

## CONVOLUTIONS OF STABLE LAWS AS LIMIT DISTRIBUTIONS OF PARTIAL SUMS<sup>1</sup>

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**0. Summary.** Let  $\{X_n\}$  be a sequence of independent random variables and let  $Y_n = B_n^{-1} \sum_{i=1}^n X_i - A_n$  be a sequence of normed, centered sums such that, for appropriately chosen normalizing coefficients ( $B_n \rightarrow \infty$ ) and centering constants,  $\{Y_n\}$  converges in law to a nondegenerate limit distribution  $G$ . B. V. Gnedenko asked the following question: What characterizes the class of limit distributions  $\{G\}$  when there are  $r$  different distribution functions among those of the random variables  $\{X_n\}$ ?

Let  $\mathcal{P}_r$  denote this class of distribution functions. As is well-known,  $\mathcal{P}_1$  is the class of stable distributions. V. M. Zolotarev and V. S. Korolyuk [8] have shown that  $\mathcal{P}_2$  consists solely of stable distributions and convolutions of two stable distributions. It was thought that O. K. Lebedintseva [4] had shown that this was true for  $r > 2$  (with *two* replaced by *less than or equal to  $r$* ) with the added hypothesis that one of the  $r$  possible distribution functions of the summands  $X_n$  belongs either to the domain of attraction of a stable distribution, or to a domain of partial attraction of only one type. However, V. M. Zolotarev and V. S. Korolyuk [8] gave an example that showed that O. K. Lebedintseva's theorem did not completely settle the matter. However, A. A. Zinger [7] gave a necessary and sufficient condition on the Lévy spectral function of  $G$  in order that  $G$  be in  $\mathcal{P}_r$ . His theorem shows that Lebedintseva's result is incorrect.

In this same paper, A. A. Zinger proved a theorem that gives a necessary condition on the distribution functions of the summands  $X_n$  in order that  $G$  be a convolution of  $r$  distinct stable distributions. Here we expand Zinger's theorem to obtain a necessary and sufficient condition that  $G$  be a convolution of  $r$  distinct stable distributions. Some related results are also obtained.

**1. Main results.** In this section, we state results connected with B. V. Gnedenko's conjecture on limit distribution functions of normed, centered sums of sequences of independent random variables when there are  $r$  different distribution functions among those of the sequence. The proofs are given in Section 3.

A distribution function  $F$  being in the domain of attraction of a stable distribution of characteristic exponent  $\alpha$ , denoted by  $F \in \mathcal{D}(\alpha)$ , means that if  $\{X_n\}$  is a sequence of independent identically distributed random variables with common

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Received May 31, 1968.

<sup>1</sup> This research is from the author's doctoral dissertation submitted to the Department of Mathematics, University of California, Riverside. It was supported in part by the Air Force Office of Scientific Research, Grant Nos. AF-AFOSR 851-65 and 851-66, and by the U.S. Army Research Office, Grant No. DA-ARO(D)-31-124-G383.

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distribution function  $F$ , then there exist a sequence of positive constants  $\{B_n\}$ , called *normalizing coefficients*, with  $B_n \rightarrow \infty$ , and a sequence of real numbers  $\{A_n\}$ , called *centering constants*, such that the limiting distribution of  $B_n^{-1}(X_1 + \cdots + X_n) - A_n$  is a stable distribution with characteristic exponent  $\alpha$ . Lamperti [3] essentially proved the “if” part and Tucker [5] proved the “converse” part of the following useful lemma.

LEMMA 1. *If  $F \in \mathcal{D}(\alpha)$ ,  $0 < \alpha \leq 2$ , and if  $\{B_n\}$  is a sequence of normalizing coefficients for  $F$ , then there is a measurable slowly varying function  $L$  defined over  $(0, \infty)$ , which must be asymptotic to a non-decreasing function when  $\alpha = 2$ , such that  $B_n \sim n^{\alpha-1}L(n)$ . Conversely, if  $L$  is a measurable slowly varying function over  $(0, \infty)$ , and if  $0 < \alpha < 2$ , or if  $\alpha = 2$  and  $L$  is a measurable slowly varying function asymptotic to a non-decreasing function, then there is an  $F \in \mathcal{D}(\alpha)$  such that  $\{n^{\alpha-1}L(n)\}$  is a sequence of normalizing coefficients for  $F$ .*

For the rest of this section, we will let  $\{X_n\}$  be a sequence of independent random variables such that the distribution function of  $X_n$  is one of  $F_1, \dots, F_r$ , with  $r \geq 2$ , for all  $n$ . We also assume that there exist normalizing coefficients  $\{B_n\}$  and centering constants  $\{A_n\}$  such that  $B_n^{-1}(X_1 + \cdots + X_n) - A_n$  converges in law to a non-degenerate distribution function  $G$ .

For each positive integer, let  $n_i(n)$  denote the number of random variables among  $X_1, \dots, X_n$  which have  $F_i$  as their distribution function,  $1 \leq i \leq r$ . We assume that  $n_i(n) \rightarrow \infty$  as  $n \rightarrow \infty$  for each  $i$ .

THEOREM 1. *If  $F_i \in \mathcal{D}(\lambda_i)$ ,  $1 \leq i \leq r$ , with  $0 < \lambda_1 < \cdots < \lambda_r \leq 2$ , if there is a  $k < r$  such that  $\limsup_{n \rightarrow \infty} (n_k(n)/n) = s_k > 0$  and  $\lim_{n \rightarrow \infty} (n_i(n)/n) = 0$  for  $i < k$ , then  $G$  is the convolution of not more than  $k$  stable distributions with characteristic exponents in the set  $\{\lambda_1, \dots, \lambda_k\}$ .*

Zinger ([7], Theorem 3) has proved the following theorem.

THEOREM (Zinger). *If  $G$  is a convolution of  $r$  stable distributions with characteristic exponents  $\lambda_i$  such that  $0 < \lambda_1 < \cdots < \lambda_r \leq 2$ , then under some permutation of the indices of the  $F$ 's, we have  $F_i \in \mathcal{D}(\lambda_i)$ ,  $1 \leq i \leq r$ .*

Now, in the statement of our principal theorem, two other conclusions are added to the conclusion of Zinger's theorem, which together are not only necessary but also sufficient.

THEOREM 2. *Let  $\{X_n\}$  be independent random variables with distribution functions among  $\{F_1, \dots, F_r\}$  such that  $\lim_{n \rightarrow \infty} n_i(n) = \infty$  for all  $i$ . Let  $\{B_n\}$  and  $\{A_n\}$  be such that  $0 < B_n \rightarrow \infty$  and  $B_n^{-1}(X_1 + \cdots + X_n) - A_n$  converges in law to  $G$ , where  $G$  is a non-degenerate distribution function. Then,  $G$  is a convolution of  $r$  stable distributions with characteristic exponents  $\lambda_1, \dots, \lambda_r$ , ( $0 < \lambda_1 < \cdots < \lambda_r \leq 2$ ), if and only if*

- (i)  $F_i \in \mathcal{D}(\lambda_i)$  for  $1 \leq i \leq r$ , (possibly requiring a permutation of the indices of the  $F$ 's);
- (ii)  $\lim_{n \rightarrow \infty} n_r(n)/n = 1$ ;
- (iii)  $\lim_{n \rightarrow \infty} B(i, n_i(n))/B(j, n_j(n)) = P_{ij}$ , with  $0 < P_{ij} < \infty$ , for all  $i, j \in \{1, \dots, r\}$ ,

where  $\{B(i, n)\}$  are normalizing coefficients for  $F_i$ ,  $1 \leq i \leq r$ . Further, in this case, the normalizing coefficients  $\{B_n\}$  can be chosen to be normalizing coefficients for  $F_r$ .

**2. Further results and some examples.** One might question whether or not the existence of a non-degenerate distribution function forces  $n_i(n)/n$  to converge. The following example shows that this need not occur.

Let  $0 < \lambda_1 < \lambda_2 \leq 2$ . Let  $F_1$  and  $F_2$  be symmetric distribution functions defined by, for  $i = 1, 2$ ,

$$\begin{aligned} F_i(x) &= 1 - x^{-\lambda_i}, & x \geq 2^{\lambda_i-1}, \\ &= \frac{1}{2}, & 0 \leq x \leq 2^{\lambda_i-1}. \end{aligned}$$

Let  $\#\{ \}$  denote the number of elements in the set defined within the braces. Define a function  $e(\cdot)$  on the positive integers as follows:

$$\begin{aligned} e(1) &= 1, & e(2) &= 2, & \text{and for } n > 2 \\ e(n) &= 1 & \text{if } e(n-1) = 2 & \text{ and } \#\{i: 1 \leq i \leq n-1, e(i) = 1\}/(n-1) = \frac{1}{2}, \\ &= 1 & \text{if } e(n-1) = 1 & \text{ and } \#\{i: 1 \leq i \leq n-1, e(i) = 1\}/(n-1) < \frac{3}{4}, \\ &= 2 & \text{if } e(n-1) = 2 & \text{ and } \#\{i: 1 \leq i \leq n-1, e(i) = 1\}/(n-1) > \frac{1}{2}, \\ &= 2 & \text{if } e(n-1) = 1 & \text{ and } \#\{i: 1 \leq i \leq n-1, e(i) = 1\}/(n-1) = \frac{3}{4}. \end{aligned}$$

In other words,  $e(n)$  is 1 until the proportion of ones assigned reaches  $\frac{3}{4}$ , then  $e(n)$  is 2 until the proportion of ones reaches  $\frac{1}{2}$  at which time  $e(n)$  becomes 1 again and we repeat the process.

Let  $\{X_n\}$  be a sequence of independent random variables such that  $F_{X_n} = F_{e(n)}$  for all  $n$ . Then  $\limsup_{n \rightarrow \infty} n_1(n)/n = \frac{3}{4}$  and  $\liminf_{n \rightarrow \infty} n_1(n)/n = \frac{1}{2}$ .

Let  $B_n = (n_1(n))^{\lambda_1-1}$  for all  $n$ .

According to Gnedenko and Kolmogorov ([2], 124, Theorem 4), in order to show that there exists  $\{A_n\}$  such that  $B_n^{-1}(X_1 + \cdots + X_n) - A_n$  converges in law to a stable distribution with characteristic exponent  $\lambda_1$ , it is sufficient to show that, for any  $x > 0$ ,

- (i)  $n_1(n)F_1(-B_n x) + n_2(n)F_2(-B_n x) \rightarrow x^{-\lambda_1}$ ,
- (ii)  $n_1(n)(1 - F_1(B_n x)) + n_2(n)(1 - F_2(B_n x)) \rightarrow x^{-\lambda_1}$ ,
- (iii)  $\lim_{\varepsilon \rightarrow 0+} \limsup_{n \rightarrow \infty} D_n(\varepsilon) = \lim_{\varepsilon \rightarrow 0+} \liminf_{n \rightarrow \infty} D_n(\varepsilon) = 0$ ,

where

$$D_n(\varepsilon) = \sum_{i=1}^2 n_i(n) \left\{ \int_{|x| < \varepsilon} x^2 dF_i(B_n x) - \left( \int_{|x| < \varepsilon} x dF_i(B_n x) \right)^2 \right\}.$$

By the definition of  $B_n$ , we have that  $n_1(n)/B_n^{\lambda_1} \rightarrow 1$  and  $n_2(n)/B_n^{\lambda_2} \leq (n/2)^{1-\lambda_2/\lambda_1} \rightarrow 0$ .

Hence, for  $x > 0$ ,  $n_1(n)F_1(-B_n x) + n_2(n)F_2(-B_n x) \rightarrow x^{-\lambda_1}$  as  $n \rightarrow \infty$ . Therefore, (i) holds, and by the symmetry involved, (ii) also holds.

To show that (iii) holds, we note that  $\int_{|x| < \varepsilon} x dF_i(B_n x) = 0$ , for  $i = 1, 2$ . An easy calculation shows that  $\lim_{n \rightarrow \infty} D_n(\varepsilon) = 2\lambda_1 \varepsilon^{2-\lambda_1} (2-\lambda_1)^{-1}$ .

Another question we ask is: given  $0 < \lambda_1 < \cdots < \lambda_r < 2$ , does there exist a sequence  $\{X_n\}$  of independent random variables containing  $r$  distributions such that, for some sequences  $\{B_n\}$  and  $\{A_n\}$ ,  $B_n^{-1}(X_1 + \cdots + X_n) - A_n$  converges in law, as  $n \rightarrow \infty$ , to a convolution of stable distributions with exponents  $\lambda_1, \dots, \lambda_r$ ? Zinger ([7], 620–621) has suggested a method to generate more general examples. Following his suggestion we define  $r$  distribution functions by

$$\begin{aligned} F_i(x) &= |x|^{-\lambda_i} \quad \text{for } x \leq -2^{-\lambda_i-1}, \\ &= \frac{1}{2} \quad \text{for } |x| \leq 2^{-\lambda_i-1}, \\ &= 1 - x^{-\lambda_i} \quad \text{for } x \geq 2^{-\lambda_i-1} \end{aligned}$$

For  $n$  a positive integer, define  $B_n$  to be the smallest positive solution of  $y^{\lambda_1} + \cdots + y^{\lambda_r} = n$ .

For  $1 \leq i \leq r-1$ , let  $n_i(n) = [B_n^{\lambda_i}]$ , and let  $n_r(n) = n - \sum_{i=1}^{r-1} n_i(n)$ , where  $[a]$  denotes the integral part of the real number  $a$ . Since  $n_i(n)$  is integer-valued, non-decreasing, and  $\sum_{i=1}^r n_i(n+1) - \sum_{i=1}^r n_i(n) = n+1 - n = 1$ , we have that exactly one of the  $n_i(n)$ s increases by one when the argument increases from  $n$  to  $n+1$ . Let  $e(n)$  be the index  $i$  such that  $n_i(n)$  increases when the argument goes from  $n-1$  to  $n$ .

Let  $\{X_n\}$  be a sequence of independent random variables such that  $F_{X_n}$  is  $F_{e(n)}$ .

In order to show that, for appropriately chosen constants  $\{A_n\}$ ,  $B_n^{-1}(X_1 + \cdots + X_n) - A_n$  converges in law to a convolution of  $r$  stable distribution with characteristic exponents  $\lambda_1, \dots, \lambda_r$ , it is sufficient to show that, for  $x > 0$ ,

- (i)  $\sum_{i=1}^r n_i(n) F_i(-B_n x) \rightarrow \sum_{i=1}^r x^{-\lambda_i}$ ,
- (ii)  $\sum_{i=1}^r n_i(n) (1 - F_i(B_n x)) \rightarrow \sum_{i=1}^r x^{-\lambda_i}$ ,
- (iii)  $\lim_{\varepsilon \rightarrow 0+} \limsup_{n \rightarrow \infty} D_n(\varepsilon) = \lim_{\varepsilon \rightarrow 0+} \liminf_{n \rightarrow \infty} D_n(\varepsilon) = 0$

where

$$D_n(\varepsilon) = \sum_{i=1}^r n_i(n) \left\{ \int_{|x| < \varepsilon} x^2 dF_i(B_n x) - \left( \int_{|x| < \varepsilon} x dF_i(B_n x) \right)^2 \right\}.$$

For  $1 \leq i \leq r-1$ ,  $n_i(n)/B_n^{\lambda_i} \rightarrow 1$  as  $n \rightarrow \infty$ . Letting  $\{a\}$  denote the fractional part of the real number  $a$ , we have that

$$n_r(n)/B_n^{\lambda_r} = \left( \sum_{i=1}^r \{B_n^{\lambda_i}\} + [B_n^{\lambda_r}] \right) / B_n^{\lambda_r}.$$

The right-hand side is bounded above by  $(r + [B_n^{\lambda_r}])B_n^{-\lambda_r}$  and is bounded below by  $[B_n^{\lambda_r}]B_n^{-\lambda_r}$ , both of which converge to unity as  $n \rightarrow \infty$ . This establishes (i) and (ii).

It can be shown that  $\lim_{n \rightarrow \infty} D_n(\varepsilon) = \sum_{i=1}^r 2\lambda_i \varepsilon^{2-\lambda_i} (2-\lambda_i)^{-1}$ . Hence, (iii) holds and the posed question is answered in the affirmative.

V. M. Zolotarev and V. S. Korolyuk [8] gave an example that shows that the limit distribution being stable does not imply that at least one of the distribution functions  $F_1$  and  $F_2$  of the summands belongs to the domain of attraction, or even the domain of partial attraction, of any distribution. However, there are some errors in their example:  $c_1$  should be  $c_2$  and  $c_2$  should be  $c_1$ ;  $n_1$  should be  $[n/3]$  and  $n_2$  should be  $2[n/3] + 3\{n/3\}$ .

Zinger [6] gave an example when  $r = 3$  and the limit distribution is neither a stable distribution nor a convolution of any number of stable distributions.

We will conclude this section with three theorems and their corollaries. For the rest of this section we assume that  $\{X_n\}$  is a sequence of independent random variables such that the distribution function of  $X_n$  is one of  $F_1, \dots, F_r$ , and  $F_i \in \mathcal{D}(\lambda_i)$  for  $1 \leq i \leq r$ , with  $0 < \lambda_1 < \lambda_2 < \lambda_3 \leq \dots \leq \lambda_r \leq 2$ . We also assume that there exist sequences  $\{B_n\}$  and  $\{A_n\}$  such that  $B_n^{-1}(X_1 + \dots + X_n) - A_n$  converges in law to a non-degenerate distribution function  $G$ .

**THEOREM 3.** *If, for some  $\eta \in (0, 1 - \lambda_1/\lambda_2)$ ,  $\limsup_{n \rightarrow \infty} (n_1(n)/n^{1-\eta}) = s_1$ , with  $0 < s_1 < \infty$ , then  $G$  is a stable distribution with characteristic exponent  $\lambda_1$ .*

It can occur, even though each of the  $r$  distribution functions  $F_1, \dots, F_r$  belongs to the domain of attraction of a stable distribution with characteristic exponent from the set  $\{\lambda_1, \dots, \lambda_r\}$ , that the normalizing coefficients are not normalizing coefficients for any distribution function in the domain of attraction of a stable distribution with characteristic exponent from the set  $\{\lambda_1, \dots, \lambda_r\}$ . This follows from the following Corollary and Lemma 1.

**COROLLARY TO THEOREM 3.** *If, for some  $\eta \in (0, 1 - \lambda_1/\lambda_2)$ ,  $\lim_{n \rightarrow \infty} (n_1(n)/n^{1-\eta}) = s_1$ , with  $0 < s_1 < \infty$ , if, for  $i > 1$ ,  $\lim_{n \rightarrow \infty} (n_i(n)/n) = s_i > 0$ , then  $G$  is a stable distribution with characteristic exponent  $\lambda_1$  and  $B_n \sim n^{\lambda^{-1}} L(n)$ , where  $\lambda = \lambda_1/(1-\eta)$  and  $L$  is a measurable slowly varying function defined over  $(0, \infty)$ .*

**THEOREM 4.** *If there exists  $\eta > 1 - \lambda_1/\lambda_2$  such that  $\lim_{n \rightarrow \infty} (n_1(n)/n^{1-\eta}) = s_1$ , with  $0 < s_1 < \infty$ , and if  $\limsup_{n \rightarrow \infty} (n_2(n)/n) = s_2 > 0$ , then  $G$  is a stable distribution with characteristic exponent  $\lambda_2$ .*

**COROLLARY TO THEOREM 4.** *If there exists  $\eta > 1 - \lambda_1/\lambda_2$  such that  $\lim_{n \rightarrow \infty} (n_1(n)/n^{1-\eta}) = s_1$ , with  $0 < s_1 < \infty$ , and if  $\lim_{n \rightarrow \infty} (n_i(n)/n) = s_i > 0$  for all  $i > 1$ , then  $G$  is a stable distribution with characteristic exponent  $\lambda_2$  and normalizing coefficients of  $F_2$  can be used for the sequence  $\{B_n\}$ .*

**THEOREM 5.** *If  $n_1(n)/n^{\lambda_1/\lambda_2} \rightarrow s_1$ , with  $0 < s_1 < \infty$ , if  $n_2(n)/n \rightarrow s_2 > 0$ , then*

- (i)  $G$  is a stable distribution with characteristic exponent  $\lambda_1$  if and only if  $L_2(n)/L_1(n^{\lambda_1/\lambda_2}) \rightarrow 0$  as  $n \rightarrow \infty$ ;
- (ii)  $G$  is a stable distribution with characteristic exponent  $\lambda_2$  if and only if  $L_2(n)/L_1(n^{\lambda_1/\lambda_2}) \rightarrow \infty$  as  $n \rightarrow \infty$ ;
- (iii)  $G$  is a convolution of two stable distributions with characteristic exponents  $\lambda_1$  and  $\lambda_2$  if and only if  $L_2(n)/L_1(n^{\lambda_1/\lambda_2})$  is bounded away from zero and infinity.

We should note that there are many assumptions made before (i), (ii) and (iii) of Theorem 5. In particular, not all distribution functions in  $\mathcal{D}(\lambda_1)$  and  $\mathcal{D}(\lambda_2)$  can be paired and satisfy the hypothesis of Theorem 5. For example, Tucker [5] gives an example of two non-decreasing slowly varying functions  $L_1$  and  $L_2$  such that  $\limsup_{x \rightarrow \infty} L_1(x)/L_2(x) = \infty$  and  $\liminf_{x \rightarrow \infty} L_1(x)/L_2(x) = 0$ . Let  $L_1'(x) = L_1(x^{\lambda_2/\lambda_1})$ . Then  $L_1'$  is a slowly varying function. Also,  $\limsup_{x \rightarrow \infty} L_1'(x^{\lambda_1/\lambda_2})/L_2(x) = \infty$  and  $\liminf_{x \rightarrow \infty} L_1'(x^{\lambda_1/\lambda_2})/L_2(x) = 0$ . By the converse of Lemma 1, there are  $F_1 \in \mathcal{D}(\lambda_1)$

and  $F_2 \in \mathcal{D}(\lambda_2)$  such that  $\{n^{\lambda_1-1}L_1'(n)\}$  and  $\{n^{\lambda_2-1}L_2(n)\}$  are normalizing coefficients for  $F_1$  and  $F_2$ , respectively. If a sequence  $\{X_n\}$  of independent random variables exists such that  $F_{X_n}$  is either  $F_1$  or  $F_2$ , and if there are normalizing coefficients and centering constants so that the normed, centered sums of the  $X_n$ s converge in law, then we must have either  $n_1(n)/n^{\lambda_1/\lambda_2} \rightarrow 0$  or  $\infty$  as  $n \rightarrow \infty$ . Otherwise, Theorem 5 would apply, which obviously would be a contradiction to our choice of  $L_1'$  and  $L_2$ .

### 3. Proofs.

PROOF OF THEOREM 1. Let  $X(i, m)$  be the  $m$ th random variable in the sequence  $\{X_n\}$  whose distribution function is  $F_i$ , for  $1 \leq i \leq r$ . Then, for each  $i$ , there are constants  $\{B(i, n)\}$  and  $\{A(i, n)\}$  such that

$$[B(i, n)]^{-1}[X(i, 1) + \cdots + X(i, n)] - A(i, n)$$

converges in law to a stable distribution with characteristic exponent  $\lambda_i$ .

By Lemma 1, for each  $i$ , there exists a measurable slowly varying function  $L_i$  defined over  $(0, \infty)$  such that  $B(i, n) \sim n^{\lambda_i-1}L_i(n)$ .

Let  $\{m(j)\}$  be an increasing sequence of positive integers such that  $\lim_{j \rightarrow \infty} (n_i(m(j))/m(j))$  exists and is equal to  $s_i$ , with  $s_i > 0$ , for  $1 \leq i \leq r$ .

Let

$$f_i(j) = \left( \frac{n_i(m(j))}{(m(j))^{\lambda_i/\lambda_k}} \right)^{1/\lambda_i} \frac{L_i(n_i(m(j)))}{L_k(m(j))}, \quad \text{for } 1 \leq i \leq k-1.$$

Now, select an increasing sequence  $\{j(p)\}$  of positive integers such that  $\lim_{p \rightarrow \infty} f_i(j(p))$  exists and is equal to  $\alpha_i$ , for  $1 \leq i \leq k-1$ . It is possible that some  $\alpha_i$  may be  $\infty$ . Let  $a(p) = m(j(p))$  for all  $p$ . Then, for  $i > k$ ,  $B(i, n_i(a(p)))/B(k, n_k(a(p))) \rightarrow 0$  as  $p \rightarrow \infty$ .

Let

$$f(i, p) = \left( \frac{n_i(a(p))}{(a(p))^{\lambda_i/\lambda_k}} \right)^{1/\lambda_i} \frac{L_i(n_i(a(p)))}{L_k(a(p))}, \quad \text{for } 1 \leq i \leq k-1.$$

Then  $\lim_{p \rightarrow \infty} f(i, p) = \alpha_i$ ,  $1 \leq i \leq k-1$ .

The proof will be concluded by considering three exhaustive cases.

CASE 1.  $\alpha_i < \infty$  for all  $i$ . Then for  $1 \leq i \leq k-1$ , we have

$$\frac{B(i, n_i(a(p)))}{B(k, n_k(a(p)))} \sim s_k^{-1/\lambda_k} f(i, p) \rightarrow s_k^{-1/\lambda_k} \alpha_i, \quad \text{as } p \rightarrow \infty.$$

We observe that

$$\begin{aligned} & \frac{X_1 + \cdots + X_{a(p)}}{B(k, n_k(a(p)))} - \frac{1}{B(k, n_k(a(p)))} \sum_{i=1}^r B(i, n_i(a(p))) \cdot A(i, n_i(a(p))) \\ &= \frac{X(k, 1) + \cdots + X(k, n_k(a(p)))}{B(k, n_k(a(p)))} - A(k, n_k(a(p))) \\ &+ \sum_{i=1, i \neq k}^r \frac{B(i, n_i(a(p)))}{B(k, n_k(a(p)))} \left\{ \frac{X(i, 1) + \cdots + X(i, n_i(a(p)))}{B(i, n_i(a(p)))} - A(i, n_i(a(p))) \right\}. \end{aligned}$$

We see that the right-hand side of the above equality converges in law to a convolution of not more than  $k$  stable laws with characteristic exponents in the set  $\{\lambda_1, \dots, \lambda_k\}$ . Hence,

$$\frac{X_1 + \dots + X_{a(p)}}{B(k, n_k(a(p)))} - \frac{1}{B(k, n_k(a(p)))} \sum_{i=1}^r B(i, n_i(a(p))) A(i, n_i(a(p)))$$

converges in law, as  $p \rightarrow \infty$ , to a convolution of not more than  $k$  stable laws with characteristic exponents in the set  $\{\lambda_1, \dots, \lambda_k\}$ , including  $\lambda_k$ .

Since  $B_{a(p)}^{-1}[X_1 + \dots + X_{a(p)}] - A_{a(p)}$  converges in law, as  $p \rightarrow \infty$ , to  $G$ , we have that  $G$  is of this same type, which establishes the theorem for Case 1.

CASE 2. For exactly one  $j \in \{1, \dots, k-1\}$ ,  $\alpha_j = \infty$  and  $\alpha_i < \infty$  for  $i \in \{1, \dots, k-1\} \setminus \{j\}$ .

Then, for  $i \in \{1, \dots, k-1\} \setminus \{j\}$ ,

$$B(i, n_i(a(p)))/B(j, n_j(a(p))) \sim f(i, p)/f(j, p) \rightarrow 0, \quad \text{as } p \rightarrow \infty.$$

Also,

$$B(k, n_k(a(p)))/B(j, n_j(a(p))) \sim s_k^{\lambda_k^{-1}}/f(j, p) \rightarrow 0, \quad \text{as } p \rightarrow \infty.$$

Continuing as in Case 1, but dividing by  $B(j, n_j(a(p)))$  instead of  $B(k, n_k(a(p)))$ , and noting that for  $i > k$ ,

$$\frac{B(i, n_i(a(p)))}{B(j, n_j(a(p)))} = \frac{B(i, n_i(a(p)))/B(k, n_k(a(p)))}{B(j, n_j(a(p)))/B(k, n_k(a(p)))} \rightarrow 0,$$

as  $p \rightarrow \infty$ , we can conclude that  $G$  is a stable distribution with characteristic exponent  $\lambda_j \in \{\lambda_1, \dots, \lambda_k\}$ . This establishes the theorem for Case 2.

CASE 3.  $\alpha_{i_s} = \infty$  for  $s \in \{1, \dots, u\}$  with  $u \geq 2$  and  $\alpha_i < \infty$  for  $i \in \{1, \dots, k-1\} \setminus \{i_1, \dots, i_u\}$ .

Assume  $i_1 < \dots < i_u$ . Let  $\{p(j)\}$  be an increasing sequence of positive integers such that  $\lim_{j \rightarrow \infty} f(i_s, p(j))/f(i_u, p(j))$  exists and is equal to  $\beta_s$ , for  $1 \leq s \leq u-1$ .

(i) If  $\beta_s < \infty$  for all  $s$ , then, for  $i \in \{i_1, \dots, i_{u-1}\}$ ,

$$\frac{B(i, n_i(a(p(j))))}{B(i_u, n_{i_u}(a(p(j))))} \sim \frac{f(i, p(j))}{f(i_u, p(j))} \rightarrow \beta_i < \infty.$$

For  $i \in \{1, \dots, k-1\} \setminus \{i_1, \dots, i_u\}$ ,

$$\frac{B(i, n_i(a(p(j))))}{B(i_u, n_{i_u}(a(p(j))))} \sim \frac{f(i, p(j))}{f(i_u, p(j))} \rightarrow 0,$$

since  $\alpha_i < \infty$  and  $\alpha_{i_u} = \infty$ . Also

$$\frac{B(k, n_k(a(p(j))))}{B(i_u, n_{i_u}(a(p(j))))} \sim \frac{s_k^{\lambda_k^{-1}}}{f(i_u, p(j))} \rightarrow 0.$$

As at the end of Case 2, for  $i > k$ ,

$$B(i, n_i(a(p(j))))/B(i_u, n_{i_u}(a(p(j)))) \rightarrow 0, \quad \text{as } j \rightarrow \infty.$$

Continuing as in Case 1, but dividing by  $B(i_u, n_{i_u}(a(p(j))))$  instead of  $B(k, n_k(a(p)))$  and summing the  $X$ s to  $a(p(j))$  instead of  $a(p)$ , we can conclude that  $G$  is a convolution of not more than  $u < k$  stable distributions with characteristic exponents in the set  $\{\lambda_{i_1}, \dots, \lambda_{i_u}\} \subset \{\lambda_1, \dots, \lambda_k\}$ .

(ii) If exactly one of the  $\beta$ s, say  $\beta_s$ , is infinite, then, for  $i \in \{i_1, \dots, i_u\} \setminus \{i_s\}$ ,

$$\frac{B(i, n_i(a(p(j))))}{B(i_s, n_{i_s}(a(p(j))))} \sim \frac{f(i, p(j))/f(i_u, p(j))}{f(i_s, p(j))/f(i_u, p(j))} \rightarrow 0.$$

For  $i \in \{1, \dots, k-1\} \setminus \{i_1, \dots, i_u\}$ ,

$$B(i, n_i(a(p(j))))/B(i_s, n_{i_s}(a(p(j)))) \sim f(i, p(j))/f(i_s, p(j)) \rightarrow 0,$$

since  $\alpha_i < \infty$  and  $\alpha_{i_s} = \infty$ .

Again  $B(k, n_k(a(p(j))))/B(i_s, n_{i_s}(a(p(j)))) \sim s_k^{\lambda_k - 1}/f(i_s, p(j)) \rightarrow 0$ .

As in (i), for  $i > k$ ,

$$B(i, n_i(a(p(j))))/B(i_s, n_{i_s}(a(p(j)))) \rightarrow 0, \quad \text{as } j \rightarrow \infty.$$

Now continuing as in (i), except dividing by  $B(i_s, n_{i_s}(a(p(j))))$ , we can conclude that  $G$  is a stable distribution with characteristic exponent  $\lambda_{i_s} \in \{\lambda_1, \dots, \lambda_k\}$ .

(iii) If more than one of the  $\beta$ s, say  $\beta_{i_1}, \dots, \beta_{i_{t_w}}$ , is infinite, then repeat the above process by forming the ratios  $f(i, a(p(j)))/f(i_{t_w}, a(p(j)))$  for  $i \in \{i_1, \dots, i_{t_w-1}\}$ . Continuing this process as long as necessary, we will eventually arrive at one of two possible stages.

*Stage 1.* We have a subset  $\{f(v_1, \cdot), \dots, f(v_t, \cdot)\}$  of  $\{f(1, \cdot), \dots, f(k-1, \cdot)\}$  with  $t \geq 2$  and a subsequence  $\{p''(j)\}$  of  $\{p(j)\}$  such that, for  $i \in \{v_1, \dots, v_{t-1}\}$ ,

$$\lim_{j \rightarrow \infty} f(i, a(p''(j)))/f(v_t, a(p''(j)))$$

exists and is finite.

Then, for  $i \in \{1, \dots, k-1\} \setminus \{v_1, \dots, v_t\}$ ,  $f(i, \cdot) \notin \{f(v_1, \cdot), \dots, f(v_t, \cdot)\}$  for one of the three following reasons:

(i)  $\alpha_i < \infty$ ;

(ii) there exists an  $f(z, \cdot) \in \{f(1, \cdot), \dots, f(k-1, \cdot)\}$  and there exists a subsequence  $\{p'(j)\}$  of  $\{p(j)\}$  containing  $\{p''(j)\}$  such that

$$\lim_{j \rightarrow \infty} f(i, a(p'(j)))/f(z, a(p'(j))) \text{ is finite and}$$

$$\lim_{j \rightarrow \infty} f(v_t, a(p'(j)))/f(z, a(p'(j))) = \infty;$$

(iii)  $\lim_{j \rightarrow \infty} f(v_t, a(p'(j)))/f(i, a(p'(j))) = \infty$ , for some subsequence  $\{p'(j)\}$  of  $\{p(j)\}$  containing  $\{p''(j)\}$ .

In each of these three cases, we have

$$\lim_{j \rightarrow \infty} f(i, a(p''(j)))/f(v_t, a(p''(j))) = 0.$$



Hence, for  $i \in \{1, \dots, k-1\} \setminus \{v_1, \dots, v_t\}$ ,

$$B(i, n_i(a(p''(j))))/B(v_t, n_{v_t}(a(p''(j)))) \rightarrow 0 \quad \text{as } j \rightarrow \infty.$$

Clearly, for  $i \geq k$ , we also have convergence to zero.

Continuing as always, we can conclude that  $G$  is a convolution of not more than  $t < k$  stable distributions with characteristic exponents in the set  $\{\lambda_{v_1}, \dots, \lambda_{v_t}\} \subset \{\lambda_1, \dots, \lambda_k\}$ .

*Stage 2.* We have a subset  $\{f(v_1, \cdot), \dots, f(v_t, \cdot)\}$  of  $\{f(1, \cdot), \dots, f(k-1, \cdot)\}$  with  $t \geq 2$  and we have a sequence  $\{p'(j)\}$  of  $\{p(j)\}$  such that for some  $i_* \in \{v_1, \dots, v_{t-1}\}$

$$\lim_{j \rightarrow \infty} f(i_*, a(p'(j)))/f(v_t, a(p'(j))) = \infty$$

and for  $i \in \{v_1, \dots, v_{t-1}\} \setminus \{i_*\}$

$$\lim_{j \rightarrow \infty} f(i, a(p'(j)))/f(v_t, a(p'(j))) \quad \text{exists and is finite.}$$

As in Stage 1, we have for  $i \in \{1, \dots, k-1\} \setminus \{v_1, \dots, v_t\}$ ,  $\lim_{j \rightarrow \infty} f(i, a(p'(j)))/f(i_*, a(p'(j))) = 0$ . Hence, for  $i \in \{1, \dots, k-1\} \setminus \{v_1, \dots, v_t\}$ ,

$$B(i, n_i(a(p'(j))))/B(i_*, n_{i_*}(a(p'(j)))) \rightarrow 0 \quad \text{as } j \rightarrow \infty.$$

Clearly, for  $i \geq k$ , the above convergence holds.

Since, for  $i \in \{v_1, \dots, v_t\} \setminus \{i_*\}$

$$\lim_{j \rightarrow \infty} f(i, a(p'(j)))/f(i_*, a(p'(j))) = 0,$$

we have

$$B(i, n_i(a(p'(j))))/B(i_*, n_{i_*}(a(p'(j)))) \rightarrow 0$$

as  $j \rightarrow \infty$ .

Therefore, we can conclude that  $G$  is a stable distribution with characteristic exponent  $\lambda_{i_*} \in \{\lambda_1, \dots, \lambda_k\}$ .

We have finally exhausted all cases and established the theorem in each of these.  $\square$

For the proof of Theorem 2, we need the following lemma.

**LEMMA 2.** *If  $Y_1, \dots, Y_r$  are independent stable random variables with the characteristic exponent of  $Y_i$  being  $\alpha_i$  and  $0 < \alpha_1 < \dots < \alpha_r \leq 2$ , if  $a_1, \dots, a_r, b_1, \dots, b_r$  are constants such that  $a_1 Y_1 + \dots + a_r Y_r$  and  $b_1 Y_1 + \dots + b_r Y_r$  have the same distribution function, then  $a_i = b_i$  for all  $i$ .*

**PROOF.** Let  $Z_1 = a_1 Y_1 + \dots + a_r Y_r$  and  $Z_2 = b_1 Y_1 + \dots + b_r Y_r$ . Also, let  $Z_i', i = 1, 2$ , be random variables such that  $F_{Z_i} = F_{Z_i'}$  and  $(Z_1, Z_2), (Z_1', Z_2')$  are independent.

Let  $\psi$  be the logarithm of the characteristic function of  $Z_1 - Z_1'$ . Using the canonical representation of the logarithm of a stable distribution ([2], 164), we have  $\psi(t) = a_1 |t|^{\alpha_1} + \dots + a_r |t|^{\alpha_r}$  and  $\psi(t) = b_1 |t|^{\alpha_1} + \dots + b_r |t|^{\alpha_r}$ , for all  $t$ .

Since  $\alpha_1, \dots, \alpha_r$  are distinct,  $|t|^{\alpha_1}, \dots, |t|^{\alpha_r}$  are linearly independent, hence,  $a_i = b_i$  for all  $i$ .  $\square$

PROOF OF SUFFICIENCY PART OF THEOREM 2. Let  $X(i, m)$  be the  $m$ th random variable in the sequence  $\{X_n\}$  whose distribution function is  $F_i$ , for  $1 \leq i \leq r$ . Then for each  $i$ , there are constants  $\{B(i, n)\}$  and  $\{A(i, n)\}$ , with  $B(i, n) \rightarrow \infty$  as  $n \rightarrow \infty$  for all  $i$ , and there is a random variable  $Y_i$  such that the joint distribution function of

$$\left\{ \frac{X(i, 1) + \cdots + X(i, n)}{B(i, n)} - A(i, n), \quad 1 \leq i \leq r \right\}$$

converges in law to  $\{Y_1, \cdots, Y_r\}$ , where  $Y_i$  has a stable distribution with characteristic exponent  $\lambda_i$ , and where  $Y_1, \cdots, Y_r$  are independent.

We observe that

$$\begin{aligned} \frac{X_1 + \cdots + X_n}{B(r, n_r(n))} - \frac{1}{B(r, n_r(n))} \sum_{i=1}^r B(i, n_i(n)) A(i, n_i(n)) \\ = \frac{X(r, 1) + \cdots + X(r, n_r(n))}{B(r, n_r(n))} - A(r, n_r(n)) + \sum_{i=1}^{r-1} \frac{B(i, n_i(n))}{B(r, n_r(n))} \\ \cdot \left\{ \frac{X(i, 1) + \cdots + X(i, n_i(n))}{B(i, n_i(n))} - A(i, n_i(n)) \right\}. \end{aligned}$$

Since normalizing coefficients for a distribution function are unique up to a positive constant factor (see, e.g. Feller [1], Lemma 1, 246), Theorem 2 (iii) implies that, for any choice of normalizing coefficients for the  $F_i$ s,  $B(i, n_i(n))/B(r, n_r(n))$  converges to a positive, finite constant, for  $1 \leq i \leq r-1$ . Hence, by the argument used at the end of Case 1 in the proof of Theorem 1, we have that  $G$  is the distribution function of a linear combination of  $Y_1, \cdots, Y_r$ , with each  $Y_i$  needed.  $\square$

PROOF OF NECESSITY PART OF THEOREM 2. We now assume that  $G$  is a convolution of  $r$  stable distributions with characteristic exponents  $0 < \lambda_1 < \cdots < \lambda_r \leq 2$ .

By Zinger's Theorem, Theorem 2 (i) holds.

By Theorem 1,  $n_i(n)/n \rightarrow 0$  as  $n \rightarrow \infty$  for  $1 \leq i \leq r-1$ , hence,  $n_r(n)/n \rightarrow 1$  as  $n \rightarrow \infty$ , i.e. Theorem 2 (ii) holds.

Now we will show that Theorem 2 (iii) holds also. Let  $X(i, n)$ ,  $B(i, n)$ ,  $A(i, n)$  and  $Y_i$  be as in the proof of the sufficiency. Then by Lemma 1, for each  $i$ , there exists a measurable slowly varying function  $L_i$  such that  $B_n \sim n^{\lambda_i-1} L_i(n)$ .

Hence, for  $1 \leq i \leq r-1$ ,

$$\frac{B(i, n_i(n))}{B(r, n_r(n))} \sim \left( \frac{n_i(n)}{n^{\lambda_i/\lambda_r}} \right)^{\lambda_i-1} \frac{L_i(n_i(n))}{L_r(n)}.$$

For  $1 \leq i \leq r-1$ , we define

$$g(i, n) = \left( \frac{n_i(n)}{n^{\lambda_i/\lambda_r}} \right)^{\lambda_i-1} \frac{L_i(n_i(n))}{L_r(n)}.$$

In order to conclude the proof of the necessity, we need to show that  $\lim_{n \rightarrow \infty} g(i, n)$  exists and is positive and finite, for  $1 \leq i \leq r-1$ .

Let  $a_i = \liminf_{n \rightarrow \infty} g(i, n)$  and  $b_i = \limsup_{n \rightarrow \infty} g(i, n)$  for  $1 \leq i \leq r-1$ . We will first show that  $b_i < \infty$  for all  $i$ .

Suppose not. Then there is a  $k$  such that  $b_k = \infty$ . Select an increasing sequence  $\{m(j)\}$  of positive integers such that  $\lim_{j \rightarrow \infty} g(i, m(j)) = b_i'$ , for  $1 \leq i \leq r-1$ , with  $b_k' = \infty$ .

If  $b_k'$  is the only infinite one among  $b_1', \dots, b_{r-1}'$ , then, by the argument in Case 2 of the proof of Theorem 1, we can conclude that  $G$  is a stable distribution with characteristic exponent  $\lambda_k$ . This contradicts  $G$  being a convolution of  $r$  distinct stable distributions.

If several of the  $b_i'$ 's, say  $b_{i_1}', \dots, b_{i_u}'$ , are infinite, then, by an argument analogous to the one used in Case 3 of the proof of Theorem 1, we can conclude that  $G$  is a convolution of not more than  $u < r$  stable distributions. Again, we have a contradiction.

Therefore,  $b_i < \infty$  for all  $i$ .

Next, we will show that  $a_i > 0$ , for all  $i$ .

Suppose not. Let  $k$  be such that  $a_k = 0$ . Select an increasing sequence  $\{m(j)\}$  of positive integers such that  $\lim_{j \rightarrow \infty} g(i, m(j)) = a_i'$  for  $1 \leq i \leq r-1$ , with  $a_k' = 0$  and  $a_i' < \infty$  for all  $i$ . Then

$$B(i, n_i(m(j)))/B(r, n_r(m(j))) \sim g(i, m(j)) \rightarrow a_i' \quad \text{as } j \rightarrow \infty,$$

for  $1 \leq i \leq r-1$ . As in the proof of the sufficiency, this implies that  $G$  is the distribution function of  $c\{a_1'Y_1 + \dots + a_k'Y_k + \dots + a_{r-1}'Y_{r-1} + Y_r\}$  for some  $c > 0$ . Since  $a_k' = 0$ ,  $G$  is a convolution of not more than  $r-1$  stable distributions, which contradicts our hypothesis of  $r$  distinct stable components. Therefore,  $a_i > 0$  for all  $i$ .

Finally, we will show that  $a_i = b_i$  for all  $i$ .

Suppose not. Let  $k$  be such that  $a_k < b_k$ . Select increasing sequences  $\{m(j)\}$  and  $\{m'(j)\}$  of positive integers such that  $\lim_{j \rightarrow \infty} g(i, m(j)) = b_i'$  and  $\lim_{j \rightarrow \infty} g(i, m'(j)) = a_i'$  for  $1 \leq i \leq r-1$ , with  $b_k' = b_k$  and  $a_k' = a_k$ . Then, for  $1 \leq i \leq r-1$ ,

$$B(i, n_i(m(j)))/B(r, n_r(m(j))) \sim g(i, m(j)) \rightarrow b_i' \quad \text{as } j \rightarrow \infty,$$

and

$$B(i, n_i(m'(j)))/B(r, n_r(m'(j))) \sim g(i, m'(j)) \rightarrow a_i' \quad \text{as } j \rightarrow \infty.$$

Hence, there exist constants  $d_1 > 0$  and  $d_2 > 0$  such that  $G$  is the distribution function of both  $d_1\{b_1'Y_1 + \dots + b_{r-1}'Y_{r-1} + Y_r\}$  and  $d_2\{a_1'Y_1 + \dots + a_{r-1}'Y_{r-1} + Y_r\}$ . By Lemma 2,  $d_1b_i' = d_2a_i'$ , and  $1 \leq i \leq r-1$ , and  $d_1 = d_2$ . Hence,  $a_k' = b_k'$ , which implies  $a_k = b_k$ . We have the desired contradiction which shows that  $\lim_{n \rightarrow \infty} g(i, n)$  exists and is positive and finite for  $1 \leq i \leq r-1$ .  $\square$

To complete the proof of Theorem 2, we need to show that  $\{B_n\}$  can be chosen as normalizing coefficients for  $F_r$ . We have shown that there exists a constant  $d > 0$  such that  $B_n \sim dB(r, n_r(n))$ . Hence,

$$B_n \sim dn^{\lambda_r-1} \left( \frac{n_r(n)}{n} \right)^{\lambda_r-1} L_r(n) \frac{L_r(n(n_r(n)/n))}{L_r(n)} \\ \sim dn^{\lambda_r-1} L_r(n).$$

Therefore,  $B_n \sim dn^{\lambda_r-1} L_r(n)$ .  $\square$

PROOF OF THEOREM 3. Let  $\{X(i, n)\}$ ,  $\{B(i, n)\}$  and  $\{A(i, n)\}$  be as in the proof of Theorem 1. Then, by Lemma 1, for each  $i$  there exists a measurable slowly varying function  $L_i$  defined over  $(0, \infty)$  such that  $B(i, n) \sim n^{\lambda_i-1} L_i(n)$ .

Also, as in the proof of Theorem 1, select an increasing sequence  $\{m(j)\}$  of positive integers such that  $\lim_{j \rightarrow \infty} (n_1(m(j))/m(j)^{1-\eta}) = s_1$  and, for all  $i > 1$ ,  $\lim_{j \rightarrow \infty} (n_i(m(j))/m(j)) = s_i$  say.

Since  $\eta \in (0, 1 - \lambda_1/\lambda_2)$ , we see that for all  $i > 1$ ,  $B(i, n_i(m(j)))/B(1, n_1(m(j))) \rightarrow 0$  as  $j \rightarrow \infty$ . Hence, the distribution function of

$$\frac{X(1, 1) + \cdots + X(1, n_1(m(j)))}{B(1, n_1(m(j)))} - A(1, n_1(m(j))) + \sum_{i=2}^r \frac{B(i, n_i(m(j)))}{B(1, n_1(m(j)))} \\ \cdot \left\{ \frac{X(i, 1) + \cdots + X(i, n_i(m(j)))}{B(i, n_i(m(j)))} - A(i, n_i(m(j))) \right\}$$

converges to a stable distribution with characteristic exponent  $\lambda_1$ . Therefore,  $G$  is a stable distribution with characteristic exponent  $\lambda_1$ .  $\square$

PROOF OF THEOREM 4. Notation is as usual. Then, there exists a sequence  $\{m(j)\}$  of positive integers such that  $\lim_{j \rightarrow \infty} (n_1(m(j))/m(j)^{1-\eta}) = s_1$ ,  $\lim_{j \rightarrow \infty} (n_2(m(j))/m(j)) = s_2$  and, for all  $i > 2$ ,  $\lim_{j \rightarrow \infty} (n_i(m(j))/m(j)) = s_i$  say. Then

$$\frac{B(1, n_1(m(j)))}{B(2, n_2(m(j)))} \sim \frac{s_1^{\lambda_1-1} m(j)^{(1-\eta)/\lambda_1} L_1(m(j))^{(1-\eta)/\lambda_1}}{s_2^{\lambda_2-1} m(j)^{\lambda_2-1} L_2(m(j))} \rightarrow 0$$

as  $j \rightarrow \infty$ , since  $\eta > 1 - \lambda_1/\lambda_2$  implies that  $(1-\eta)/\lambda_1 - 1/\lambda_2 < 0$ .

Since  $\lambda_i^{-1} - \lambda_2^{-1} < 0$  for all  $i > 2$ , we also have that, for all  $i > 2$ ,  $B(i, n_i(m(j)))/B(2, n_2(m(j))) \rightarrow 0$  as  $j \rightarrow \infty$ .

Therefore we can conclude that  $G$  is a stable distribution with characteristic exponent  $\lambda_2$ .  $\square$

The proofs of the Corollaries to Theorems 3 and 4 are obvious.

PROOF OF THEOREM 5. Note that

$$B(2, n_2(n))/B(1, n_1(n)) \sim s_2^{\lambda_2-1} s_1^{-\lambda_1-1} L_2(n)/L_1(n^{\lambda_1/\lambda_2}).$$

(i)  $L_2(n)/L_1(n^{\lambda_1/\lambda_2}) \rightarrow 0$  as  $n \rightarrow \infty$  implies that  $B(2, n_2(n))/B(1, n_1(n)) \rightarrow 0$  as  $n \rightarrow \infty$ . Since  $\lambda_i^{-1} - \lambda_2^{-1} < 0$ , for all  $i > 2$ , we have that there exists a sequence  $\{m(j)\}$  of positive integers such that, for all  $i > 2$ ,  $B(i, n_i(m(j)))/B(1, n_1(m(j))) \rightarrow 0$  as  $j \rightarrow \infty$ . This concludes (i).

(ii)  $L_2(n)/L_1(n^{\lambda_1/\lambda_2}) \rightarrow \infty$  as  $n \rightarrow \infty$  implies that  $B(1, n_1(n))/B(2, n_2(n)) \rightarrow 0$  as  $n \rightarrow \infty$ . For all  $i > 2$ , there exists a sequence  $\{m(j)\}$  of positive integers such that  $B(i, n_i(m(j)))/B(2, n_2(m(j))) \rightarrow 0$  as  $j \rightarrow \infty$ . This concludes (ii).

(iii) Let  $a = \liminf_{n \rightarrow \infty} L_2(n)/L_1(n^{\lambda_1/\lambda_2})$  and  $b = \limsup_{n \rightarrow \infty} L_2(n)/L_1(n^{\lambda_1/\lambda_2})$ . We first wish to prove that  $b < \infty$ . Suppose to the contrary that  $b = \infty$ . We have  $a < \infty$ . Then select a sequence  $\{m_1(j)\}$  of positive integers such that

$$L_2(m_1(j))/L_1(m_1(j)^{\lambda_1/\lambda_2}) \rightarrow \infty$$

as  $j \rightarrow \infty$ . Hence,  $B(1, n_1(m_1(j)))/B(2, n_2(m_1(j))) \rightarrow 0$  as  $j \rightarrow \infty$ . For all  $i > 2$ ,  $B(i, n_i(m_1(j)))/B(2, n_2(m_1(j))) \rightarrow 0$  as  $j \rightarrow \infty$ . Therefore,  $G$  is a stable distribution with characteristic exponent  $\lambda_2$ . However, since  $\{L_2(n)/L_1(n^{\lambda_1/\lambda_2})\}$  does not converge to  $\infty$ , we can also select a sequence  $\{m_2(j)\}$  of positive integers such that

$$L_2(m_2(j))/L_1(m_2(j)^{\lambda_1/\lambda_2}) \rightarrow a < \infty$$

as  $j \rightarrow \infty$ . Hence,  $B(2, n_2(m_2(j)))/B(1, n_1(m_2(j))) \rightarrow s_2^{\lambda_2^{-1}} s_1^{-\lambda_1 a^{-1}}$  as  $j \rightarrow \infty$ . For all  $i > 2$ ,  $B(i, n_i(m_2(j)))/B(1, n_1(m_2(j))) \rightarrow 0$  as  $j \rightarrow \infty$ . If  $a = 0$ , then  $G$  is a stable distribution with characteristic exponent  $\lambda_1$ , and if  $a > 0$ , then  $G$  is a convolution of two stable distributions with characteristic exponents  $\lambda_1$  and  $\lambda_2$ . If  $a \neq b = \infty$ , there are two possible combinations: (i)  $b = \infty$  and  $a = 0$ , (ii)  $b = \infty$  and  $0 < a < \infty$ . Under (i), we have a contradiction since  $b = \infty$  implies that  $G$  is a stable distribution with characteristic exponent  $\lambda_2$ , while  $a = 0$  implies that  $G$  is a stable distribution with characteristic exponent  $\lambda_1$ . Also, we have a contradiction under (ii) since  $0 < a < \infty$  implies that  $G$  is a convolution of two stable distributions with characteristic exponents  $\lambda_1 \neq \lambda_2$ .

Therefore  $b < \infty$ . We next show that  $a > 0$ .

Suppose to the contrary that  $a = 0$ . We have  $0 < b < \infty$ . Select a sequence  $\{m_1(j)\}$  of positive integers such that  $L_2(m_1(j))/L_1(m_1(j)^{\lambda_1/\lambda_2}) \rightarrow 0$  as  $j \rightarrow \infty$ . Hence,  $B(2, n_2(m_1(j)))/B(1, n_1(m_1(j))) \rightarrow 0$  as  $j \rightarrow \infty$ . For all  $i > 2$ ,

$$B(i, n_i(m_1(j)))/B(1, n_1(m_1(j))) \rightarrow 0$$

as  $j \rightarrow \infty$ . Therefore,  $G$  is a stable distribution with characteristic exponent  $\lambda_1$ . However, we can also select a sequence  $\{m_2(j)\}$  of positive integers such that  $L_2(m_2(j))/L_1(m_2(j)^{\lambda_1/\lambda_2}) \rightarrow b$  as  $j \rightarrow \infty$ . Hence,  $B(2, n_2(m_2(j)))/B(1, n_1(m_2(j))) \rightarrow s_2^{\lambda_2^{-1}} s_1^{-\lambda_1 b^{-1}} b$ , as  $j \rightarrow \infty$ , which is positive and finite. For all  $i > 2$ ,

$$B(i, n_i(m_2(j)))/B(1, n_1(m_2(j))) \rightarrow 0$$

as  $j \rightarrow \infty$ . Therefore,  $G$  is a convolution of two stable distributions with characteristic exponents  $\lambda_1$  and  $\lambda_2$ . Again, we have a contradiction since  $\lambda_1 \neq \lambda_2$ .

Therefore,  $0 < a \leq b < \infty$ . Hence, we can select a sequence  $\{m(j)\}$  of positive integers such that  $L_2(m(j))/L_1(m(j)^{\lambda_1/\lambda_2}) \rightarrow a$ , as  $j \rightarrow \infty$ , which is positive and finite. As above, this implies that  $G$  is a convolution of two stable distributions with characteristic exponents  $\lambda_1$  and  $\lambda_2$ . This concludes (iii).  $\square$

**Acknowledgment.** I wish to thank Professor Howard G. Tucker whose guidance and encouragement contributed greatly to the completion of this work.

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