_M(x)/F_M(x)$  is a non-increasing function on  $(0, \infty)$ .

Proof. For  $z \in (0, 1)$ ,  $M \stackrel{d}{=} (M \vee Y) z$  implies

(a)  $F_M(xz)/F_M(x)$  is a nondecreasing function on  $(0, \infty)$  for every  $0 < z < 1$ .

Let  $H(x) = -\log F_M(x)$ . Then (a) is equivalent to

(b)  $H(x/z) - H(x)$  is a nonincreasing function on  $(0, \infty)$  for every  $0 < z < 1$ .

This is equivalent to  $H(e^x)$  being convex, then  $H$ , and thus  $F_M$  as well, are absolutely continuous.

Then (b) is equivalent to

(c)  $yH'(y)$  is a nondecreasing function on  $(0, \infty)$ .

Since  $H'(y) = -f_M(y)/F_M(y)$ , the proof is now complete.  $\square$

#### 4. THE RATE OF CONVERGENCE PROBLEM IN A GENERAL SETTING

Throughout this section,  $(\mathbb{B}, \|\cdot\|)$  is the separable Banach space  $C(T)$  of continuous mappings  $x: T \rightarrow \mathbb{R}$  where  $T$  is compact and  $\|\cdot\|$  is the usual supremum norm in  $C(T)$ . For any  $x, y \in \mathbb{B}$  we denote  $(x \cdot y)(t) := x(t) \cdot y(t)$ ,  $(x \vee y)(t) = x(t) \vee y(t)$ ,  $t \in T$ .

Given a nonatomic probability space, let  $\mathfrak{F}(\mathbb{B})$  be the space of all random fields (r.f.'s)  $X$ ,

$\mathbb{B}$ -valued random variables, and let  $\mathcal{L}(\mathbb{B})$  be the space of all laws  $P_X$ . Suppose  $Y_1, Y_2, \dots$  are i.i.d. r.f.'s,  $Z_1, Z_2, \dots$  are independent of  $Y_i$ 's i.i.d. r.f.'s and define

$$(4.1) \quad S_n := \sum_{i=1}^n X_i, \quad X_i := Y_i \prod_{j=1}^i Z_j.$$

The r.f.  $S_n$  can be interpreted as the total crop yields of crop producing areas  $t \in T = T(U)$  for a period of  $n$  years. Here  $U$  is a compact metric space,  $U = (U, \rho)$ , and  $T(U)$  is the set of all closed subsets (crop producing areas)  $t$  endowed with the Hausdorff metric

$$h(t_1, t_2) = \inf \{ \epsilon > 0 : t_1 \subset t_2^\epsilon, t_2 \subset t_1^\epsilon \},$$

where  $t^\epsilon$  stands for the  $\epsilon$ -neighborhood of  $t$  (Hausdorff (1969), Sect 29). Then  $(T, h)$  is a compact metric space (cf. Hausdorff (1969), Sect. 29; Kuratowski (1966) § 21; Kuratowski (1969) § 31).

Similarly we define the maximal crop yields

$$(4.2) \quad M_n = \bigvee_{i=1}^n X_i.$$

Next we are interested in the conditions providing exponential rate of convergence of  $S_n$  and  $M_n$  to a finite limits  $S$  and  $M$  respectively. The obtained theorem improves on the existing results in this area (cf. Rachev and Todorovich (1990)).

The rate of convergence of the laws  $P_{X_n}$  to  $P_X$  will be expressed, as usual in the Banach space setting, in terms of the Prokhorov metric

$$(4.3) \quad \pi(X, Y) := \pi(P_X, P_Y)$$

$$:= \inf \{ \epsilon > 0 : P(X \in A) \leq P(Y \in A^\epsilon) + \epsilon \text{ for all Borel subsets } A \text{ of } \mathbb{B} \},$$

where  $A^\epsilon$  is the open  $\epsilon$ -neighborhood of  $A$ . Further, we shall use also the following metrics and functions in  $\mathfrak{F}(\mathbb{B})$  and  $\mathcal{L}(\mathbb{B})$ :

(i)  $\chi_p$  - metric in  $\mathfrak{F}(\mathbb{B})$  (cf. Pisier and Zinn (1977)):

$$(4.4) \quad \chi_p(X, Y) = \left[ \sup_{t>0} t^p P(\|X - Y\| > t) \right]^{\frac{1}{1+p}}, \quad p > 0;$$

(ii)  $\hat{\chi}_p$  - minimal metric in  $\mathcal{L}(\mathbb{B})$  (cf. de Haan and Rachev (1989)):

$$\begin{aligned}
(4.5) \quad \hat{\chi}_p(X, Y) &:= \hat{\chi}_p(P_X, P_Y) \\
&= \inf \{ \chi_p(\tilde{X}, \tilde{Y}) : \tilde{X}, \tilde{Y} \in \mathfrak{B}(\mathbb{B}), \tilde{X} \stackrel{d}{=} X, \tilde{Y} \stackrel{d}{=} Y \}, p > 0; \\
\text{(iii) } \omega_{p,N}(X)^{p+1} &:= \sup_{t > N} t^p P(\|X\| > t), \\
\omega_p(X) &:= \omega_{p,0}(X), \\
N_p(X) &:= \{E \|X\|^p\}^{1/(1+p)} \quad p > 0.
\end{aligned}$$

Note that  $\hat{\chi}_p$  is a metric in  $\mathcal{L}(\mathbb{B})$ ,  $\hat{\chi}_p \geq \pi$  and the following convergence criterion holds: if

$$(4.6) \quad \omega_{p,N}(X_n) + \omega_{p,N}(X) \rightarrow 0 \text{ as } N \rightarrow \infty \text{ for any } n = 1, 2, \dots,$$

then

$$(4.7) \quad \hat{\chi}_p(X_n, X) \xrightarrow{n \rightarrow \infty} 0 \text{ iff } X_n \xrightarrow{n \rightarrow \infty}^{\mathcal{Q}} X \text{ and } \lim_{N \rightarrow \infty} \overline{\lim}_{n \rightarrow \infty} \omega_{p,N}(X_n) = 0.$$

Similarly, if (4.6) holds then

$$(4.8) \quad \chi_p(X_n, X) \xrightarrow{n \rightarrow \infty} 0 \text{ iff } X_n \xrightarrow{n \rightarrow \infty}^P X \text{ and } \lim_{N \rightarrow \infty} \overline{\lim}_{n \rightarrow \infty} \omega_{p,N}(X_n) = 0.$$

Theorem 4.1.

a) If for some  $p > 0$ ,  $N_p(Z_1) < 1$  and  $\omega_p(Y_1) < \infty$  then  $S_n$  converges in probability to a  $P$ -a.e. finite limit  $S = \sum_{i=1}^{\infty} X_i$ , and

$$(4.9) \quad \pi(S_n, S) \leq \chi_p(S_n, S) \leq \phi_n(Z_1) \omega_p(Y_1),$$

where  $\phi_n(Z_1) := N_p(Z_1)^{n+1} / (1 - N_p(Z_1))$ .

b) Under the above conditions, if  $Y_i^*$ 's are i.i.d. with  $\chi_p(Y_1, Y_1^*) < \infty$  then

$$(4.10) \quad S_n^* := \sum_{i=1}^n Y_i^* \prod_{j=1}^i Z_j \xrightarrow{P} S^*.$$

Moreover, suppose that  $\tilde{Z}_i$ 's are independent of  $Z_i$ 's and  $Y_i$ 's i.i.d. copies of  $Z_1$ . Then, as  $n \rightarrow \infty$ ,

$$(4.11) \quad \pi(S^* - S_n^*, \tilde{Z}_1 \cdots \tilde{Z}_n \cdot S) \leq \phi_n(Z_1) \hat{\chi}_p(Y_1, Y_1^*) \rightarrow 0.$$

Remark 4.1. Clearly, in (4.11) the factor  $\tilde{Z}_1 \cdots \tilde{Z}_n$  plays the same role as the normalizing scaling in the usual CLT.

Proof.

a) Since  $\chi_p \geq \kappa$  where  $\kappa$  is the distance in probability (the Ky Fan metric):

$$(4.12) \quad \kappa(X, Y) = \inf\{u > 0: P(\|X - Y\| > u) < u\},$$

then to prove  $S_n \xrightarrow{P} S$  it is enough to show that  $S_n$  is  $\chi_p$ -fundamental. In fact, for any  $k = 1, 2, \dots$ ,

$$(4.13) \quad \begin{aligned} \chi_p(S_{n+k}, S_n) &= \omega_p\left(\sum_{i=n+1}^{n+k} X_i\right) \leq \sum_{i=n+1}^{n+k} \omega_p(X_i) \\ &\leq \sum_{i=n+1}^{n+k} \left\{ \mathbb{E}_{Z_1=z_1, \dots, Z_n=z_n} \omega_p^{p+1}\left(Y_i \prod_{j=1}^i z_j\right) \right\}^{\frac{1}{1+p}} \\ &= \sum_{i=n+1}^{n+k} \left[ \mathbb{E}_{Z_1=z_1, \dots, Z_n=z_n} \left( \prod_{j=1}^i \|z_j\| \right)^p \right]^{\frac{1}{1+p}} \omega_p(Y_1) \\ &\leq \sum_{n>i} N_p(Z_1)^i \omega_p(Y_1) \\ &= N_p(Z_1)^{n+1} \frac{1}{1 - N_p(Z_1)} \omega_p(Y_1) \rightarrow 0 \text{ as } n \rightarrow \infty. \end{aligned}$$

Next, (4.9) follows by the same arguments as (4.13).

b) By (4.5),  $\hat{\chi}_p$  is a minimal metric with respect to  $\chi_p$  and thus, for any joint distribution of  $Y_1$  and  $Y_1^*$ ,

$$\hat{\chi}_p(S^* - S_n^*, \tilde{Z}_1 \cdots \tilde{Z}_n \cdot S) \leq \chi_p\left(\sum_{i>n} Y_i^* \prod_{j=1}^i Z_j, \sum_{i>n} Y_i \prod_{j=1}^i Z_j\right).$$

Now proceed as in (4.13) to obtain

$$\leq \chi_p(Y_1, Y_1^*) N_p(Z_1)^{n+1} \frac{1}{1-N_p(Z_1)}.$$

Taking the infimum is the last inequality over of all joint distributions of  $(Y_1, Y_1^*)$  with fixed marginals, we use  $\hat{\chi}_p \geq \pi$  to complete the proof of (4.11).  $\square$

Theorem 4.2. Suppose  $Y_i$ 's and  $Z_i$ 's are nonnegative r.f.'s. Then under the assumptions of Theorem 4.1 (a),  $M_n \xrightarrow{P} M$  and

$$\pi(M_n, M) \leq \phi_n(Z_1) \omega_p(Y_1) \rightarrow 0, \text{ as } n \rightarrow \infty.$$

Moreover if  $\chi_p(Y_1, Y_1^*) < \infty$  then

$$\pi\left(\bigvee_{i=n+1}^{\infty} Y_i, \prod_{j=1}^i Z_j, \tilde{Z}_1 \cdot \dots \cdot \tilde{Z}_n \cdot M\right) \leq \phi_n(Z_1) \hat{\chi}_p(Y_1, Y_1^*) \rightarrow 0.$$

Remark 4.2. If  $N$  is an integer valued r.v. independent of  $Y_i$ 's and  $Z_i$ 's then under the conditions of Theorem 4.1,

$$\pi(S_N, S) \leq \psi_N(Z_1) \omega_p(Z_1),$$

where  $\psi_N(Z_1) = (E N_p(Z_1)^N) / (1 - N_p(Z_1))$ , and moreover,

$$\pi(S^* - S_N^*, \tilde{Z}_1 \cdot \dots \cdot \tilde{Z}_N \cdot S) \leq \psi_N(Z_1) \hat{\chi}_p(Y_1, Y_1^*).$$

Similar results regarding the limit behavior of the distribution of the maximum  $\bigvee_{i=1}^N X_i$  can be obtained as a corollary to Theorem 4.1.

## 5. DISCUSSION OF THE LITERATURE

The equivalence of (bii) and (biv) in Lemma 2.1 has been proved by Vervaat (1979) (Lemmas 1.7, 1.8 and Corollary 1.9). Vervaat (1979) (Theorem 6.1) proved weak convergence of

$$(5.1) \quad S^{(n)} \stackrel{d}{=} Z^{1/n} S^{(n)} + B, \quad (Z, B) \text{ and } S^{(n)} \text{ independent}$$

to the normal law under conditions similar to that of Theorem 2.1(b) ( $\alpha = 2$ ). The equation  $S \stackrel{d}{=} (S + Y) Z$  with  $Z \in \mathcal{Z}_1$ , has been studied by Chandrasekhar and Munch (1950), Takacz (1953), Weiss (1973), Yeo (1974), Vervaat (1974), Chamayou and Schorr (1975) who proved particular cases of Theorem 3.1. A weaker version of Lemma 3.2 was proved by Vervaat (1979) p. 764.

The class  $\mathcal{F}_2$  is also known as the class of self-decomposable r.v.'s, see Lukacz (1970, Section 5.11). The random sums

$$(5.2) \quad S_n = \sum_{k=1}^n \left( \prod_{j=k+1}^n Z_j \right) Y_k + \left( \prod_{j=1}^n Z_j \right) Y_0$$

where  $Y_0$  is independent of the  $\mathbb{R}^2$ -valued  $(Y_j, Z_j)$  are of interest in economics (see de Haan, Resnick, Rootzen and Vries (1989)), physics (Chamayou (1973)), environmental processes (Solomon (1975)), sociology (Cavalli-Sforza and Feldman (1973)). Excellent reviews are given by Kesten (1973) and Vervaat (1979), more references can be found in de Haan, Resnick, Rootzen and Vries (1979) and de Haan and Karandikar (1979). It will be of future interest to obtain a set of limit results as in Section 2 for the sums (5.2) under mild condition on the joint distribution of  $(Y_1, Z_1)$ .

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