

# A THEOREM ON CONVERGENCE TO A LÉVY PROCESS

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## 1. Introduction.

Let  $X_{n,1}, X_{n,2}, \dots, X_{n,k_n}$ ,  $n = 1, 2, \dots$ , be a double sequence of random variables such that

(1)  $X_{n,1}, X_{n,2}, \dots, X_{n,k_n}$  are independent for every  $n$

and

(2)  $\lim_{n \rightarrow \infty} \max_j P(|X_{n,j}| > \varepsilon) = 0$  for every  $\varepsilon > 0$ .

For the study of the asymptotic behaviour of the partial sums:

(3) 
$$S_{n,m} = \sum_{j=1}^m X_{n,j}, \quad m = 1, 2, \dots, k_n, \quad n = 1, 2, \dots,$$

it is natural to consider a sequence of random broken lines:

(4) 
$$\begin{aligned} Y_{n,\omega}(t) &= 0 && \text{if } 0 = t_{n,0} \leq t < t_{n,1}, \\ &= S_{n,m} - \gamma_{n,m} && \text{if } t_{n,m} \leq t < t_{n,m+1}, \quad 0 < m < k_n, \\ &= S_{n,k_n} && \text{if } t = t_{n,k_n} = 1, \end{aligned}$$

where  $0 = t_{n,0} < t_{n,1} < \dots < t_{n,k_n} = 1$  is a set of division time-points for each  $n = 1, 2, \dots$  and  $\gamma_{n,m}$  are adjusting constants.  $Y_n(t)$  is a random process whose sample functions are in the space  $D[0, 1]$  of all functions on  $[0, 1]$  with no discontinuities of the second kind. The space  $D[0, 1]$  with the Skorohod topology is a Polish space, i.e., a topological space homeomorphic with a complete separable metric space. Let  $P_n$  be the probability law governing the sample function of  $Y_n$  and hence a regular probability measure on  $D[0, 1]$  for every  $n$ . Thus the study of partial sums  $S_{n,m}$  is reduced to the problem of convergence of the sequence  $\{P_n\}$ .

Suppose that  $\{P_n\}$  is weakly convergent. Then  $\{S_{n,k_n}, n = 1, 2, \dots\}$  is necessarily convergent in law, because  $S_{n,k_n} = Y_n(1)$ . Suppose conversely that  $\{S_{n,k_n}\}_n$  is convergent in law. Then we can find  $\gamma_{n,m}$  and  $t_{n,m}$  such that  $\{P_n\}$  is conditionally weakly compact. This is the main result of our present paper. Prohorov [1] discussed similar problems for the following special cases:

- (a)  $X_{n,1}, X_{n,2}, \dots, X_{n,k_n}$  are identically distributed, where we can take  $\gamma_{n,m} = 0$  and  $t_{n,m} = m/k_n$ .
- (b)  $X_{n,1}, X_{n,2}, \dots, X_{n,k_n}$  satisfy the Lindeberg conditions, where we can take  $\gamma_{n,m} = E(S_{n,m})$  and  $t_{n,m} = V(S_{n,m})/V(S_{n,k_n})$ .

In the general case we are going to discuss in this paper we will use the central value and the dispersion to determine  $\gamma_{n,m}$  and  $t_{n,m}$ .

In Section 2 we will first review the definitions and the basic properties of the central value and the dispersion following Ito [2] and then prove a few new facts which will be used in Section 3. In Section 3 we will prove our main theorem on convergence to a Lévy process and use it to show the equivalence of several definitions of the infinitely divisible laws.

**2. Definitions and elementary facts about dispersion and central value.**

Following [2] or [3] we adopt the following definition.

**DEFINITION.** The central value  $\gamma = \gamma(\mu)$  of a one-dimensional probability measure  $\mu$  is defined to be the unique real number  $\gamma$  such that

$$\int_{\mathbb{R}^1} \arctan(x - \gamma) \mu(dx) = 0 .$$

The dispersion  $\delta(\mu)$  is defined by

$$\delta(\mu) = -\log \int \int_{\mathbb{R}^2} e^{-|x-y|} \mu(dx) \mu(dy) .$$

For a realvalued random variable  $X$  with probability law  $\mu$  we have the natural definitions  $\gamma(X) = \gamma(\mu)$  and  $\delta(X) = \delta(\mu)$ .

A measure  $\mu_1$  is called a factor of  $\mu$  if  $\mu = \mu_1 * \mu_2$ .

**PROPOSITION 1.** *Let  $\varphi_X$  be the characteristic function of  $X$ . Then*

$$\delta(X) = -\log \pi^{-1} \int_{\mathbb{R}^1} |\varphi_X(\xi)|^2 \frac{d\xi}{1 + \xi^2} .$$

**PROPOSITION 2.**  $\mu = \delta_c$  (the  $\delta$ -distribution concentrated at  $c$ ) if and only if  $\delta(\mu) = 0$  and  $\gamma(\mu) = c$ .

**PROPOSITION 3.** *If  $X \rightarrow Y$  i.p., then  $\delta(X) \rightarrow \delta(Y)$  and  $\gamma(X) \rightarrow \gamma(Y)$ .*

PROPOSITION 4.  $\delta(X) \rightarrow 0$  if and only if  $X - \gamma(X) \rightarrow 0$  i.p.

PROPOSITION 5. If  $X$  and  $Y$  are independent, then  $\delta(X) \leq \delta(X + Y)$  with equality if and only if  $Y = \text{const. a.s.}$

PROPOSITION 6. Let  $\mathcal{M}$  be a set of one-dimensional probability measures and  $\mathcal{M}'$  the set of all factors of probability measures in  $\mathcal{M}$ . Then  $\{\mu' - \gamma(\mu')\}_{\mu' \in \mathcal{M}'}$  is conditionally compact if  $\mathcal{M}$  is conditionally compact.

**2.2. Further properties of the central value and dispersion.**

For the proof of our main result we need some more facts about central value and dispersion.

LEMMA 1. Given  $\varepsilon > 0$  and  $c > 0$ , we can find  $d = d(\varepsilon, c)$  such that if  $\delta(X + Y) - \delta(X) < d$  for some random variable  $X$  independent of  $Y$  with  $\delta(X) \leq c$ , then  $\delta(Y) < \varepsilon$ .

PROOF. First we note the following fact: If  $[-a, a]$  is a compact interval and  $B$  a Borel set contained in  $[-a, a]$  such that  $|B| \geq c > 0$ , then for some constant  $K = K(c, a) > 0$  and for all  $x$  and all  $B$  of the above type,

$$(5) \quad \int_B (1 - \cos \xi x) d\xi \geq K x^2 / (1 + x^2).$$

Now let  $X$  and  $Y$  be independent random variables with characteristic functions  $\varphi_X$  and  $\varphi_Y$ , respectively. Then

$$e^{\delta(X+Y) - \delta(X)} = I_1 / (I_1 - I_2),$$

where

$$I_1 = \int_{-\infty}^{\infty} |\varphi_X(\xi)|^2 \frac{d\xi}{1 + \xi^2}, \quad I_2 = \int_{-\infty}^{\infty} |\varphi_X(\xi)|^2 (1 - |\varphi_Y(\xi)|^2) \frac{d\xi}{1 + \xi^2}.$$

If  $\delta(X) \leq c$  we get  $\pi \geq I_1 \geq \pi e^{-c} = K_1(c) > 0$ , and

$$1 - |\varphi_Y(\xi)|^2 = \int_{-\infty}^{\infty} (1 - \cos \xi x) \mu(dx)$$

for some probability measure  $\mu$ . By interchanging the order of integration,

$$I_2 = \int_{-\infty}^{\infty} \mu(dx) \int_{-\infty}^{\infty} |\varphi_X(\xi)|^2 \frac{1 - \cos \xi x}{1 + \xi^2} d\xi.$$

Now choose  $K_2 = K_2(c)$  such that  $\int_{K_2}^{\infty} d\xi/(1+\xi^2) = \frac{1}{4}K_1$ . Then

$$\int_{-K_2}^{K_2} |\varphi_X(\xi)|^2 d\xi/(1+\xi^2) \geq \frac{1}{2}K_1.$$

Let  $E = \{\xi; |\xi| \leq K_2, |\varphi_X(\xi)|^2 \geq \frac{1}{8}K_1/K_2\}$ . Then we have

$$\int_{[-K_2, K_2] \setminus E} |\varphi_X(\xi)|^2 d\xi/(1+\xi^2) \leq 2K_2 \frac{1}{8}K_1/K_2 = \frac{1}{4}K_1$$

and therefore

$$\int_E |\varphi_X(\xi)|^2 d\xi/(1+\xi^2) \geq \frac{1}{4}K_1, \quad |E| \geq \frac{1}{4}K_1.$$

By (5) there exists a constant  $K_3 = K_3(c)$  such that

$$\begin{aligned} I_2 &\geq \int_{-\infty}^{\infty} \mu(dx) \int_E |\varphi_X(\xi)|^2 \frac{1 - \cos \xi x}{1 + \xi^2} d\xi \\ &\geq \int_{-\infty}^{\infty} \mu(dx) \frac{K_1}{8K_2} \frac{1}{1 + K_2^2} \int_E (1 - \cos \xi x) d\xi \geq K_3 \int_{-\infty}^{\infty} \frac{x^2}{1 + x^2} \mu(dx). \end{aligned}$$

It follows that

$$\int_{-\infty}^{\infty} \frac{x^2}{1 + x^2} \mu(dx) \leq \frac{I_2}{K_3} = \frac{I_1(e^{\delta(X+Y)} - 1)}{K_3 e^{\delta(X+Y)} - 1} \leq \frac{\pi}{K_3} (e^{\delta(X+Y)} - 1).$$

Observing that

$$\int_{-\infty}^{\infty} \frac{x^2}{1 + x^2} \mu(dx) \rightarrow 0 \Rightarrow \delta(Y) \rightarrow 0,$$

the lemma is proved.

**LEMMA 2.** *Given  $\varepsilon > 0$  there exists a  $d = d(\varepsilon) > 0$  such that, for all  $k$  and  $X_1, X_2, \dots, X_k$  independent,*

$$|\gamma(X_1 + \dots + X_k) - \sum_{j=1}^k \gamma(X_j)| < \varepsilon$$

as soon as  $\delta(X_1 + \dots + X_k) < d$ .

PROOF. Assume that the lemma is false. Then for some  $\varepsilon > 0$  we can find a double sequence  $X_{n,1}, X_{n,2}, \dots, X_{n,k_n}, n = 1, 2, \dots$  of random variables such that:

$$\begin{aligned} & X_{n,1}, X_{n,2}, \dots, X_{n,k_n} \text{ are independent for each } n, \\ & \delta(X_{n,1} + X_{n,2} + \dots + X_{n,k_n}) \rightarrow 0 \text{ as } n \rightarrow \infty, \\ & \gamma(X_{n,1}) = \gamma(X_{n,2}) = \dots = \gamma(X_{n,k_n}) = 0, \quad n = 1, 2, 3, \dots, \\ & |\gamma(X_{n,1} + X_{n,2} + \dots + X_{n,k_n})| \geq \varepsilon, \quad n = 1, 2, 3, \dots \end{aligned}$$

It is easy to see that we can assume that  $\gamma(X_{n,1} + \dots + X_{n,k_n}) \rightarrow \varepsilon$  as  $n \rightarrow \infty$ .

Let  $\varphi_{n,j}$  denote the characteristic function of  $X_{n,j}$ . Then we have, by Proposition 4,

$$(6) \quad \varphi_{n,1}(\xi) \varphi_{n,2}(\xi) \dots \varphi_{n,k_n}(\xi) \rightarrow e^{i\xi\varepsilon}.$$

Letting  $\xi_0 = \pi/\varepsilon$  in (6) we get

$$\begin{aligned} & \left[ \int_{-\infty}^{\infty} \cos \xi_0 x \mu_{n,1}(dx) + i \int_{-\infty}^{\infty} \sin \xi_0 x \mu_{n,1}(dx) \right] \dots \\ & \left[ \int_{-\infty}^{\infty} \cos \xi_0 x \mu_{n,k_n}(dx) + i \int_{-\infty}^{\infty} \sin \xi_0 x \mu_{n,k_n}(dx) \right] \rightarrow -1. \end{aligned}$$

Since  $\varphi_{n,j}(\xi) \rightrightarrows 1$  uniformly in  $j$  as  $n \rightarrow \infty$ , we have

$$\left| \sum'_j \int_{-\infty}^{\infty} \sin \xi_0 x \mu_{n,j}(dx) \right| > 1$$

for all  $n$  large enough, where  $\sum'$  denotes the sum over all positive or all negative terms. Further,

$$\sin \xi_0 x = \frac{\xi_0 x + H_1(x)x^2}{1+x^2}, \quad \arctan x = \frac{x + H_2(x)x^2}{1+x^2},$$

where  $H_1(x)$  and  $H_2(x)$  are bounded on the real line. By assumption

$$\sum' \int_{-\infty}^{\infty} \xi_0 \arctan x \mu_{n,j}(dx) = 0.$$

Therefore

$$\left| \sum' \int_{-\infty}^{\infty} \frac{[H_1(x) - \xi_0 H_2(x)]x^2}{1+x^2} \mu_{n,j}(dx) \right| > 1,$$

and so we can find a constant  $c > 0$  such that

$$(7) \quad \sum_{j=1}^{k_n} \int_{-\infty}^{\infty} \frac{x^2}{1+x^2} \mu_{n,j}(dx) > c > 0 \quad \text{for all } n .$$

On the other hand it follows from (6) that

$$|\varphi_{n,1}(\xi)|^2 |\varphi_{n,2}(\xi)|^2 \dots |\varphi_{n,k_n}(\xi)|^2 \rightrightarrows 1 \quad \text{as } n \rightarrow \infty .$$

Therefore

$$\sum_{j=1}^{k_n} (1 - |\varphi_{n,j}(\xi)|^2) = \sum_{j=1}^{k_n} \int_{-\infty}^{\infty} (1 - \cos \xi x) \tilde{\mu}_{n,j}(dx) \rightrightarrows 0 \quad \text{as } n \rightarrow \infty$$

( $\tilde{\mu}_{n,j}$  denotes the symmetrization of  $\mu_{n,j}$ ), and from this

$$\int_{-1}^1 \sum_{j=1}^{k_n} (1 - |\varphi_{n,j}(\xi)|^2) d\xi \rightarrow 0 ,$$

which implies

$$(8) \quad \sum_{j=1}^{k_n} \int_{-\infty}^{\infty} \frac{x^2}{1+x^2} \tilde{\mu}_{n,j}(dx) \rightarrow 0 \quad \text{as } n \rightarrow \infty .$$

Comparing (7) and (8) we can see that Lemma 2 is proved, when we have verified:

**LEMMA 3.** *If  $\mu$  is a one-dimensional probability measure with  $\gamma(\mu) = 0$  and if  $\tilde{\mu}$  denotes the symmetrization of  $\mu$ , we have*

$$\int_{-\infty}^{\infty} \frac{x^2}{1+x^2} \tilde{\mu}(dx) \geq \frac{1}{16} \int_{-\infty}^{\infty} \frac{x^2}{1+x^2} \mu(dx) .$$

**PROOF.** We can assume that  $m \geq 0$ , where  $m$  is a median of  $\mu$ , that is,  $\mu(-\infty, m] \geq \frac{1}{2}$  and  $\mu[m, \infty) \geq \frac{1}{2}$ . First we note that

$$\begin{aligned} I &= \int_{-\infty}^{\infty} \frac{x^2}{1+x^2} \tilde{\mu}(dx) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{(x-y)^2}{1+(x-y)^2} \mu(dx) \mu(dy) \\ &\geq \int_{x>0} \int_{y \leq 0} \frac{x^2}{1+x^2} \mu(dx) \mu(dy) + \int_{x \leq 0} \int_{y \geq 0} \frac{x^2}{1+x^2} \mu(dx) \mu(dy) \\ &\geq \mu(-\infty, 0] \int_{x>0} \frac{x^2}{1+x^2} \mu(dx) + \mu[0, \infty) \int_{x \leq 0} \frac{x^2}{1+x^2} \mu(dx) , \end{aligned}$$

and from this also

$$I \geq \frac{1}{2} \int_{-\infty}^{\infty} \frac{(x-m)^2}{1+(x-m)^2} \mu(dx).$$

If  $\mu(-\infty, 0] \geq \frac{1}{16}$ , it follows immediately that

$$I \geq \frac{1}{16} \int_{-\infty}^{\infty} \frac{x^2}{1+x^2} \mu(dx),$$

so it is enough to consider the case  $\mu(-\infty, 0] \leq \frac{1}{16}$ . Then

$$\int_{x \leq 0} \frac{x^2}{1+x^2} \mu(dx) \leq \frac{16}{15} I$$

and

$$\int_{2m}^{\infty} \frac{x^2}{1+x^2} \mu(dx) \leq 4 \int_{2m}^{\infty} \frac{(x-m)^2}{1+(x-m)^2} \mu(dx) \leq 8I.$$

Since

$$\begin{aligned} \frac{1}{2}\pi \frac{1}{16} &\geq \int_{x < 0} \arctan(-x) \mu(dx) = \int_{x > 0} \arctan x \mu(dx) \\ &\geq \frac{1}{2} \arctan m \geq \frac{1}{4} \arctan 2m, \end{aligned}$$

we first get  $m \leq \frac{1}{4}$  and then we have

$$\begin{aligned} \int_{x \leq 0} \frac{x^2}{1+x^2} \mu(dx) &\geq \int_{x \leq -2m} \frac{x^2}{1+x^2} \mu(dx) \\ &\geq \int_{x \leq -2m} m \arctan(-x) \mu(dx) \\ &\geq \frac{3}{4} \int_{x < 0} m \arctan(-x) \mu(dx) \\ &= \frac{3}{4} \int_{x > 0} m \arctan x \mu(dx) \\ &\geq \frac{3m}{4} \int_0^{2m} \arctan x \mu(dx) \\ &\geq \frac{3m}{4} \int_0^{2m} \frac{1}{4m} \frac{x^2}{1+x^2} \mu(dx) = \frac{3}{16} \int_0^{2m} \frac{x^2}{1+x^2} \mu(dx). \end{aligned}$$

Therefore

$$\int_0^{2m} \frac{x^2}{1+x^2} \mu(dx) \leq \frac{16}{3} \int_{x \leq 0} \frac{x^2}{1+x^2} \mu(dx) \leq 6I,$$

and so we have

$$\int_{-\infty}^{\infty} \frac{x^2}{1+x^2} \mu(dx) = \int_{x \leq 0} + \int_0^{2m} + \int_{x \geq 2m} \frac{x^2}{1+x^2} \mu(dx) \leq 16I.$$

LEMMA 4. Given  $\varepsilon > 0$  and  $c > 0$  we can find a  $d = d(\varepsilon, c) > 0$  such that  $\delta(X + Y) - \delta(X) < \varepsilon$  for all pairs  $(X, Y)$  of independent random variables satisfying  $\delta(Y) < d$  and  $\delta(X) < c$ .

PROOF. Use Proposition 1.

### 3.1. A theorem on convergence to a Lévy process.

Let  $X_{n,1}, X_{n,2}, \dots, X_{n,k_n}$ ,  $n = 1, 2, \dots$ , be a double sequence of random variables subject to conditions (1) and (2). Assume further that  $X_{n,j}$  has distribution  $\mu_{n,j}$  and that

$$(9) \quad \mu_{n,1} * \mu_{n,2} * \dots * \mu_{n,k_n} \xrightarrow{w} \mu,$$

where  $\delta(\mu) > 0$ . Define for fixed  $\omega \in (\Omega, \mathcal{B}, P)$  and  $n \in N$  a broken line

$$Y_{n,\omega}(t) = \begin{cases} 0 & \text{if } 0 \leq t < t_{n,1} \\ S_{n,m} - \sum_{j=1}^m \gamma(X_{n,j}) + t_{n,m} \sum_{j=1}^{k_n} \gamma(X_{n,j}) & \text{if } t_{n,m} \leq t < t_{n,m+1} \\ S_{n,k_n} & \text{if } t = 1. \end{cases}$$

We take

$$t_{n,m} = \frac{\delta(\mu_{n,1} * \dots * \mu_{n,m})}{\delta(\mu_{n,1} * \dots * \mu_{n,k_n})},$$

$n = 1, 2, \dots$ . Then  $Y_{n,\omega}(t)$  defines a distribution  $P_n$  in the function space  $D[0, 1]$ , which is assumed to have the Skorohod topology ([4, p.109]). Without restriction we can assume that  $D[0, 1]$  is a complete separable metric space ([4, p.113]). We are going to prove

THEOREM 1. The sequence  $\{P_n\}_n$  of distributions in  $D[0, 1]$  defined above is weakly conditionally compact and every limit distribution  $P$  corresponds stochastically to a process  $X_t(\omega)$ ,  $t \in [0, 1]$ , continuous in probability and with independent increments (a Lévy process). Moreover,  $X_1(\omega)$  has probability law  $\mu$ .

### 3.2. Prohorov's lemma.

We now consider an arbitrary double sequence  $\tilde{X}_{n,j}$ ,  $j = 1, 2, \dots, k_n$ ,  $n = 1, 2, \dots$ , of random variables, and a family of division time-points



$t_{n,j}, j = 1, 2, \dots, k_n, n = 1, 2, \dots$ . We construct the random broken lines with vertices  $(t_{n,m}, \sum_{j=1}^m \tilde{X}_{n,j})$ . In Prohorov [1, p.193] the following lemma is proved.

LEMMA 5. *The sequence  $\{\tilde{P}_n\}_n$  of distributions in  $D[0,1]$  defined by the above-mentioned random broken lines is conditionally compact and every limit distribution for  $\{\tilde{P}_n\}_n$  corresponds stochastically to a continuous process with independent increments if*

- (i)  $\theta_n = \max_j (t_{n,j} - t_{n,j-1}) \rightarrow 0$  as  $n \rightarrow \infty$ ,
- (ii)  $\max_{|d| \leq d} P\{|\sum_{t_{n,j} \in \Delta} \tilde{X}_{n,j}| > \lambda\} \rightarrow 0$  as  $d \rightarrow 0$ ,

*uniformly in  $n$  for every fixed  $\lambda > 0$ ,*

- (iii)  $\max_{\Delta} P\{|\sum_{t_{n,j} \in \Delta} \tilde{X}_{n,j}| > \lambda\} \rightarrow 0$

*uniformly in  $n$  as  $\lambda \rightarrow \infty$ . The maximum in (ii) and (iii) is taken over all  $\Delta$  of form  $(t_{n,j}, t] \subset [0, 1]$ .*

REMARK. The proof in [1] is not completely correct because Lemma 2.4 on p. 182 is false. But using a similar theorem given by Billingsley in [4, p. 125], we can easily correct Prohorov's proof, and proceeding in this way we automatically get a proof in the case when  $D[0, 1]$  has the Skorohod topology. (The topology used by Prohorov is in fact equivalent to the Skorohod topology.)

### 3.3. Proof of Theorem 1.

We now return to the theorem on convergence to a Lévy process stated in Section 3.1. First we prove that the Prohorov broken line  $\tilde{Y}_{n,\omega}(t)$  determined by the double sequence  $\tilde{X}_{n,j}, j = 1, 2, \dots, k_n, n = 1, 2, \dots$ , defined by

$$\tilde{X}_{n,j} = X_{n,j} - \gamma(X_{n,j}),$$

and the time-points  $t_{n,m} = \delta(\mu_{n,1} * \dots * \mu_{n,m}) / \delta(\mu_{n,1} * \dots * \mu_{n,k_n})$  satisfies the conditions in Lemma 5.

(i) Since  $\delta(\mu_{n,1} * \dots * \mu_{n,k_n}) \rightarrow \delta(\mu) > 0$  as  $n \rightarrow \infty$ , condition (i) follows immediately from Lemma 4 and the fact that  $\delta(\mu_{n,j}) \rightarrow 0$  uniformly in  $j$  as  $n \rightarrow \infty$ .

(ii) Using Lemma 1 we get

$$\sup_{|d| \leq d} \delta(\sum_{t_{n,j} \in \Delta} \tilde{X}_{n,j}) \rightarrow 0 \text{ uniformly in } n \text{ as } d \rightarrow \infty.$$

From Lemma 2 we have

$$\sup_{|d| \leq d} |\gamma(\sum_{t_{n,j} \in \Delta} \tilde{X}_{n,j})| \rightarrow 0 \text{ uniformly in } n \text{ as } d \rightarrow 0,$$

and so by Proposition 4,

$$\sup_{|\Delta| \leq d} P\{|\sum_{t_{n,j} \in \Delta} \tilde{X}_{n,j}| > \lambda\} \rightarrow 0 \quad \text{uniformly in } n \text{ as } d \rightarrow 0.$$

(iii) We can choose  $n_0$  so large that

$$\left| \sum_{t_{n,j} \in \Delta} \gamma(X_{n,j}) - \gamma(\sum_{t_{n,j} \in \Delta} X_{n,j}) \right| \leq 1$$

for all  $\Delta$  with  $|\Delta| \leq n_0^{-1}$ . Since  $\mathcal{M} = \{\mu_{n,1} * \dots * \mu_{n,k_n}\}_n$  is conditionally compact, the family

$$\mathcal{M}' = \left\{ \mu_{n,m} * \mu_{n,m+1} * \dots * \mu_{n,p} - \sum_{j=m}^p \gamma(\mu_{n,j}) : 1 \leq m \leq p \leq k_n, \right. \\ \left. t_{n,m}, t_{n,m+1}, \dots, t_{n,p} \in \Delta, |\Delta| \leq n_0^{-1} \right\}$$

is also conditionally compact by Proposition 6. Since  $D[0,1]$  is a complete separable metric space,  $\mathcal{M}'$  is tight. Therefore, given  $\varepsilon > 0$ , we can choose  $\lambda$  so large that

$$P\{|\sum_{t_{n,j} \in \Delta} \tilde{X}_{n,j}| > \frac{1}{2}\lambda n_0^{-1}\} \leq \frac{1}{2}\varepsilon n_0^{-1} \quad \text{for all } \Delta, |\Delta| \leq n_0^{-1}.$$

If now  $\Delta$  is an arbitrary subinterval of  $[0,1]$  of form  $(t_{n,j}, t]$ , we can for all large enough  $n$  divide it into at most  $2n_0$  intervals of form  $(t_{n,k}, t]$ , each one of length at most  $n_0^{-1}$ . Therefore taking  $\lambda$  large enough,

$$P\{|\sum_{t_{n,j} \in \Delta} \tilde{X}_{n,j}| > \lambda\} < \varepsilon \quad \text{for all } \Delta \text{ and all } n,$$

which proves (iii).

Lemma 5 can now be applied. We see that  $\{\tilde{P}_n\}_n$  is conditionally compact and every limit distribution  $\tilde{P}$  corresponds stochastically to a Lévy process  $X_t(\omega), t \in [0,1]$ . Let  $\{P_q\}$  be a subsequence converging weakly to  $\tilde{P}$ . If  $\pi_1$  is the projection, which takes  $Y(\cdot) \in D[0,1]$  into  $Y(1)$ , then  $\pi_1$  is continuous and therefore,

$$\tilde{P}_q \xrightarrow{w} \tilde{P} \Rightarrow \tilde{P}_q \pi_1^{-1} \rightarrow \tilde{P} \pi_1^{-1}.$$

But  $\tilde{P}_q \pi_1^{-1}$  is the probability law of  $\sum_{j=1}^{k_q} \tilde{X}_{q,j}$ , that is,

$$\tilde{P}_q \pi_1^{-1} = \mu_{q,1} * \mu_{q,2} * \dots * \mu_{q,k_q} - \sum_{j=1}^{k_q} \gamma(\mu_{q,j}),$$

and  $\tilde{P} \pi_1^{-1}$  is the probability law of  $\tilde{X}_1(\omega)$ . Since

$$\mu_{q,1} * \mu_{q,2} * \dots * \mu_{q,k_q} \xrightarrow{w} \mu \quad \text{as } q \rightarrow \infty,$$

there exists a constant  $c$  such that

$$(10) \quad \sum_{j=1}^{k_q} \gamma(\mu_{q,j}) \rightarrow c \quad \text{as } q \rightarrow \infty,$$

and  $\tilde{X}_1(\omega)$  has the probability law  $\mu - c$ .

From (10) it follows that  $\{\sum_{j=1}^{k_n} \gamma(X_{n,j})\}_n$  is bounded. Therefore we

can easily see that Lemma 5 can also be applied to the random broken lines  $Y_{n,\omega}(t)$  in our theorem. In this case also every limit process  $X_t(\omega)$ ,  $t \in [0, 1]$ , is such that  $X_1(\omega)$  has the probability law  $\mu$ .

REMARK 1. The sequence  $\{P_n\}_n$  in the theorem is not convergent in general, as is easily seen.

REMARK 2. Since Lévy's decomposition of the sample path of a Lévy process can be proved directly [3, Section 1.7], the theorem in this note can be used to prove the equivalence of the following characterizations of an infinitely divisible distribution.

(i) There exists a family  $\{\mu_n\}_n$  of distributions such that  $\mu = \mu_n^{n*}$  for all  $n$ .

(ii) There exists a family  $\{\mu_{n,k}\}_{n,k}$  of distributions such that

$$\mu_{n,1} * \mu_{n,2} * \dots * \mu_{n,k_n} \xrightarrow{w} \mu \quad \text{and} \quad \mu_{n,j}[-\varepsilon, +\varepsilon] \rightarrow 0$$

uniformly in  $j$  as  $n \rightarrow \infty$  for every  $\varepsilon > 0$ .

(iii) There exists a Lévy process  $X_t(\omega)$ ,  $t \in [0, 1]$ , such that  $\mu$  is the probability law of  $X_1(\omega)$ .

(iv) The characteristic function  $\varphi_\mu(z)$  of  $\mu$  can be written in the form

$$\varphi_\mu(z) = \exp\{imz - \frac{1}{2}vz^2 + \int_{-\infty}^{\infty} (e^{iuz} - 1 - iuz)/(1+u^2)n(du)\}$$

where  $v \geq 0$ ,  $n(du) \geq 0$ , and  $\int_{-\infty}^{\infty} (u^2/(1+u^2)) n(du) < \infty$ .

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