

THE ACCURACY OF INFINITELY DIVISIBLE APPROXIMATIONS  
TO SUMS OF INDEPENDENT VARIABLES WITH  
APPLICATION TO STABLE LAWS

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**1. Introduction and summary.** Let  $\{F_n\}$  be a sequence of distribution functions defined on the real line, and suppose  $\{F_n(x)\}$  converges to some limiting distribution function  $F(x)$ . It is of interest to investigate the error involved in using  $F(x)$  as an approximation to  $F_n(x)$ , that is to investigate the rate of convergence of  $\{F_n\}$  to  $F$ . This leads to the problem of finding bounds on  $M_n = \sup_{-\infty < x < \infty} |F_n(x) - F(x)|$ . In particular, this problem has been studied by several authors for cases where  $F_n(x)$  represents the distribution function of a certain sum of independent random variables.

For cases involving the classical forms of the central limit theorem Berry [1] and Esseen [3] have obtained certain bounds on  $M_n$  which have been reinvestigated and improved by many authors (c.f. [4] Chapter XVI).

Let  $(X_{nk}), k = 1, 2, \dots, k_n; n = 1, 2, \dots$  be a system of random variables such that for each  $n, X_{n1}, \dots, X_{nk}$  are independent (we say the system is independent within each row). In [6], under suitable conditions, bounds have been obtained on  $M_n$  for the case where  $F_n(x)$  is the distribution function of  $S_n = X_{n1} + \dots + X_{nk}$  and  $F(x)$  is an infinitely divisible distribution. A basic assumption made in [6] was that both  $X_{nk}$  and  $F(x)$  have finite variances.

The purpose of this study is to extend the results of [6] to include the case where neither  $F(x)$  nor  $X_{nk}$  need have finite variance. Our main theorem (Theorem 1) gives a bound on  $M_n$  under a mild assumption on  $X_{nk}$  and a certain assumption on the derivative of the infinity divisible distribution  $F(x)$ . It is shown in Section 4, that if  $F(x)$  satisfies an additional condition which is considerably weaker than that having finite variance, then the bound obtained tends to zero as  $n$  becomes infinite under necessary and sufficient conditions that  $\{F_n(x)\}$  converge to  $F(x)$ .

In Section 5 our general results are applied to the case of convergence of distribution functions of normed sums of independent identically distributed random variables to an arbitrary stable law with exponent  $\alpha, 0 < \alpha < 2$ .

**2. Notation and preliminaries.** Let  $F(x)$  be an infinitely divisible distribution with characteristic function  $\varphi(t)$ . According to the Lévy-Khintchine formula we have

$$(2.1) \quad \log \varphi(t) = i\gamma t + \int_{-\infty}^{\infty} \left( e^{itu} - 1 - \frac{i+u}{1+u^2} \right) \frac{1+u^2}{u^2} dG(u)$$

Received April 18, 1968.

where  $\gamma$  is a constant and  $G(u)$  is a bounded nondecreasing function. The logarithm of  $\varphi(t)$  can also be represented by Lévy's formula :

$$(2.2) \quad \log \varphi(t) = i\gamma t - \frac{1}{2}(t^2\sigma^2) + \int_{-\infty}^{0-} \left( e^{itx} - 1 - \frac{itx}{i+x^2} \right) dM(x) \\ + \int_{0+}^{\infty} \left( e^{itx} - 1 - \frac{itx}{i+x^2} \right) dN(x),$$

where the connection between  $\sigma^2$ ,  $M$ ,  $N$  and  $G$  is given by

$$(2.3) \quad M(x) = \int_{-\infty}^x \frac{1+u^2}{u^2} dG(u) \quad \text{for } x < 0, \\ N(x) = - \int_x^{\infty} \frac{1+u^2}{u^2} dG(u) \quad \text{for } x > 0, \\ \sigma^2 = G(0+) - G(0-).$$

If an infinitely divisible distribution function  $F$  has finite variance, Kolmogorov's formula yields

$$(2.4) \quad \log \varphi(t) = i\mu t + \int_{-\infty}^{\infty} (e^{itv} - 1 - itv)v^{-2} dK(v),$$

where  $\mu$  is the mean of  $F$ , and  $K$  is a bounded nondecreasing function with  $K(-\infty) = 0$  and  $K(+\infty)$  equal to the variance of  $F$ . The relationship between (2.4) and (2.1) is given by

$$(2.5) \quad \mu = \gamma + \int_{-\infty}^{\infty} u dG(u) \quad \text{and} \\ K(v) = \int_{-\infty}^v (1+u^2) dG(u).$$

A system of random variables  $(X_{nk})$  as considered in Section 1 is said to be infinitesimal if for any  $\varepsilon > 0 \lim_{n \rightarrow \infty} \max_{1 \leq k \leq k_n} P(|X_{nk}| > \varepsilon) = 0$ . Given such a system, for any  $a > 0$  let  $X_{nk}^a$  be defined by

$$X_{nk}^a = X_{nk} \quad \text{if } -a < x_{nk} \leq a, \\ = 0 \quad \text{otherwise;}$$

and let  $F_{nk}^a(x)$ ,  $\varphi_{nk}^a(t)$ ,  $\mu_{nk}(a)$ ,  $\sigma_{nk}^2(a)$  denote respectively, the distribution function, characteristic function, mean and variance of  $X_{nk}^a$ . Let  $S_n = X_{n1} + \dots + X_{nk_n}$  and  $S_n^a = X_{n1}^a + \dots + X_{nk_n}^a$ . Let  $F_n(x)$  and  $\varphi_n(t)$  denote the distribution function and characteristic function of  $S_n$  and let  $F_n^a(x)$ ,  $\varphi_n^a(t)$ ,  $\mu_n(a)$  and  $\sigma_n^2(a)$  denote respectively the distribution function, characteristic function, mean, and variance of  $S_n^a$ .

Let  $F(x)$  be an infinitely divisible distribution with corresponding  $G(u)$  and  $\gamma$  given by (2.1). For each  $a > 0$  such that  $\pm a$  are continuity points of  $G(u)$  we define

$$(2.6) \quad G^a(u) = 0, \quad u \leq -a \\ = G(u) - G(-a), \quad -a < u \leq a \\ = G(a) - G(-a), \quad u > a$$

and

$$\gamma^a = \gamma - \int_{|u|>a} u^{-1} dG(u).$$

The nondecreasing function  $G^a$  and the constant  $\gamma^a$  determine a unique infinitely divisible distribution  $F^a(x)$  through the formula (2.1). In [8] it is shown that if  $F_n(x) \rightarrow F(x)$  then for any  $a > 0$  (with  $\pm a$  continuity points of  $G$ )  $F_n^a(x) \rightarrow F^a(x)$  and that  $\lim_{n \rightarrow \infty} \int_{-\infty}^{\infty} x^k dF_n^a(x) = \int_{-\infty}^{\infty} x^k dF^a(x)$  for any positive integer  $k$  (the result of [7] shows that limit is finite). We let  $\varphi^a(t)$ ,  $\mu(a)$ , and  $\sigma^2(a)$  denote the characteristic function mean, and variance of  $F^a(x)$ . From (2.4) we have

$$(2.7) \quad \log \varphi^a(t) = i\mu(a)t + \int_{-\infty}^{\infty} (e^{itv} - 1 - itv)v^{-2} dK^a(v)$$

where

$$\mu(a) = \gamma^a + \int_{-\infty}^{\infty} u dG^a(u) \quad \text{and}$$

$$K^a(v) = \int_{-\infty}^v (1 + u^2) dG^a(v).$$

Let

$$K_n^a(v) = \sum_{k=1}^{k_n} \int_{-\infty}^v x^2 dF_{nk}^a(x + \mu_{nk}(a)).$$

For any  $A > 0$  such that  $-A$  and  $A$  are continuity points of  $G(u)$ , and hence continuity points of  $K^a(v)$ , let  $0 < \delta \leq 2A$  and define

$$m = m(A, \delta) = [2A/\delta] + 1.$$

Let  $-A = x_0 < x_1 < \dots < x_m = A$  be such that  $x_i$  is a continuity point of  $K^a(v)$  and such that  $\max_{1 \leq i \leq m} (x_i - x_{i-1}) < \delta$ .

Define

$$(2.8) \quad \begin{aligned} E^a(n, t, m(A, \delta)) = & \frac{5}{2} \delta |t|^3 (\sigma_n^2(a) + \sigma^2(a)) + \frac{1}{2} t^2 \sum_{i=0}^m |K_n^a(x_i) - K^a(x_i)| \\ & + 2A^{-1} |t| [(K_n^a(+\infty) - K_n^a(A) + K^a(+\infty) - K^a(A)) \\ & + K_n^a(-A) + K^a(-A)]. \end{aligned}$$

$E^a$  is used in obtaining the desired bounds which will involve  $g^a(n, m(A, \delta), r)$  defined by

$$(2.9) \quad \begin{aligned} g^a(n, m(A, \delta), r) = & [\frac{1}{3} \sigma_n^2(a) \max_{1 \leq k \leq k_n} \sigma_{nk}^2(a)]^{1/5} + [\frac{5}{6} \delta (\sigma_n^2(a) + \sigma^2(a))]^{1/4} \\ & + [\frac{1}{2} \sum_{i=0}^m |K_n^a(x_i) - K^a(x_i)|]^{1/3} \\ & + \{4A^{-1} [K_n^a(+\infty) - K_n^a(A) + K^a(+\infty) - K^a(A) \\ & + K_n^a(-A) + K^a(-A)] + 2|\mu_n(a) - \mu(a)|\}^{1/2} \\ & + [8r^{-1} \int_{|u|>a} |u|^r dG(u)]^{1/(1+r)} \end{aligned}$$

where  $r$  is a positive real number.

**3. The general bound.**

We require two lemmas.

LEMMA 1. *With  $F_n(x)$  and  $F_n^a(x)$  defined as in Section 2, it follows that*

$$|F_n(x) - F_n^a(x)| \leq \sum_{k=1}^{k_n} \{F_{nk}(-a) + 1 - F_{nk}(a)\}.$$

PROOF. We have that

$$\begin{aligned} F_n(x) &= P(S_n \leq x) = P(S_n \leq x, X_{nk} \in (-a, a] \text{ for all } k) \\ &\quad + P(S_n \leq x \text{ and } X_{nk} \notin (-a, a] \text{ for some } k) \\ &\leq P(S_n^a \leq x) + P(X_{nk} \notin (-a, a] \text{ for some } k) \\ &\leq F_n^a(x) + \sum_{k=1}^{k_n} P(X_{nk} \notin (-a, a]). \end{aligned}$$

Thus  $F_n(x) - F_n^a(x) \leq \sum_{k=1}^{k_n} [F_{nk}(-a) + 1 - F_{nk}(a)]$ . A similar argument, starting with  $F_n^a(x)$  proves the lemma.

LEMMA 2. *Let  $F(x)$  be an infinitely divisible distribution with characteristic function  $\varphi(t)$  and with corresponding  $G(u)$  given by (2.1). Then for any real numbers  $a$  and  $r$  with  $a > 1$ , ( $\pm a$  continuity points of  $G$ ) and  $0 < r \leq 1$  we have  $|\varphi^a(t) - \varphi(t)| \leq 4|t|^r \int_{|u|>a} |u|^r dG(u)$  where  $\varphi^a(t)$  is given by (2.7).*

PROOF. Since  $\log \varphi^a(t)$  is given by formula (2.1) using  $G^a(u)$  and  $\gamma^a$  given by (2.6), we have using Lemma 1 of [6] that

$$\begin{aligned} |\varphi^a(t) - \varphi(t)| &\leq |\log \varphi^a(t) - \log \varphi(t)| \\ &= \left| -it \int_{|u|>a} u^{-1} dG(u) - \int_{|u|>a} \left( e^{itu} - 1 - \frac{itu}{1+u^2} \right) \frac{1+u^2}{u^2} dG(u) \right| \\ &= \left| \int_{|u|>a} (e^{itu} - 1) u^{-2} (1+u^2) dG(u) \right|. \end{aligned}$$

Now for  $|u| > a > 1$  we have

$$\left| (e^{itu} - 1) \frac{1+u^2}{u^2} \right| \leq 2|e^{itu} - 1| = 4|\sin \frac{1}{2}tu|.$$

Furthermore, for  $|\frac{1}{2}tu| \geq 1$  we have  $4|\sin \frac{1}{2}tu| \leq 4 \leq 4|\frac{1}{2}tu|^r \leq 4|tu|^r$ , and for  $|\frac{1}{2}tu| \leq 1$  and  $0 < r \leq 1$  we have  $4|\sin \frac{1}{2}tu| \leq 4|\frac{1}{2}tu| \leq 4|\frac{1}{2}tu|^r \leq 4|tu|^r$ .

It follows that  $|\varphi^a(t) - \varphi(t)| \leq 4|t|^r \int_{|u|>a} |u|^r dG(u)$ , which proves the lemma.

We are now in a position to state and prove the theorem giving a general bound mentioned in Section 1. We use the notation developed in Section 2.

THEOREM 1. *Let  $F(x)$  be an infinitely divisible distribution function with corresponding  $G(u)$  given by the Levy-Khintchine formula (2.1). Let  $(X_{nk})_{k=1, \dots, k_n}$ ;  $n = 1, 2, \dots$  be a system of random variables independent within each row. Let  $F_n(x)$  be the distribution function of the sum  $S_n = X_{n1} + \dots + X_{nk_1}$  and suppose that*

$$dF(x)/dx = F'(x) \text{ exists and } |F'(x)| < B \text{ for all } x.$$

Assume that  $\sigma_{nk}^2(a) \leq 1$  for all  $n$  and  $k$  (as will be seen in the proof of Theorem 2, this assumption is quite weak). Then it follows that for  $0 < r \leq 1$  and  $a \geq 1$ ,

$$M_n = \sup_{-\infty < x < \infty} |F_n(x) - F(x)| \leq k(B)g^a(n, m(A, \delta), r) + \sum_{k=1}^{k_n} \{F_{nk}(-a) + 1 - F_{nk}(a)\}$$

where  $k(B)$  is a constant depending only on  $B$  and where  $g^a(n, m(A, \delta), r)$  is given by (2.9).

PROOF. We have

$$(3.1) \quad |F_n(x) - F(x)| \leq |F_n(x) - F_n^a(x)| + |F_n^a(x) - F(x)|.$$

Letting  $\varphi(t)$  be the characteristic function of  $F(x)$ , from Lemma 2 it follows that

$$\begin{aligned} |\varphi_n^a(t) - \varphi(t)| &\leq |\varphi_n^a(t) - \varphi^a(t)| + |\varphi^a(t) - \varphi(t)| \\ &\leq |\varphi_n^a(t) - \varphi^a(t)| + 4|t|^r \int_{|u|>a} |u|^r dG(u). \end{aligned}$$

If we let  $T_n = [g^a(n, m(A, \delta), r)]^{-1}$  and restrict  $|t| \leq T_n$ , by an argument analogous to the proof of Theorem 3 of [6], it follows that

$$\begin{aligned} |\varphi_n^a(t) - \varphi^a(t)| &\leq \frac{5}{8}t^4 \max_{1 \leq k \leq k_n} \sigma_{nk}^2(a) \sigma_n^2(a) \\ &\quad + |\mu_n(a) - \mu(a)| |t| + E^a(n, t, m(A, \delta)) \end{aligned}$$

where  $E^a$  is given by (2.8). Thus

$$\begin{aligned} |\varphi_n^a(t) - \varphi(t)| &\leq \frac{5}{8}t^4 \max_{1 \leq k \leq k_n} \sigma_{nk}^2(a) \sigma^2(a) \\ &\quad + |\mu_n(a) - \mu(a)| |t| + E^a(n, t, m(A, \delta)) + 4|t|^r \int_{|u|>a} |u|^r dG(u) \\ &\equiv h^a(t, n, m(A, \delta), r). \end{aligned}$$

From this it follows that

$$\int_{-T_n}^{T_n} |t|^{-1} (\varphi_n^a(t) - \varphi(t)) dt \leq 2 \int_0^{T_n} t^{-1} [h^a(t, n, m(A, \delta), r)] dt \leq g^a(n, m(A, \delta), r).$$

Now applying a Theorem of Esseen ([3] Theorem 2a, page 32), we have

$$M_n^a \equiv \sup_{-\infty < x < \infty} |F_n^a(x) - F(x)| \leq (2\pi)^{-1} p g^a(n, m(A, \delta), r) + c(p) \cdot B/T_n$$

where  $p > 1$  and  $c(p)$  is a constant depending only on  $p$  ( $p > 1$  is arbitrary). Thus

$$M_n^a \leq (p/(2\pi) + c(p)B)g^a(n, m(A, \delta), r).$$

Applying Lemma 1 and (3.1) we have the proof of the Theorem.

**4. Behavior of the estimate.** In this section we examine, under suitable conditions, the behavior of the bound

$$(4.1) \quad D(n, A, \delta, a, r) = k(B)g^a(n, m(A, \delta), r) + \sum_{k=1}^{k_n} \{F_{nk}(-a) + 1 - F_{nk}(a)\}$$

given in Theorem 1. Several lemmas will be needed.

LEMMA 3. Let  $\{Q_n(a)\}$   $n = 0, 1, 2, \dots$  be defined for  $a > 0$  and be such that

- (i) for each  $n$ ,  $Q_n(a) \geq Q_n(b) \geq 0$  for  $a < b$ .
- (ii)  $\lim_{a \rightarrow +\infty} Q_0(a) = 0$ .
- (iii)  $\lim_{n \rightarrow \infty} Q_n(a) = Q_0(a)$  at every continuity point of  $Q_0(a)$ .

Then for any nondecreasing sequence  $\{a_n\}$  such that each  $a_n$  is a continuity point of  $Q_0(a)$ , and such that  $\lim_{n \rightarrow \infty} a_n = +\infty$  we have  $\lim_{n \rightarrow \infty} Q_n(a_n) = 0$ .

PROOF. Given  $\varepsilon > 0$ , let  $k$  be such that  $Q_0(a_k) < \varepsilon/2$ . Let  $n_k$  be such that  $n > n_k$  implies  $Q_n(a_k) < Q_0(a_k) + \varepsilon/2 < \varepsilon$ . Now for  $n > n_k$  we have  $n > k$  so that by (i),  $Q_n(a_n) \leq Q_n(a_k) < \varepsilon$ , which proves the lemma.

LEMMA 4. Let  $(X_{nk})$ ,  $F_n(x)$ ,  $F(x)$  and  $G(u)$  be as in Theorem 1. If

- (i) the system  $(X_{nk})$  is infinitesimal,
  - (ii)  $\lim_{n \rightarrow \infty} F_n(x) = F(x)$  at all continuity points of  $F(x)$ ,
- then for any nondecreasing sequence  $\{a_n\}$  such that  $-a_n$  and  $a_n$  are continuity points of  $G$  and  $\lim_{n \rightarrow \infty} a_n = \infty$  we have

$$\lim_{n \rightarrow \infty} \sum_{k=1}^{k_n} \{F_{nk}(-a_n) + 1 - F_{nk}(a_n)\} = 0.$$

PROOF. Let  $M(x)$  and  $N(x)$  be given by (2.2) or (2.3) and let

$$Q_n(a) = \sum_{k=1}^{k_n} \{F_{nk}(-a) + 1 - F_{nk}(a)\} \quad \text{and} \quad Q_0(a) = M(-a) - N(a).$$

By Theorem 1, page 116 of [5] we have  $\lim_{n \rightarrow \infty} Q_n(a) = Q_0(a)$  at continuity points of  $Q_0(a)$  so that by Lemma 3,  $\lim_{n \rightarrow \infty} Q_n(a_n) = 0$ .

LEMMA 5. Let  $g(n, a, \delta)$  be nonnegative and be such that to each  $a$  there corresponds a sequence  $\{\delta_n(a)\}$  of positive real numbers such that  $\lim_{n \rightarrow \infty} g(n, a, \delta_n(a)) = 0$ . Then there exists a nondecreasing sequence  $\{a_n\}$  such that  $\lim_{n \rightarrow \infty} a_n = \infty$  and a sequence  $\{\delta_n\}$  such that

$$\lim_{n \rightarrow \infty} g(n, a_n, \delta_n) = 0.$$

PROOF. Let  $\{\varepsilon_k\}$  be a sequence such that  $\varepsilon_k \downarrow 0$  and let  $\{\bar{a}_n\}$  be such that  $\bar{a}_n \uparrow \infty$ . By hypothesis we can choose  $\{n_k\}$  such that  $n_k > n_{k-1}$  and such that  $g(n, \bar{a}_k, \delta_n(\bar{a}_k)) < \varepsilon_k$  for  $n > n_k$ .

If we define

$$\begin{aligned} \delta_n &= \delta_1(\bar{a}_1) \quad \text{for } n \leq n_1, \\ &= \delta_n(\bar{a}_k) \quad \text{for } n_k < n \leq n_{k+1}, \end{aligned} \quad k = 1, 2, \dots$$

and

$$\begin{aligned} a_n &= \bar{a}_1 \quad \text{for } n \leq n_1 \\ &= \bar{a}_k \quad \text{for } n_k < n \leq n_{k+1}, \end{aligned} \quad k = 1, 2, \dots$$

it is not difficult to see that  $\{a_n\}$  and  $\{\delta_n\}$  satisfy the conclusion of the lemma.

We now show that under suitable conditions we can choose sequences  $\{A_n\}$ ,  $\{\delta_n\}$  and  $\{a_n\}$  such that the bound,  $D(n, A_n, \delta_n, a_n, r)$  given by (4.1) approaches zero as  $n$  becomes infinite.

Using the same notation as in Theorem 1, let  $0 < \delta < 1$  be such that  $\pm\delta^{-\frac{1}{2}}$  are continuity points of  $G(u)$ . Using (2.9) define  $g(n, a, \delta)$  by

$$(4.2) \quad g^a(n, m(\delta^{-\frac{1}{2}}, \delta), r) = g(n, a, \delta) + \{8r^{-1} \int_{|u|>a} |u|^r dG(u)\}^{1/(1+r)}.$$

This leads to the main result of this section.

**THEOREM 2.** *Let  $(X_{nk}), F_n(x)$  and  $F(x)$  satisfy the conditions of Theorem 1. Assume further that the random variables  $(X_{nk})$  are infinitesimal,  $F_n(x)$  converges to  $F(x)$  at continuity points of  $F$ , and that for some  $r$ , (without loss of generality assume  $0 < r \leq 1$ )*

$$\int_{-\infty}^{\infty} |u|^r dG(u) < \infty.$$

*Then there exist sequences  $\{a_n\}$  and  $\{\delta_n\}$  such that, for large  $n$*

$$(4.3) \quad M_n = \sup_{-\infty < x < \infty} |F_n(x) - F(x)| \leq D(n, \delta_n^{-\frac{1}{2}}, \delta_n, a_n, r), \quad \text{and}$$

$$(4.4) \quad \lim_{n \rightarrow \infty} D(n, \delta_n^{-\frac{1}{2}}, \delta_n, a_n, r) = 0.$$

(The function  $D$  is given by (4.1) and (2.9)).

**PROOF.** Since  $(X_{nk})$  are infinitesimal,  $(X_{nk}^a)$  as well as  $(X_{nk}^a - \mu_{nk}(a))$  are also infinitesimal. By Theorems 3 and 6 of [8], if  $\pm a$  are continuity points of  $G$ , we have  $\lim_{n \rightarrow \infty} F_n^a(x) = F^a(x)$  and  $\lim_{n \rightarrow \infty} \sigma_n^2(a) = \sigma^2(a)$ . It follows from the proof of Theorem 5 of [6] that there exists a sequence  $\{\delta_n(a)\}$  such that

$$(4.5) \quad \lim_{n \rightarrow \infty} g(n, a, \delta_n(a)) = 0.$$

By Lemma 5 we can find sequences  $\{a_n\}$  and  $\{\delta_n\}$  such that

$$(4.6) \quad \lim_{n \rightarrow \infty} g(n, a_n, \delta_n) = 0,$$

$a_n \leq a_{n+1}$ , and  $\lim_{n \rightarrow \infty} a_n = \infty$ . (Note that from (4.2) and (2.9) we see that (4.5) and (4.6) imply  $\lim_{n \rightarrow \infty} \max_k \sigma_{nk}^2(a) = \lim_{n \rightarrow \infty} \max_k \sigma_{nk}^2(a_n) = 0$ . This justifies the parenthetical remark in the statement of Theorem 1.) Clearly from the proof of Lemma 5,  $a_n$  can be chosen so that  $\pm a_n$  are continuity points of  $G(u)$  so that the conclusion of Lemma 4 holds. Now since  $\int_{-\infty}^{\infty} |u|^r dG(u) < \infty$  it follows that  $\lim_{n \rightarrow \infty} \int_{|u|>a_n} |u|^r dG(u) = 0$ . Since

$$D(n, \delta_n^{-\frac{1}{2}}, \delta_n, a_n, r) = k(B)\{g(n, a_n, \delta_n) + [8r^{-1} \int_{|u|>a_n} |u|^r dG(u)]^{1/(1+r)}\} \\ + \sum_{k=1}^{k_n} \{F_{nk}(-a) + 1 - F_{nk}(a)\}$$

we see that (4.4) holds. Finally from Theorem 1, as soon as  $n$  is so large that  $a_n \geq 1$  and  $\max_k \sigma_{nk}^2(a_n) \leq 1$ , it follows that (4.3) holds.

**5. Application to stable laws.** The basic properties of stable distributions can be found in [4] or [5]. We recall that a distribution function  $F(x)$  is stable if and only if there exists a sequence of independent identically distributed random variables  $\{X_n\}$  and constants  $A_n$  and  $B_n > 0$  such that

$$(5.1) \quad \lim_{n \rightarrow \infty} P \left\{ \frac{X_1 + \cdots + X_n}{B_n} - A_n \leq x \right\} = F(x).$$

Since stable distributions are infinitely divisible, by letting  $X_{nk} = X_k/B_n$  it can be seen that our previous results can be applied to limit theorems of the type (5.1).

As is well known to every stable distribution, there corresponds an exponent  $\alpha$  ( $0 < \alpha \leq 2$ ). The case  $\alpha = 2$  corresponds to the normal distribution and will not be discussed here. From the theorem on page 164 of [5] we know that for stable distributions with exponent  $\alpha$ , ( $0 < \alpha < 2$ ), the functions  $M$  and  $N$  and the constant  $\sigma^2$  in (2.3) are given

$$(5.2) \quad \begin{aligned} M(x) &= c_1 |x|^{-\alpha}, & x < 0 \\ N(x) &= -c_2 x^{-\alpha}, & x > 0 \\ \sigma^2 &= 0 = G(0+) - G(0-) \end{aligned}$$

where  $c_1$  and  $c_2$  are nonnegative and  $c_1 + c_2 > 0$ . From Section 36 of [5] it follows that all proper stable distributions (and hence all stable distributions with  $0 < \alpha < 2$ ) have bounded derivatives of all orders of every point. Thus for (proper) stable distributions the assumption in Theorem 1 on the derivative of  $F(x)$  is always satisfied.

The next lemma removes one of the hypotheses of Theorem 2.

**LEMMA 6.** *If  $F(x)$  is a stable distribution function with corresponding function  $G(u)$  given by the formula (2.1), then there exists a real number  $r > 0$  such that*

$$\int_{-\infty}^{+\infty} |u|^r dG(u) < +\infty.$$

**PROOF.** From (5.2) and (2.3) we note that

$$(5.3) \quad \begin{aligned} dG(u) &= c_1 \frac{|u|^{1-\alpha}}{1+u^2} du, & \text{for } u < 0 \\ &= c_2 \frac{u^{1-\alpha}}{1+u^2} du, & \text{for } u > 0. \end{aligned}$$

$$\text{Thus} \quad \int_{-\infty}^{+\infty} |u|^r dG(u) = (c_1 + c_2) \int_{0+}^{+\infty} \frac{x^{1-(\alpha-r)}}{1+x^2} dx$$

which is finite if  $r < \alpha$ .

\* The next two lemmas will be used to simplify the general estimate given in Theorem 1 to the stable case.



LEMMA 7. Let  $F(x)$  be a stable distribution function with representation given by (2.2) and (5.2). Then using the notation of Section 2,

$$(5.4) \quad \mu(a) = \gamma + (c_1 - c_2) \left\{ \int_a^{+\infty} \frac{du}{(1+u^2)u^\alpha} - \int_0^a \frac{u^{2-\alpha} du}{1+u^2} \right\},$$

$$(5.5) \quad \sigma^2(a) = (c_1 + c_2) \frac{a^{2-\alpha}}{2-\alpha},$$

and

$$(5.6) \quad \begin{aligned} K^a(v) &= 0, & v < -a, \\ &= c_1 \left( \frac{a^{2-\alpha}}{2-\alpha} - \frac{|v|^{2-\alpha}}{2-\alpha} \right), & -a \leq v < 0, \\ &= c_1 \left( \frac{a^{2-\alpha}}{2-\alpha} \right) + c_2 \left( \frac{v^{2-\alpha}}{2-\alpha} \right), & 0 \leq v < a \\ &= (c_1 + c_2) \left( \frac{a^{2-\alpha}}{2-\alpha} \right), & a \leq v. \end{aligned}$$

PROOF. From (5.3) and (2.6) we have

$$\gamma^a = \gamma + (c_1 - c_2) \int_a^{+\infty} \frac{du}{(1+u^2)u^\alpha}.$$

Hence, from (2.7)

$$\begin{aligned} \mu(a) &= \gamma^a + \int_{-a}^a u dG(u) \\ &= \gamma^a + \int_{-a}^0 uc_1 \frac{|u|^{1-\alpha}}{1+u^2} du + \int_0^a uc_2 \frac{u^{1-\alpha}}{1+u^2} du \\ &= \gamma^a - (c_1 - c_2) \int_0^a \frac{u^{2-\alpha}}{1+u^2} du \\ &= \gamma + (c_1 - c_2) \left\{ \int_a^{+\infty} \frac{du}{(1+u^2)u^\alpha} - \int_0^a \frac{u^{2-\alpha}}{1+u^2} du \right\}. \end{aligned}$$

This proves (5.4). For  $-a < v < 0$ , we have

$$\begin{aligned} K^a(v) &= \int_{-\infty}^v (1+u^2) dG^a(u) = \int_{-a}^v (1+u^2) dG(u) \\ &= \int_{-a}^v c_1 |u|^{1-\alpha} du = c_1 \left( \frac{a^{2-\alpha}}{2-\alpha} - \frac{v^{2-\alpha}}{2-\alpha} \right). \end{aligned}$$

For  $0 < v < a$ , we have

$$\begin{aligned} K^a(v) &= \int_{-\infty}^v (1+u^2) dG^a(u) \\ &= \int_{-a}^0 (1+u^2) dG(u) + [G(0+) - G(0-)] + \int_0^v (1+u^2) dG(u) \\ &= c_1 \left( \frac{a^{2-\alpha}}{2-\alpha} \right) + c_2 \left( \frac{v^{2-\alpha}}{2-\alpha} \right), \end{aligned}$$

using (5.3) and the fact that  $G(0+) - G(0-) = \sigma^2 = 0$  from (5.2). From this (5.6) follows. Formula (5.5) follows from (5.6) and the fact that  $\sigma^2(a) = K^a(+\infty)$ .

LEMMA 8. *Let  $F(x)$  be a stable distribution function as given in Lemma 7. If  $0 < r < \alpha$ , then*

$$\int_{|u|>a} |u|^r dG(u) \leq \frac{c_1 + c_2}{(\alpha - r)a^{\alpha-r}}.$$

PROOF. Using (5.3) we have

$$\begin{aligned} \int_{|u|>a} |u|^r dG(u) &= (c_1 + c_2) \int_a^{+\infty} \frac{u^{1+r-\alpha}}{1+u^2} du \\ &\leq (c_1 + c_2) \int_a^{+\infty} \frac{u^{1+r-\alpha}}{u^2} du \\ &= \frac{c_1 + c_2}{(\alpha - r)a^{\alpha-r}}, \end{aligned}$$

which proves the lemma.

As was done at the beginning of this section, let  $\{X_n\}$ ,  $n = 1, 2, \dots$  be a sequence of independent random variables with a common distribution function  $\bar{F}(x)$ . For each  $n$ , let

$$S_n = \frac{X_1 + \dots + X_n}{B_n},$$

where  $B_n$  are suitably chosen positive constants. Let  $F_n(x)$  denote the distribution function of  $S_n$ . To apply our general result, we let

$$X_{nk} = X_k/B_n.$$

The results expressed in (5.7)–(5.13) follow easily.

$$(5.7) \quad \begin{aligned} X_{nk}^a &= X_k/B_n & -aB_n < X_k \leq aB_n, \\ &= 0 & \text{otherwise.} \end{aligned}$$

$$(5.8) \quad F_{nk}(x) = \bar{F}(xB_n).$$

$$(5.9) \quad \begin{aligned} F_{nk}^a(x) &= 0 & \text{for } x \leq -a, \\ &= \bar{F}(xB_n) - \bar{F}(-aB_n) & \text{for } -a < x < 0, \\ &= \bar{F}(xB_n) + 1 - \bar{F}(aB_n) & \text{for } 0 \leq x \leq a, \\ &= 1 & \text{for } a \leq x. \end{aligned}$$

$$(5.10) \quad \mu_{nk}(a) = B_n^{-1} \int_{-aB_n}^{aB_n} x d\bar{F}(x) \quad \text{and} \quad \mu_n(a) = Bn_n^{-1} \int_{-aB_n}^{aB_n} x d\bar{F}(x).$$

$$(5.11) \quad \begin{aligned} \sigma_{nk}^2(a) &= B_n^{-2} \left\{ \int_{-aB_n}^{aB_n} x^2 d\bar{F}(x) - \left( \int_{-aB_n}^{aB_n} x d\bar{F}(x) \right)^2 \right\} & \text{and} \\ \sigma_n^2(a) &= Bn_n^{-2} \left\{ \int_{-aB_n}^{aB_n} x^2 d\bar{F}(x) - \left( \int_{-aB_n}^{aB_n} x d\bar{F}(x) \right)^2 \right\}. \end{aligned}$$

$$\begin{aligned}
 K_n^a(v) &= n \int_{-\infty}^{v+\mu_{nk}(a)} (u - \mu_{nk}(a))^2 dF_{nk}^a(u) \\
 &= 0, \quad v + \mu_{nk}(a) \leq -a, \\
 &= n \int_{-a}^{v+\mu_{nk}(a)} (u - \mu_{nk}(a))^2 d\bar{F}(uB_n), \quad -a < v + \mu_{nk}(a) < 0, \\
 (5.12) \quad &= n \int_{-a}^{v+\mu_{nk}(a)} (u - \mu_{nk}(a))^2 d\bar{F}(uB_n) \\
 &\quad + [\mu_{nk}(a)]^2 [1 - \bar{F}(aB_n) + \bar{F}(-aB_n)], \quad 0 \leq v + \mu_{nk}(a) < a, \\
 &= n \int_{-a}^a (u - \mu_{nk}(a))^2 d\bar{F}(uB_n) \\
 &\quad + [\mu_{nk}(a)]^2 [1 - \bar{F}(aB_n) + \bar{F}(-aB_n)], \quad a < v + \mu_{nk}(a).
 \end{aligned}$$

Finally we have

$$(5.13) \quad \sum_{k=1}^{k_n} \{F_{nk}(-a) + 1 - F_{nk}(a)\} = n\{\bar{F}(-aB_n) + 1 - \bar{F}(aB_n)\}.$$

The following theorem is an immediate consequence of Theorem 1, Lemmas 7 and 8, and (5.7)–(5.13).

**THEOREM 3.** *Let  $\{X_n\}$ ,  $n = 1, 2, \dots$  be a sequence of independent random variables with a common distribution function  $\bar{F}(x)$ . For each  $n$ , let*

$$S_n = \frac{X_1 + \dots + X_n}{B_n},$$

where  $B_n$  are positive constants. Let  $F_n(x)$  denote the distribution function of  $S_n$ . Let  $F(x)$  be a stable distribution function with exponent  $\alpha$  given by (2.2) and (5.2). Assume that

$$B_n^{-2} \{ \int_{-aB_n}^{aB_n} x^2 dF(x) - (\int_{-aB_n}^{aB_n} x dF(x))^2 \} \leq 1.$$

(We note that this is  $\sigma_{nk}^2(a) \leq 1$  so that by the proof of Theorem 2 this is a weak assumption.)

Then for,  $0 < r \leq 1, a > 1$ , we have

$$\begin{aligned}
 M_n &= \sup_{-\infty < x < +\infty} |F_n(x) - F(x)| \\
 &\leq kg^a(n, m(A, \delta), r) + n\{\bar{F}(-aB_n) + 1 - \bar{F}(aB_n)\},
 \end{aligned}$$

where  $k$  is a constant depending only on the bound of the derivative of  $F$ , and

$g^a(n, m(A, \delta), r)$

$$\begin{aligned}
 &= \left\{ \frac{n}{3B_n^4} \left[ \int_{-aB_n}^{aB_n} x^2 dF(x) - \left( \int_{-aB_n}^{aB_n} x dF(x) \right)^2 \right]^2 \right\}^{1/5} \\
 &+ \left\{ \frac{5}{6} \delta \left[ \frac{n}{B_n^2} \left\{ \int_{-aB_n}^{aB_n} x^2 dF(x) - \left( \int_{-aB_n}^{aB_n} x dF(x) \right)^2 \right\} + \frac{(c_1 + c_2)a^{2-\alpha}}{2-\alpha} \right] \right\}^{1/4} \\
 &+ \left\{ \frac{1}{2} \sum_{i=0}^m |K_n^a(x_i) - K^a(x_i)| \right\}^{1/3} \\
 &+ \left\{ \frac{4}{A} \left[ \frac{n}{B_n^2} \left( \int_{-aB_n}^{aB_n} x^2 dF(x) - \left( \int_{-aB_n}^{aB_n} x dF(x) \right)^2 \right) - K_n^a(A) + \frac{(c_1 + c_2)a^{2-\alpha}}{2-\alpha} \right. \right. \\
 &\quad \left. \left. - K^a(A) + K_n^a(-A) + K^a(-A) \right] + 2 \left| \frac{n}{B_n} \int_{-aB_n}^{aB_n} x dF(x) - \gamma - (c_1 - c_2) \right. \right. \\
 &\quad \left. \left. \times \left( \int_a^{+\infty} \frac{du}{(1+u^2)u^\alpha} - \int_0^a \frac{u^{2-\alpha} du}{1+u^2} \right) \right| \right\}^{1/2} + \left\{ \frac{8(c_1 + c_2)}{r(\alpha - r)a^{2-r}} \right\}^{1+r^{-1}}.
 \end{aligned}$$

The functions  $K_n^a(v)$  and  $K^a(v)$  are given by (5.12) and (5.6), and  $A$ ,  $\delta$ ,  $m(A, \delta)$  are given in Section 2.

**6. An example.** As an example we consider a sequence of independent random variables  $\{X_n\}$ ,  $n = 1, 2, \dots$  with a common distribution function  $\bar{F}(x)$ . Let  $\bar{F}(x)$  have density function

$$\bar{f}(x) = \pi^{-1}(1 - \cos x),$$

and consider the normed sums

$$S_n = n^{-1}(X_1 + \dots + X_n).$$

Again, let  $F_n(x)$  denote the distribution function of  $S_n$ . The characteristic function of  $\bar{F}(x)$  (c.f. [2] page 94 ff) is

$$\begin{aligned} \bar{\varphi}(t) &= 1 - |t| & \text{or } |t| \leq 1, \\ &= 0 & \text{for } |t| > 1 \end{aligned}$$

and hence the characteristic function of  $F_n(x)$  is

$$\begin{aligned} \varphi_n(t) &= (1 - n^{-1}|t|)^n & \text{for } |t| \leq n, \\ &= 0 & \text{for } |t| > n. \end{aligned}$$

Clearly  $\varphi_n(t)$  converges to  $\varphi(t) = e^{-|t|}$  which is the characteristic function of the well-known Cauchy distribution function

$$F(x) = \pi^{-1}(\frac{1}{2}\pi + \arctan x).$$

We shall establish that the rate of convergence of  $\sup_{-\infty < x < +\infty} |F_n(x) - F(x)|$  to zero is bounded by  $C/n^{1/15}$  where  $C$  is a constant.

As is well known,  $F(x)$  is stable with (c.f. [9] Section 4) the constants in (5.2) and (2.2) given by

$$(6.1) \quad \begin{aligned} \alpha &= 1, \\ c_1 = c_2 &= \pi^{-1} & \text{and} \\ \gamma &= 0. \end{aligned}$$

To apply the result of Theorems 1 or 3, we put  $X_{nk} = n^{-1}X_k$ . We have

$$\begin{aligned} \mu_{nk}(a) &= 0 = \mu_n(a), \\ \sigma_{nk}^2(a) &= \frac{1}{n^2} \int_{-na}^{na} x^2 \frac{1}{\pi} \frac{1 - \cos x}{x^2} dx \\ &= \frac{2}{\pi n^2} (na - \sin na), & \text{and} \\ \sigma_n^2(a) &= \frac{2}{\pi n} (na - \sin na). \end{aligned}$$

For  $-a < v < a$  we have

$$K_n^a(v) = \frac{n}{n^2} \int_{-na}^{nv} u^2 \frac{1 - \cos u}{\pi u^2} du$$

$$= \frac{1}{n\pi} [n(v+a) - (\sin nv + \sin na)].$$

Hence

$$K_n^a(v) = 0 \quad \text{for } v < -a,$$

$$= \frac{1}{n\pi} [n(v+a) - (\sin nv - \sin na)] \quad \text{for } -a \leq v < a,$$

$$= \frac{2}{n\pi} [na - \sin na] \quad \text{for } a \leq v.$$

From (6.1) and (5.3) we have  $dG(u) = \pi^{-1}(1+u^2)^{-1}du$ . Applying Lemma 7 we have

$$\mu(a) = 0, \quad \sigma^2(a) = 2a/\pi, \quad \text{and}$$

$$K^a(v) = 0 \quad \text{for } v < -a,$$

$$= (a+v)/\pi \quad \text{for } -a \leq v < a,$$

$$= 2a/\pi \quad \text{for } a \leq v.$$

For simplicity we let  $a = A = \delta^{-\frac{1}{2}}$ . Then we have

$$(6.2) \quad \left\{ \frac{1}{3} \sigma_n^2(a) \max \sigma_{nk}^2(a) \right\}^{1/5} \leq \left\{ \frac{16}{3\pi^2 n \delta} \right\}^{1/5},$$

$$(6.3) \quad \left\{ \frac{5}{6} \delta(\sigma_n^2(a) + \sigma^2(a)) \right\}^{1/4} \leq \left\{ \frac{5}{\pi} \sqrt{\delta} \right\}^{1/4}, \quad \text{and}$$

$$(6.4) \quad \left\{ \frac{1}{2} \sum_{i=0}^m |K_n^a(x_i) - K^a(x_i)| \right\}^{1/3} = \left\{ \frac{1}{2} \sum_{i=0}^m \frac{|\sin nx_i - \sin na|}{n\pi} \right\}^{1/3}$$

$$\leq \left\{ \frac{1}{2} (m+1) \frac{2}{n\pi} \right\}^{1/3} \leq \left\{ \frac{4}{n\pi \delta \sqrt{\delta}} \right\}^{1/3},$$

from the definition of  $m$ . Furthermore,

$$(6.5) \quad \{4A^{-1}(K_n^a(+\infty) - K_n^a(A) + K^a(+\infty) - K^a(A) + K_n^a(-A) + K^a(-A))$$

$$+ 2|\mu_n(a) - \mu(a)|\}^{1/2} = 0.$$

Applying Lemma 8, and (6.1)–(6.5) to  $g^a(n, m(A, \delta), r)$  as given in (2.9) we have

$$g^a(n, m(A, \delta), r) \leq \left\{ \frac{16}{3\pi^2 n \delta} \right\}^{1/5} + \left\{ \frac{5}{\pi} \sqrt{\delta} \right\}^{1/4}$$

$$+ \left\{ \frac{4}{\pi n \delta \sqrt{\delta}} \right\}^{1/3} + \left\{ \frac{16}{\pi r(1-r)} \right\}^{(1+r)^{-1}} \{\sqrt{\delta}\}^{(1-r)/(1+r)}.$$

Applying (5.13) we have

$$\begin{aligned} \sum_{k=1}^{k_n} \{F_{nk}(-a) + 1 - F_{nk}(a)\} &= n\{\bar{F}(-an) + 1 - \bar{F}(an)\} \\ &= n \frac{2}{\pi} \int_{an}^{+\infty} \frac{1 - \cos x}{x^2} dx \leq \frac{4}{\pi a} = \frac{4}{\pi} \sqrt{\delta}. \end{aligned}$$

From Theorem 1 or 3 we have

$$\begin{aligned} \sup_{-\infty < x < +\infty} |F_n(x) - F(x)| &\leq k \left\{ \frac{16}{3\pi^2 n \delta} \right\}^{1/5} + \left\{ \frac{5}{\pi} \sqrt{\delta} \right\}^{1/4} \\ &\quad + \left\{ \frac{4}{\pi n \delta \sqrt{\delta}} \right\}^{1/3} + \left\{ \frac{16}{\pi r(1-r)} \right\}^{(r+1)^{-1}} \{\sqrt{\delta}\}^{(1-r)/(1+r)} + \frac{4}{\pi} \sqrt{\delta}. \end{aligned}$$

Taking for example  $\delta = 1/n^{8/15}$ ,  $r = \frac{3}{5}$ , we find that

$$\sup_{-\infty < x < +\infty} |F_n(x) - F(x)| \leq C/n^{1/15},$$

where  $C$  is a constant.

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