

## ERROR ESTIMATION FOR A LIMIT THEOREM FOR DEPENDENT RANDOM VARIABLES<sup>1</sup>

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**1. Introduction.** Let  $(X_{nk}), k = 1, 2, \dots, k_n; n = 1, 2, \dots$  be a system of random variables. We investigate a limit theorem of a type studied by Loève [3] and particularly an error estimate for this theorem. The method used in finding the error estimate in this case applies to finding estimates for several of the theorems of [3].

**2. A convergence theorem for independent systems.** Let each  $X_{nk}$  have mean  $\mu_{nk}$  and variance  $\sigma_{nk}^2$  which we shall assume exists. Let  $S_n = \sum_{k=1}^{k_n} X_{nk}$  and  $\sigma_n^2 = \sum_{k=1}^{k_n} \sigma_{nk}^2$ . If for each  $n, X_{n1}, X_{n2}, \dots, X_{nk_n}$  are independent we say that  $(X_{nk})$  is an *independent system*. We write  $\mathcal{L}(S_n) \rightarrow \mathcal{L}(X)$  if  $F_n(x)$ , the distribution function of  $S_n$ , converges to  $F(x)$ , the distribution function of  $X$ , at each continuity point of  $X$ . We write  $\mathcal{L}(X) = \mathcal{L}(Y)$  when  $X$  and  $Y$  have the same distribution.

It is well known that if a random variable is infinitely divisible and has finite variance it can be represented by the formula of Kolmogorov [2] with unique real constant  $c$  and bounded nondecreasing function  $K(u)$  which is right continuous and  $K(-\infty) = 0$ . (Henceforth, we shall call a function with these properties a *Kolmogorov function*). Also it is known [2] that if  $(X_{nk})$  is an independent system of random variables having finite variances and such that  $(X_{nk} - \mu_{nk})$  is infinitesimal, then  $\mathcal{L}(S_n) \rightarrow \mathcal{L}(X)$  if there is a Kolmogorov function  $K(u)$  and a constant  $c$  such that as  $n \rightarrow \infty$

$$(2.1) \quad \sum_{k=1}^{k_n} \int_{-\infty}^u x^2 dF_{nk}(x + \mu_{nk}) \rightarrow K(u) \quad \text{at continuity points of } K(u),$$

$$(2.2) \quad \sum_{k=1}^{k_n} \int_{-\infty}^{\infty} x^2 dF_{nk}(x + \mu_{nk}) \rightarrow K(\infty),$$

$$(2.3) \quad \sum_{k=1}^{k_n} \mu_{nk} \rightarrow c,$$

where  $\mathcal{L}(X)$  is the infinitely divisible distribution determined by  $K(u)$  and  $c$ .

**3. A convergence theorem for dependent systems.** The following notation is the same as that used in [3] and [4] with the exception that distributions will be used in the usual sense (i.e.  $F(\infty) = P(-\infty < X < \infty) = 1$ ) rather than in the more generalized sense of Loève (i.e.  $F(\infty) \leq 1$ ). We recall that

$$F'_{nk}(x) = P(X_{nk} \leq x | \sum_{j=1}^{k-1} X_{nj})$$

$$E'(X_{nk}) = E(X_{nk} | \sum_{j=1}^{k-1} X_{nj})$$

$$\sigma'_{nk}{}^2 = E((X_{nk} - E(X_{nk}))^2 | \sum_{j=1}^{k-1} X_{nj}).$$

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We say  $S$  is a *constancy set* of the function  $K(u)$  if  $S = \bigcup_{v=1}^r I_v$  and  $I_v$  is an interval of constancy of  $K(u)$  and has for its endpoints continuity points of  $K(u)$ .

The following is representative of a class of limit theorems which can be obtained using the Comparison Theorem of Loève.

**THEOREM 1.** *Let  $(X_{nk})$  be a system of random variables such that  $(X_{nk} - \mu_{nk})$  is infinitesimal. Then  $\mathcal{L}(S_n) \rightarrow \mathcal{L}(X)$  if there exists a Kolmogorov function  $K(u)$  and a constant  $c$  such that as  $n \rightarrow \infty$ , (2.1), (2.2) and (2.3) hold,*

$$(3.1) \quad \sum_{k=1}^{k_n} E |E'(X_{nk}) - E(X_{nk})| \rightarrow 0,$$

*there is a constancy set  $S$  ( $S$  can depend on  $n$  and/or  $k$ ) of  $k(u)$ , whose complement,  $S^c$ , is a bounded set, such that*

$$(3.2) \quad \sum_{k=1}^{k_n} E \left( \int_{S^c} |d(F'_{nk}(x + \mu_{nk}) - F_{nk}(x + \mu_{nk}))| \right) \rightarrow 0.$$

*The random variable  $X$  has the infinitely divisible distribution determined by  $K(u)$  and  $c$ .*

**PROOF.** The proof follows from Loève's Comparison Theorem ([3] or [4]) since

$$\begin{aligned} & \sum_{k=1}^{k_n} E \left| \int_S x d(F'_{nk}(x + \mu_{nk}) - F_{nk}(x + \mu_{nk})) \right| \\ & \leq \sum_{k=1}^{k_n} E |E'(X_{nk}) - E(X_{nk})| \\ & \quad + (\max_{x \in S^c} |x|) \cdot \sum_{k=1}^{k_n} E \left( \int_{S^c} |d(F'_{nk}(x + \mu_{nk}) - F_{nk}(x + \mu_{nk}))| \right) \end{aligned}$$

and

$$\sum_{k=1}^{k_n} E \left| \int_S |x|^2 |d(F'_{nk}(x + \mu_{nk}) - F_{nk}(x + \mu_{nk}))| \right| \leq 2 \sum_{v=1}^r \sum_{k=1}^{k_n} \int_{I_v} x^2 dF_{nk}(x + \mu_{nk}).$$

In this finite variance case a similar theorem has been obtained by the author for convergence to the Poisson distribution. Convergence theorems to the normal distribution are proven in [3] and [4].

**4. Error estimates.** We consider the behavior of  $M_n = \sup_{-\infty < x < \infty} |F_n(x) - F(x)|$  where  $F_n(x)$  is the distribution function of  $S_n$  and  $F(x)$  is the distribution function of  $X$  in Theorem 1.

**LEMMA 1.** *Let  $K_1(x)$  and  $K_2(x)$  be functions of the form*

$$K_j(x) = \int_{-\infty}^x u^2 dF_j(u + \mu_j)$$

*where  $F_j(x)$  is the distribution function of a random variable  $Y_j$  which has mean  $\mu_j$  and variance  $\sigma_j^2$  for  $j = 1, 2$ .*

*Then*

$$\begin{aligned} & \left| \int_{-\infty}^{\infty} x^{-2} (e^{itx} - 1 - itx) d(K_1(x) - K_2(x)) \right| \\ & \leq (5/4) |t|^3 \delta (\sigma_1^2 + \sigma_2^2) + \frac{1}{2} t^2 \sum_{j=0}^m |K_1(x_j) - K_2(x_j)| \\ & \quad + 2 |t| A^{-1} [K_1(\infty) - K_1(A) + K_2(\infty) - K_2(A) + K_1(-A) + K_2(-A)] \end{aligned}$$

for any  $A > 0$ ,  $0 < \delta \leq 2A$  with  $m = [2A/\delta] + 1$  and any  $x_j$ ,  $j = 0, 1, 2, \dots, m$ , satisfying  $-A = x_0 < x_1 < \dots < x_m = A$  and  $\max_{1 \leq j \leq m} |x_j - x_{j-1}| < \delta$ . (The proof of this result is the same as the proof of the first part of Lemma 3 page 619 of [5].)

Let  $(X_{nk})$  be a system of random variables and let  $(X_{nk}^*)$  be its independent version (i.e.  $\mathcal{L}(X_{nk}) = \mathcal{L}(X_{nk}^*)$  and  $(X_{nk}^*)$  is an independent system). Also define

$$K_{nk}(x) = \int_{-\infty}^x u^2 dF_{nk}(u + \mu_{nk}).$$

$$K'_{nk}(x) = \int_{-\infty}^x u^2 dF'_{nk}(u + \mu_{nk}).$$

$$K_n(x) = \sum_{k=1}^{k_n} K_{nk}(x).$$

LEMMA 2. Let  $(X_{nk})$  be a system of random variables and let  $(X_{nk}^*)$  be its independent version. Then for any  $A > 0, 0 \leq \delta < 2A, m = [2A/\delta] + 1$  and any  $x_j, j = 0, 1, 2, \dots, m$ , satisfying  $-A = x_0 < x < \dots < x_m = A$  and  $\max_{1 < j < m} |x_j - x_{j-1}| < \delta$  we have

$$|f_n(t) - f_n^*(t)| \leq |t| \sum_{k=1}^{k_n} E |E'(X_{nk}) - E(X_{nk})| + (5/2) |t|^3 \delta \sigma_n^2 + \frac{1}{2} t^2 \sum_{j=0}^m \sum_{k=1}^{k_n} E |K'_{nk}(x_j) - K_{nk}(x_j)| + 4 |t| \sigma_n^2 / A$$

where  $\sigma_n^2 = \sum_{k=1}^{k_n} \sigma_{nk}^2, f_n(t)$  and  $f_n^*(t)$  are the characteristic functions of  $S_n$  and  $S_n^*$ .

PROOF. Let  $\phi'_{nk}(t) = E(e^{it}(X_{nk} - \mu_{nk}) | \sum_{l=1}^{k-1} X_{nl})$  and  $\phi_{nk}(t) = E(e^{it}(X_{nk}^* - \mu_{nk}))$ . Using the same proof as in Lemma B page 65 of [3] we have the following inequality  $|f_n(t) - f_n^*(t)| \leq \sum_{k=1}^{k_n} E |\phi'_{nk}(t) - \phi_{nk}(t)|$ . By Lemma 1, at each value of the random variable  $K'_{nk}(x)$  we have,

$$|\int_{-\infty}^{\infty} x^{-2} (e^{itx} - 1 - itx) d(K'_{nk}(x) - K_{nk}(x))| \leq (5/4) |t|^3 \delta \sigma_{nk}'^2 + \sigma_{nk}^2 + \frac{1}{2} t^2 \sum_{j=0}^m |K'_{nk}(x_j) - K_{nk}(x_j)| + 2A^{-1} |t| |K'_{nk}(\infty) - K'_{nk}(A) + K_{nk}(\infty) - K_{nk}(A) + K'_{nk}(-A) + K_{nk}(-A)|.$$

Finally

$$|f_n(t) - f_n^*(t)| \leq \sum_{k=1}^{k_n} E |\int_{-\infty}^{\infty} x^{-2} (e^{itx} - 1 - itx) d(K'_{nk}(x) - K_{nk}(x)) + it(E'(X_{nk}) - \mu_{nk})| \leq |t| \sum_{k=1}^{k_n} E |E'(X_{nk}) - E(X_{nk})| + (5/2) |t|^3 \delta \sigma_n^2 + \frac{1}{2} t^2 \sum_{j=0}^m \sum_{k=1}^{k_n} E |K'_{nk}(x_j) - K_{nk}(x_j)| + 4t \sigma_n^2 / A$$

which is the required inequality.

We now investigate the behavior of  $M_n$  for Theorem 1. A bound is first obtained for this quantity. Then under the conditions of Theorem 1, it is shown that the bound converges to 0 as  $n \rightarrow \infty$ .

THEOREM 2. Let  $X$  be an infinitely divisible random variable with distribution function  $F(x)$ , mean  $\mu$ , variance  $\sigma^2$ , characteristic function  $f(t)$ , and Kolmogorov function  $K(x)$ . Let  $(X_{nk})$  be a system of random variables with  $\sigma_{nk}^2 \leq 1$  for each  $X_{nk}$  and assume  $dF(x)/dx \leq B$  for all  $x$  where  $B$  is some positive constant. Then it follows that for any number  $a > 1$  we have

$$M_n = \sup_{-\infty < x < \infty} |F_n(x) - F(x)| \leq k(a, B)g(n, m(A, \delta))$$

where  $k(a, B)$  is a constant depending only on  $a$  and  $B$ ,

$$\begin{aligned}
 g(n, m(A, \delta)) &= [(5/16)n^2 \max_{1 \leq k \leq k_n} \sigma_{nk}^2]^{1/5} + [(5/6)\delta(3\sigma_n^2 + \sigma^2)]^{\frac{1}{2}} \\
 &\quad + [\frac{1}{2} \sum_{i=0}^m \sum_{k=1}^{k_n} E |K'_{nk}(x_i) - K_{nk}(x_i)| + \frac{1}{2} \sum_{i=0}^m |K_n(x_i) - K(x_i)|]^{\frac{1}{2}} \\
 &\quad + [2 \sum_{k=1}^{k_n} E |E'(X_{nk}) - E(X_{nk})| + 2 |\mu_n - \mu| \\
 &\quad + 4A^{-1}(2\sigma_n^2 + K_n(\infty) - K_n(A) \\
 &\quad + K(\infty) - K(A) + K_n(-A) + K(-A))]^{\frac{1}{2}},
 \end{aligned}$$

and  $\sigma_n^2, \delta, A, m(A, \delta) = m$ , and  $\{x_i\}$  are as defined in Lemma 2. (The numbers  $A$  and  $\delta$  depend on  $n$ ).

PROOF. We let  $(X_{nk}^*)$  be the independent version of  $(X_{nk})$ . We then have

$$|f_n(t) - f(t)| \leq |f_n(t) - f_n^*(t)| + |f_n^*(t) - f(t)|$$

where Lemma 2 may be applied to the first term on the right-hand side of the inequality. The second term may be bounded by  $h(t, n, m(A, \delta))$  for  $|t| \leq T_n$  where

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

and

$$\begin{aligned}
 T_n &= [(1/3)\sigma_n^2 \max_{1 \leq k \leq k_n} \sigma_{nk}^2]^{1/5} + ((5/6)\delta(\sigma_n^2 + \sigma^2))^{\frac{1}{2}} + (\frac{1}{2} \sum_{i=0}^m |K_n(x_i) - K(x_i)|)^{\frac{1}{2}} \\
 &\quad + (4A^{-1} \{K_n(\infty) - K_n(A) - K(\infty) - K(A) + K_n(-A) + K(-A)\} + 2 |\mu_n - \mu|)^{\frac{1}{2}})^{-1}.
 \end{aligned}$$

This follows from the proof of Theorem 3 of [5]. Letting  $T_n' = (g(n, m(A, \delta)))^{-1}$  it is observed that  $T_n' \geq T_n$ . It is then easily seen that

$$\int_{-T_n'}^{T_n'} |(f_n(t) - f(t))/t| dt \leq g(n, m(A, \delta)).$$

Then applying a result of Esseen [1] we obtain that for every  $a > 1$ , there is a finite positive number  $c(a)$  depending only on  $a$ , such that

$$\begin{aligned}
 M_n &\leq (2\pi)^{-1} a \int_{-T_n'}^{T_n'} |(f_n(t) - f(t))/t| dt + c(a) \cdot B \cdot (T_n')^{-1} \\
 &\leq (2\pi)^{-1} a g(n, m(A, \delta)) + c(a) \cdot B \cdot g(n, m(A, \delta)) \\
 &= k(a, B) \cdot g(n, m(A, \delta))
 \end{aligned}$$

where  $k(a, B) = (2\pi)^{-1} \cdot a + c(a) \cdot B$  which depends only on  $a$  and  $B$ . This completes the proof.

In Theorem 2 let  $\sigma < \delta < 1$  be such that  $\pm \delta$  are continuity points of  $K(x)$  and choose  $A = \delta^{-\frac{1}{2}}$ . Let  $m(\delta) = m(\delta^{-\frac{1}{2}}, \delta) = m(A, \delta)$ .

We now show that the bound obtained in Theorem 2 converges to zero as  $n \rightarrow \infty$  under the conditions of Theorem 1.

**THEOREM 3.** Let  $(X_{nk})$  be a system of random variables satisfying the conditions of Theorem 1. Let  $F(x)$  be the limit distribution and assume that there is a positive constant  $B$  such that  $dF(x)/dx < B$  for all  $x$ . Then there exists a sequence  $(\delta_n)$  with  $0 < \delta_n < 1$  such that  $\delta_n \rightarrow 0$  as  $n \rightarrow \infty$  and  $g(n, m(\delta_n)) \rightarrow 0$  as  $n \rightarrow \infty$ .

**PROOF.** We first show how the  $(\delta_n)$  are chosen. Let  $S$  be a set satisfying (3.2). Then

$$\sum_{k=1}^{k_n} E |K'_{nk}(x_i) - K_{nk}(x_i)| \leq 2 \sum_{k=1}^{k_n} \int_S x^2 d(F_{nk}(x + \mu_{nk}) + (\max_{x \in S^c} x^2) \sum_{k=1}^{k_n} E(\int_{S^c} |d(F'_{nk}(x + \mu_{nk}) - F_{nk}(x + \mu_{nk}))|))$$

which converges to 0 as  $n \rightarrow \infty$  by (2.1) and (3.2). Thus for each  $\delta < 0$

$$(4.1) \quad \lim_{n \rightarrow \infty} \sum_{i=0}^{m(\delta)} \sum_{k=1}^{k_n} E |K'_{nk}(x_i) - K_{nk}(x_i)| = 0.$$

We also have  $\lim_{n \rightarrow \infty} K_n(x_i) = K(x_i)$  by (2.1) since  $(x_i)$  were chosen to be continuity points of  $K(x)$ . Thus for each  $\delta < 0$ ,

$$(4.2) \quad \lim_{n \rightarrow \infty} \sum_{i=0}^{m(\delta)} |K_n(x_i) - K(x_i)| = 0.$$

It is then possible to choose  $(\delta_n)$  with  $0 < \delta_n < 1$  such that  $\pm \delta_n^{-\frac{1}{2}}$  are continuity points of  $K(x)$ ,  $\lim_{n \rightarrow \infty} \delta_n = 0$ , and (4.1) and (4.2) hold with  $\delta$  replaced by  $\delta_n$ .

We need only notice that under the conditions of Theorem 1  $\lim_{n \rightarrow \infty} \sigma_n^2 = \sigma^2$ , where  $\sigma^2$  is the variance of the limit distribution, and  $\lim_{n \rightarrow \infty} \max_{1 \leq k \leq k_n} \sigma_{nk}^2 = 0$ . Thus the theorem follows.

Also notice that under the conditions of the previous theorem  $\sigma_{nk}^2 \leq 1$  for large  $n$  since  $\lim_{n \rightarrow \infty} \max_{1 \leq k \leq k_n} \sigma_{nk}^2 = 0$ . This condition was needed in Theorem 2 to obtain the bound and is not a serious restriction.

Using similar techniques bounds can be obtained for the aforementioned convergence theorems to the normal and the Poisson law.

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