

On the Convergence of Convolutions of Distributions with Regularly Varying Tails

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Summary. Let x_1, \dots, x_n be a sequence of independent random variables with the common distribution F . Suppose $E x_k = 0$ and that F belongs to the domain of attraction of the normal distribution. Under conditions which do not involve the existence of any particular moment we show that

$$P \{x_1 + \dots + x_n \leq x a_n\} = \Phi(x) - \frac{n}{a_n^3} \int_{-a_n}^{a_n} |y|^3 F(dy) (\omega \Phi(x) + o(1))$$

uniformly in x , provided the norming constants a_1, a_2, \dots are properly chosen. Here Φ is the standard normal distribution and ω a certain operator (depending on F).

The local counterparts are also treated.

1. Introduction and Results

Suppose x_1, \dots, x_n are independent, identically distributed random variables with zero expectation and with a distribution F that belongs to the domain of attraction of the normal distribution. Then there is a sequence of numbers $a_1, a_2, \dots \rightarrow \infty$ satisfying

$$\frac{n}{a_n^2} \int_{-a_n}^{a_n} x^2 F(dx) \rightarrow 1 \quad \text{as } n \rightarrow \infty. \tag{1}$$

Obviously $a_n \sim \sqrt{n} \sigma$ if the variance exists and equals σ^2 . Let F_n denote the distribution of the normed sum $a_n^{-1}(x_1 + \dots + x_n)$. If we require that the third moment μ_3 exists and that F is not a lattice distribution, then (Esséen [2] p. 49)

$$F_n(x) = \Phi(x) - \frac{\mu_3}{6\sigma^3\sqrt{n}}(x^2 - 1)\varphi(x) + o\left(\frac{1}{\sqrt{n}}\right)$$

uniformly in x . Here Φ stands for the normal distribution with zero expectation and unit variance and φ for its density.

The aim of the present paper is to prove an analogous result when the third moment does not exist. (Concerning estimates of the Berry-Esséen type in this case, see Feller [5], Esséen [3] and the references given there.) Instead we have to impose other conditions. To formulate these we shall say that a positive function R defined on $(0, \infty)$ varies regularly with exponent α ($-\infty < \alpha < \infty$) if, for each $x > 0$, $\frac{R(tx)}{R(t)} \rightarrow x^\alpha$ as $t \rightarrow \infty$. R varies slowly if R varies regularly with exponent zero.

Our conditions will be: for some slowly varying function L , as $x \rightarrow \infty$,

$$\int_{-x}^x |y|^3 F(dy) \sim x^{1-\delta} L(x), \quad 0 \leq \delta \leq 1, \tag{2}$$

and

$$\frac{\int_0^x y^3 F(dy)}{\int_{-x}^x |y|^3 F(dy)} \rightarrow p, \quad \frac{\int_{-x}^0 |y|^3 F(dy)}{\int_{-x}^x |y|^3 F(dy)} \rightarrow q. \tag{3}$$

Except when $\delta = 1$, (2) and (3) are equivalent to

$$1 - F(x) + F(-x) \sim \frac{1-\delta}{2+\delta} x^{-2-\delta} L(x) \tag{2'}$$

and

$$\frac{1 - F(x)}{1 - F(x) + F(-x)} \rightarrow p, \quad \frac{F(-x)}{1 - F(x) + F(-x)} \rightarrow q. \tag{3'}$$

The verification of this equivalence is immediate, given Lemma 1 below. Conditions similar to those above have proved successful in connection with large deviation probabilities, see Heyde [8] and especially Heyde [6] where a result of McLaren concerning the convergences toward a normal distribution is cited.

Introduce the truncation function

$$\tau(x) = x \text{ if } |x| \leq 1, = 0 \text{ if } |x| > 1$$

and the functions

$$\Omega_\delta(x) = \begin{cases} p x^{1-\delta}, & x > 0, \\ q |x|^{1-\delta}, & x < 0, \end{cases}$$

where p and q are defined in (3). The results will be formulated in terms of the derivation operator D and the operators ω_δ and $\tilde{\omega}_\delta$ defined, for those u for which the definitions make sense, by

$$\begin{aligned} \omega_\delta u(x) &= \frac{1}{\delta(\delta+1)(\delta+2)} \int u(x-y) \Omega_\delta(dy), \quad 0 < \delta \leq 1 \\ \tilde{\omega}_\delta u(x) &= \int \frac{u(x-y) - u(x) + y Du(x) - \frac{1}{2} \tau(y)^2 D^2 u(x)}{y^3} \Omega_\delta(dy), \quad 0 \leq \delta \leq 1. \end{aligned} \tag{4}$$

For later use we here mention that $\tilde{\omega}_\delta \Phi$ has the Fourier-Stieltjes transform $e^{-t^2/2} h(t)$, where

$$h(t) = \int \frac{e^{itx} - 1 - itx - \frac{1}{2}(it\tau(x))^2}{x^3} \Omega_\delta(dx). \tag{5}$$

This is seen if we use the representation (4) of $\tilde{\omega}_\delta \Phi$, then apply Fubinis theorem and remember that

$$\int e^{itx} D^k \Phi(dx) = (-it)^k e^{-t^2/2}, \quad k = 0, 1, \dots$$

The following theorem is closely related to the theorem of Bergström, [1] and also to that of Heyde, [7] in which he shows that (with $a_n = \sigma \sqrt{n}$ and without

any extra assumptions such as ours (2) and (3)).

$$\sum_{n=1}^{\infty} n^{-1+\delta/2} \sup_x |F_n(x) - \Phi(x)| < \infty$$

if and only if

$$E|x_k|^{2+\delta} < \infty, \quad 0 < \delta < 1, \quad E|x_k| \log(1+|x_k|) < \infty, \quad \delta = 0.$$

Theorem 1. *Suppose F satisfies (2) and (3) and has an infinite absolute third moment.*

(i) *If $\delta > 0$ we choose $a_n = \sigma \sqrt{n}$, where $\sigma^2 = \int x^2 F(dx)$; then*

$$F_n(x) = \left(1 - \frac{L(\sqrt{n})}{\sigma^{2+\delta} n^{\delta/2}} \omega_\delta D^3 \right) \Phi(x) + o\left(\frac{L(\sqrt{n})}{n^{\delta/2}}\right) \tag{6}$$

uniformly in x .

(ii) *If $\delta = 0$ then it is possible to choose the sequence a_1, a_2, \dots such that*

$$\frac{n}{a_n^2} \int_{-a_n}^{a_n} y^2 F(dy) = 1 + o\left(\frac{n L(a_n)}{a_n^2}\right). \tag{7}$$

With such a choice of a_1, a_2, \dots ,

$$F_n(x) = \left(1 + \frac{n L(a_n)}{a_n^2} \tilde{\omega}_0 \right) \Phi(x) + o\left(\frac{n L(a_n)}{a_n^2}\right)$$

uniformly in x .

Remark. A choice of a_1, a_2, \dots according to (7) gives the fastest possible rate of convergence if $\delta = 0$; if $\delta > 0$ we have nothing to gain by another choice than $a_n = \sigma \sqrt{n}$. In order to show this, it will in the proof of Theorem 1 be shown that $F_n(x) = (1 + c_n \tilde{\omega}_\delta + \rho_n \frac{1}{2} D^2) \Phi(x) + o(c_n) + o(\rho_n)$ uniformly in x , where

$$c_n = \frac{n}{a_n^3} \int_{-a_n}^{a_n} |y|^3 F(dy) \sim \frac{n L(a_n)}{a_n^{2+\delta}}, \quad \rho_n = \frac{n}{a_n^2} \int_{-a_n}^{a_n} y^2 F(dy) - 1. \tag{8}$$

We first show that the choice of a_n does not influence the order of magnitude of c_n . Suppose $a_1^{(i)}, a_2^{(i)}, \dots, i = 1, 2$, are two sequences satisfying (1), then

$$a_n^{(1)} \sim a_n^{(2)}. \tag{9}$$

Let $c_n^{(i)}$ and $\rho_n^{(i)}$ stand for the c_n and ρ_n corresponding to $a_n^{(i)}, i = 1, 2$. Now $c_n^{(i)} = n a_n^{(i)-3} R(a_n^{(i)})$, where R is a nondecreasing, regularly varying function; because of (9) we therefore have $c_n^{(1)} \sim c_n^{(2)}$.

Thus we obtain the fastest possible rate of convergence if we choose a_n such that $\rho_n = O(c_n)$; this is certainly satisfied if we choose a_n according to (7). If the variance σ^2 exists and $a_n = \sigma \sqrt{n}$, then

$$-\rho_n = \frac{1}{\sigma^2} \int_{|y| > \sigma \sqrt{n}} y^2 F(dy).$$

It follows from Lemma 1 that in this case

$$-\frac{\rho_n}{c_n} \rightarrow \begin{cases} \frac{1-\delta}{\delta} & \text{if } \delta > 0, \\ \infty & \text{if } \delta = 0. \end{cases} \tag{10}$$

This shows that $\rho_n = O(c_n)$ if $\delta > 0$ and $a_n = \sigma \sqrt{n}$; however, if $\delta = 0$, then a choice according to (7) is preferable to the choice $a_n = \sigma \sqrt{n}$.

Theorems 2 and 3 are the local counterparts of Theorem 1, for absolutely continuous and lattice distributions respectively.

Theorem 2. *Suppose F satisfies the conditions of Theorem 1, is absolutely continuous, and has characteristic function $f \in L$ for some $1 \leq r < \infty$. Denote the density of F_n by v_n .*

(i) *If $\delta > 0$ we choose $a_n = \sigma \sqrt{n}$, where $\sigma^2 = \int x^2 F(dx)$; then*

$$v_n(x) = \left(1 - \frac{L(\sqrt{n})}{\sigma^{2+\delta} n^{\delta/2}} \omega_\delta D^3 \right) \varphi(x) + o\left(\frac{L(\sqrt{n})}{n^{\delta/2}}\right)$$

uniformly in x .

(ii) *If $\delta = 0$ then it is possible to choose the sequence a_1, a_2, \dots such that (7) is valid. With such a choice of a_1, a_2, \dots*

$$v_n(x) = \left(1 + \frac{n L(a_n)}{a_n^2} \tilde{\omega}_0 \right) \varphi(x) + o\left(\frac{n L(a_n)}{a_n^2}\right)$$

uniformly in x .

If F is a lattice distribution with span h then the variables x_j are restricted to values of the form $b, b \pm h, b \pm 2h, \dots$ and h is the largest positive number with this property. The atoms of the distribution of $a_n^{-1}(x_1 + \dots + x_n)$ are among the points of form $x = a_n^{-1}(nb + kh)$, where $k = 0, \pm 1, \pm 2, \dots$. For such x we define

$$p_n(x) = P(a_n^{-1}(x_1 + \dots + x_n) = x).$$

Theorem 3. *Suppose F is a lattice distribution with span h and assume that F satisfies the conditions of Theorem 1. Let x be of form $a_n^{-1}(nb + kh)$, k integer.*

(i) *If $\delta > 0$ we choose $a_n = \sigma \sqrt{n}$, where $\sigma^2 = \int x^2 F(dx)$, then*

$$p_n(x) = \left(1 - \frac{L(\sqrt{n})}{\sigma^{2+\delta} n^{\delta/2}} \omega_\delta D^3 \right) \varphi(x) + o\left(\frac{L(\sqrt{n})}{n^{\delta/2}}\right)$$

uniformly in x .

(ii) *If $\delta = 0$ then it is possible to choose the sequence a_1, a_2, \dots such that (7) is valid. With such a choice of a_1, a_2, \dots*

$$p_n(x) = \left(1 + \frac{n L(a_n)}{a_n^2} \tilde{\omega}_0 \right) \varphi(x) + o\left(\frac{n L(a_n)}{a_n^2}\right)$$

uniformly in x .

2. Three Lemmas

The following lemma can be found in problem 30, p. 279 of [4].

Lemma 1. *Let F be a probability distribution concentrated on $(0, \infty)$. Put*

$$U_\zeta(x) = \int_0^x y^\zeta F(dy), \quad V_\sigma(x) = \int_x^\infty y^{-\sigma} F(dy).$$

If F possesses moments of order $< \alpha$, but not of order $> \alpha > 0$, then either U_ζ and V_σ vary regularly for all $\zeta > \alpha$ and all $\sigma > -\alpha$ or for no such value. In the first case the exponents are $\zeta - \alpha$ and $\sigma - \alpha$, and this case arises if and only if

$$\frac{x^{\sigma+\zeta} V_\sigma(x)}{U_\zeta(x)} \rightarrow \frac{\zeta - \alpha}{\sigma + \alpha} \tag{11}$$

($0 < \alpha < \zeta$). If U_ζ varies slowly, (11) holds with $\alpha = \zeta$; if V_σ varies slowly the left side tends to ∞ .

In applications of Lemma 1 to distributions not concentrated on $(0, \infty)$ we treat the two tails separately.

We also state a lemma which is contained in Theorem 2a, p. 32 of [2].

Lemma 2. Let F be a probability distribution with characteristic function f . Let G be a real function of bounded variation over the whole real axis, such that $G(-\infty) = 0$, $G(+\infty) = 1$, while the derivative G' exists everywhere and satisfies $|G'| < K$ for some constant K . Write

$$g(t) = \int e^{itx} G(dx).$$

Suppose that for some positive constants T and η we have

$$\int_{-T}^T \left| \frac{f(t) - g(t)}{t} \right| dt = \eta.$$

Then there are positive constants A and B independent of T and η such that for all real x

$$|F(x) - G(x)| < A\eta + \frac{B}{T}.$$

Finally we need a bound for characteristic functions of distributions in the domain of attraction of the normal distribution.

Lemma 3. Let F be a probability distribution with characteristic function f . Suppose $\int_{-x}^x y^2 F(dy)$ is slowly varying and the sequence a_1, a_2, \dots satisfies (1). Then there are positive constants ε and c such that

$$|f^n(t/a_n)| \leq e^{-c|t|} \tag{12}$$

for $1 < |t| < \varepsilon a_n$ and all sufficiently large n .

Proof. We may without loss of generality assume that $\int x F(dx) = 0$. Introduce the symmetrized distribution ${}^0F(x) = \int F(x+y)F(dy)$; then also

$$M(x) = \int_{-x}^x y^2 {}^0F(dy)$$

is slowly varying, and

$$\frac{n}{a_n^2} M(a_n) \rightarrow 2. \tag{13}$$

This becomes obvious if we use the fact that a distribution G with zero expectation belongs to the domain of attraction of the normal distribution if and only if $\int_{-x}^x y^2 G(dy)$ varies slowly. The variance of the corresponding normal distribution then equals the limit $\lim_n \frac{n}{a_n^2} \int_{-a_n}^{a_n} y^2 G(dy)$ if we use a_n as norming constants. (See [3] p. 544.) Further 0F has characteristic function $|f(t)|^2$ and because 0F is symmetric

$$1 - |f(t)|^2 = \int (1 - \cos tx) {}^0F(dx).$$

Now $\frac{1 - \cos y}{y^2}$ is non negative for each y and larger than $\frac{2}{3}$ for $|y| \leq 1$, hence

$$1 - |f(t)|^2 \geq \int_{|x| \leq 1/|t|} \frac{1 - \cos tx}{(tx)^2} (tx)^2 {}^0F(dx) \geq \frac{2}{3} t^2 M\left(\frac{1}{|t|}\right).$$

From this and the inequality $1 - x \leq e^{-x}$ valid for all $x \geq 0$ we obtain

i. e.,
$$\begin{aligned} |f(t)|^2 &\leq e^{-\frac{2}{3} t^2 M(1/|t|)}, \\ |f(t)| &\leq e^{-\frac{1}{3} t^2 M(1/|t|)}. \end{aligned}$$

Thus it suffices to show that there are positive numbers ε and c such that

$$|t| \frac{n}{a_n^2} M\left(\frac{a_n}{|t|}\right) > c \tag{14}$$

for $1 < |t| < \varepsilon a_n$ and all sufficiently large n . From the representation

$$M(x) = b(x) \exp\left(\int_1^x \eta(y) \frac{dy}{y}\right)$$

where $b(x) \rightarrow b > 0$ and $\eta(x) \rightarrow 0$ as $x \rightarrow \infty$ (see [4], p. 274) it now follows that

$$\frac{M\left(\frac{a_n}{|t|}\right)}{M(a_n)} \geq \frac{b\left(\frac{a_n}{|t|}\right)}{b(a_n)} \exp\left(-\sup_{y > a_n/|t|} |\eta(y)| \log |t|\right) \geq \frac{1}{2} |t|^{-1}$$

for $1 \leq |t| \leq \varepsilon a_n$ and all sufficiently large n , if $\varepsilon > 0$ is chosen so small that $|\eta(y)| < 1$ and $b(y) > \frac{2}{3} b$ for all $y > 1/\varepsilon$.

Because of (13), (14) follows.

3. Proof of Theorem 1

Below we will be concerned with the sequence U_1, U_2, \dots of measures defined by

$$U_n(dx) = \frac{(a_n x)^3 F(a_n dx)}{\int_{-a_n}^{a_n} |y|^3 F(dy)}$$

Let us point out that

$$U_n(I) \rightarrow \int_I \Omega_\delta(dx) \tag{15}$$

for each bounded interval I whose boundary has Ω_δ -measure zero, and that for each $\varepsilon > 0$ there is an $x > 0$ such that for all n

$$\int_{|y|>x} y^{-2} |U_n|(dy) < \varepsilon. \tag{16}$$

The first of these statements is a direct consequence of (2) and (3). If we treat each tail of U_n separately and apply Lemma 1 we get

$$\int_{|y|>x} y^{-2} |U_n|(dy) \rightarrow \frac{1-\delta}{1+\delta} x^{-1-\delta} \quad \text{as } n \rightarrow \infty$$

from which the second statement follows.

Denote the characteristic function of F by f . Using the identity

$$u^n - v^n = (u - v)(u^{n-1} + u^{n-2}v + \dots + v^{n-1})$$

with $u = f(t/a_n)$ and $v = \exp\left(-\frac{1}{2} \frac{t^2}{n}\right)$ we obtain

$$f^n(t/a_n) - e^{-t^2/2} = d_n(t) S_n(t) \tag{17}$$

where

$$d_n(t) = n \left[f(t/a_n) - \exp\left(-\frac{1}{2} \frac{t^2}{n}\right) \right],$$

$$S_n(t) = \frac{1}{n} \sum_{j=0}^{n-1} f^j(t/a_n) \exp\left(-\frac{1}{2} t^2 \frac{n-j-1}{n}\right).$$

Remembering that F has zero expectation we get trivially

$$f(t/a_n) = \int \left[e^{itx} - 1 - itx - \frac{1}{2}(it\tau(x))^2 \right] F(a_n dx) + 1 + \frac{1}{2}(it)^2 \int \tau(x)^2 F(a_n dx).$$

Hence

$$d_n(t) = c_n h_n(t) + \rho_n \frac{1}{2}(it)^2 + r_n(t) \tag{18}$$

where c_n and ρ_n are defined in (8), and

$$h_n(t) = \int \frac{e^{itx} - 1 - itx - \frac{1}{2}(it\tau(x))^2}{x^3} U_n(dx),$$

$$r_n(t) = -n \left[\exp\left(-\frac{1}{2} \frac{t^2}{n}\right) - 1 + \frac{1}{2} \frac{t^2}{n} \right].$$

From (15) and (16) we obtain $h_n(t) \rightarrow h(t)$, where h is defined in (5). The difference $S_n(t) - e^{-t^2/2}$ we estimate in the following way:

$$\begin{aligned} \left| \frac{1}{n} \sum_{j=0}^{n-1} u^j v^{n-1-j} - v^n \right| &= \left| \frac{1}{n} \sum_{j=0}^{n-1} v^{n-j-1} (u^j - v^j) + v^{n-1} - v^n \right| \\ &\leq \frac{1}{n} \sum_{j=0}^{n-1} j |u - v| + |1 - v| \leq n |u - v| + |1 - v|. \end{aligned}$$

(Here we have used the fact that $|u| \leq 1, |v| \leq 1$ and that the inequality $|u^j - v^j| \leq j|u - v|$ is valid for such u and v .) Now $f^n(t/a_n) \rightarrow e^{-t^2/2}$ implies $n(f(t/a_n) - 1) \rightarrow -t^2/2$. Hence $S_n(t) \rightarrow e^{-t^2/2}$. The inequality $|e^{-x} - 1 + x| \leq \frac{1}{2}x^2$ valid for $x \geq 0$ gives

$$|r_n(t)| \leq \frac{1}{n}t^4/8. \tag{19}$$

A glance at (17) and (18) now suggest introducing the functions g_n defined by

$$g_n(t) - e^{-t^2/2} = [c_n h(t) + \rho_n \frac{1}{2}(it)^2] e^{-t^2/2}; \tag{20}$$

then we have

$$f^n(t/a_n) - g_n(t) = r_n(t) S_n(t) + c_n [(h_n(t) - h(t)) e^{-t^2/2} + h_n(t)(S_n(t) - e^{-t^2/2})] + \rho_n \frac{1}{2}(it)^2 [S_n(t) - e^{-t^2/2}].$$

Furthermore we have in connection with the definition (5) of h shown that g_n is the Fourier-Stieltjes transform of

$$G_n(x) = (1 + c_n \tilde{\omega}_\delta + \rho_n \frac{1}{2} D^2) \Phi(x).$$

Given $\varepsilon > 0$ we put $T = (\varepsilon c_n)^{-1} B$ in Lemma 2 and obtain

$$|F_n(x) - G_n(x)| \leq A(I_1 + I_2 + I_3) + \varepsilon c_n$$

where

$$I_1 = \int_{-T}^T |r_n(t) t^{-1} S_n(t)| dt,$$

$$I_2 = c_n \int_{-T}^T \left(\left| \frac{h_n(t) - h(t)}{t} \right| e^{-t^2/2} + |h_n(t) t^{-1} (S_n(t) - e^{-t^2/2})| \right) dt,$$

$$I_3 = \frac{1}{2} \rho_n \int_{-T}^T |t(S_n(t) - e^{-t^2/2})| dt.$$

Since

$$a_n c_n = \frac{n}{a_n^2} \int_{-a_n}^{a_n} y^2 F(dy) \cdot b_n \sim b_n,$$

where

$$b_n = \frac{\int_{-a_n}^{a_n} |y|^3 F(dy)}{\int_{-a_n}^{a_n} y^2 F(dy)},$$

the assumption of an infinite absolute third moment implies $T = o(a_n)$. Thus according to Lemma 3 there is a positive constant c such that

$$|f^n(t/a_n)| \leq e^{-c|t|} \quad \text{for } 1 < |t| < T$$

and all sufficiently large n . Hence also

$$\left| f^j(t/a_n) \exp \left(-\frac{1}{2} t^2 \frac{n-1-j}{n} \right) \right| < e^{-c_0|t|}, \quad 0 \leq j < n \tag{21}$$

for some $c_0 > 0$ if $1 < |t| < T$ and n is large enough (because (21) is trivial when $j < n/2$). Because of (21), for the above-mentioned t and n ,

$$|S_n(t)| < e^{-c_0|t|}. \tag{22}$$

From (19) and (22) we conclude $I_1 = O(1/n)$. The inequalities

$$|e^{itx} - 1 - itx - \frac{1}{2}(it\tau(x))^2| \leq \begin{cases} \frac{1}{6}|tx|^3 & \text{if } |x| \leq 1, \\ 2|tx| & \text{if } |x| > 1, \end{cases}$$

give

$$|h_n(t)| \leq \frac{1}{6}|t|^3 \int_{|x| \leq 1} |U_n(dx)| + 2|t| \int_{|x| > 1} \frac{1}{x^2} |U_n(dx)|.$$

This together with (15) and (16) shows that

$$|h_n(t)| \leq (\text{constant not depending on } n) |t|(1+t^2). \tag{23}$$

The functions under the integral signs in I_2 and I_3 tend to zero and they are according to (22) and (23) dominated by integrable functions. Hence, by Lebesgues dominated convergence theorem, $I_2 = o(c_n)$, $I_3 = o(\rho_n)$.

Thus we have shown that

$$F_n(x) = (1 + c_n \tilde{\omega}_\delta + \frac{1}{2} \rho_n D^2) \Phi(x) + o(c_n) + o(\rho_n) \tag{24}$$

uniformly in x .

If $\delta > 0$ and $a_n = \sigma \sqrt{n}$ it follows from (10) that

$$\rho_n = -c_n \int_{|x| > 1} \frac{1}{x} \Omega_\delta(dx) + o(c_n).$$

Hence we can get rid of the truncation in this case:

$$\begin{aligned} & (c_n \tilde{\omega}_\delta + \rho_n \frac{1}{2} D^2) \Phi(x) \\ &= c_n \int \frac{\Phi(x-y) - \Phi(x) + y D \Phi(x) - \frac{1}{2} y^2 D^2 \Phi(x)}{y^3} \Omega_\delta(dy) + o(c_n), \end{aligned} \tag{25}$$

uniformly in x . Three partial integrations to the right in (25) show that (24) is equivalent to (6) if $\delta > 0$.

It remains to show that a choice of the sequence a_1, a_2, \dots according to (7) is possible. Put $V(x) = \int_{|y| \leq x} y^2 F(dy)$ and

$$a_n = \sup \left\{ x > 0 \mid \frac{n}{x^2} V(x) \geq 1 \right\}$$

for n large enough. Then, since V is nondecreasing, $n a_n^{-2} V(a_n) = 1$.

4. Proof of Theorem 2 and 3

From the inversion formula for Fourier transforms we obtain in the absolutely continuous case

$$v_n(x) - (1 + c_n \tilde{\omega}_\delta + \rho_n \frac{1}{2} D^2) \varphi(x) = \frac{1}{2\pi} \int e^{-itx} (f^n(t/a_n) - g_n(t)) dt \tag{26}$$

where g_n is defined in (20). We choose $\varepsilon > 0$ so small that (12) is true for $1 < |t| < \varepsilon a_n$ and n sufficiently large and divide the integral to the right in (26) into two parts,

$$\int_{|t| < \varepsilon a_n} + \int_{|t| \geq \varepsilon a_n}.$$

We proceed with the first integral in a similar way as in the proof of Theorem 1 of this paper; with the second as in [4], p. 489.

The modifications necessary when F is a lattice distribution parallels those which permit us to conclude Theorem 3, p. 490 of [4], from Theorem 2, p. 489 of [4].

Acknowledgement. I wish to thank B. von Bahr for advice and criticism.

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(Received April 18, 1969)