

## CENTRAL LIMIT THEOREM IN $D[0, \infty)$ FOR BREAKDOWN PROCESSES

BY

WŁADYSŁAW SZCZOTKA (WROCLAW)

*Abstract.* Main results are given in Theorems 1-3. Theorem 1 asserts that a mixture of independent random elements of  $(D, d)$  satisfying the Central Limit Theorem in  $(D, d)$  also satisfies the Central Limit Theorem in  $(D, d)$ . Theorem 2 determines an upper bound for  $P\{X(t_1) = X(t_2) \neq X(t)\}$ , where  $t_1 \leq t \leq t_2$ ,  $t_2 - t_1$  is small and  $X$  is a breakdown process. Theorem 3 gives sufficient conditions under which a breakdown process satisfies the Central Limit Theorem in  $(D, d)$ .

**1. Introduction.** A stochastic process  $X$  assuming values 0 and 1 at any time is called the *binary process*. The class of binary processes plays an important role in reliability theory. The assumption that the value 0 is taken on if a component of a system is functioning and the value 1 if the component is failed allows us to use the class of binary processes for describing the behaviour of a system. The first moment at which a system fails is an important characteristic. If the state of a system is described by the number of failed components, i.e. by a sum of binary processes, then such a moment can be defined as the first passage time into some set. Thus an investigation of the asymptotic behaviour of sums of binary processes seems to be important. A subclass of the class of binary processes, called *breakdown processes*, is considered in Section 3. For those processes we give sufficient conditions under which the normalized sums of independent and identically distributed breakdown processes converge in distribution to a Gaussian process.

**2. Central Limit Theorem for mixture of independent random elements of  $(D, d)$ .** Denote by  $D$  the space of all real-valued right continuous functions on  $[0, \infty)$  which have left-hand limits in  $(0, \infty)$ . We consider  $D$  with metric

$d$  defined in [5] where it was shown that  $(D, d)$  is a complete separable metric space. Let  $\tilde{D}$  denote the space of  $D$ -valued right continuous functions on  $[0, \infty)$  which have left-hand limits in  $(0, \infty)$ . The space  $\tilde{D}$  with metric  $\tilde{d}$  defined in [6] is a complete separable metric space. By  $D_0$  we denote the set of those elements of  $D$  which are non-negative and non-decreasing. Let  $(\tilde{D}^m, \tilde{d}^m)$  and  $(D_0^m, d^m)$  denote the Cartesian product of  $m$  copies of the spaces  $(\tilde{D}, \tilde{d})$  and  $(D_0, d)$ , respectively. Define the mapping  $\tau$  of  $\tilde{D} \times D_0$  in  $\tilde{D}$  by  $\tau(x, v) = x \circ v$ , where  $(x \circ v)(t) = x(v(t))$  for  $x \in \tilde{D}$ ,  $v \in D_0$ ,  $t \geq 0$ . In [6] it has been shown that  $\tau$  is a measurable and continuous mapping on  $\tilde{C} \times C_0$ , where  $\tilde{C}$  and  $C_0$  are subsets of the sets of continuous functions belonging to  $\tilde{D}$  and  $D_0$ , respectively. By  $\tau$  we denote the mapping of  $\tilde{D}^m \times D_0^m$  in  $\tilde{D}^m$  defined by

$$\tau(x, v) = (\tau(x_1, v_1), \tau(x_2, v_2), \dots, \tau(x_m, v_m)),$$

where  $x = (x_1, x_2, \dots, x_m) \in \tilde{D}^m$  and  $v = (v_1, v_2, \dots, v_m) \in D_0^m$ . Hence  $\tau$  is continuous on  $\tilde{C}^m \times C_0^m$ , where  $\tilde{C}^m$  and  $C_0^m$  denote the Cartesian product of  $m$  copies of  $\tilde{C}$  and  $C_0$ , respectively.

A random element  $Y$  of  $(D, d)$  is said to satisfy the Central Limit Theorem (CLT) in  $(D, d)$  if there exists a Gaussian process  $Z$  with sample paths in  $D$  which is the limit in distribution in  $(D, d)$  of the sequence  $\{\zeta_n\}$ , where

$$\zeta_n(t) = \frac{1}{\sqrt{n}} (Y_1(t) + Y_2(t) + \dots + Y_n(t) - EY_1(t) - EY_2(t) - \dots - EY_n(t)),$$

and  $Y_1, Y_2, \dots$  are independent copies of  $Y$  on the same probability space.

Let  $\delta = (\delta_1, \delta_2, \dots, \delta_m)$  be the random vector of  $R^m$  the components  $\delta_i$  of which take on the values 0 or 1 and  $\delta_1 + \delta_2 + \dots + \delta_m = 1$ . Put  $p_i = P\{\delta_i = 1\}$  for  $0 < p_i < 1$ ,  $i = 1, 2, \dots, m$ .

**THEOREM 1<sup>(1)</sup>.** Let  $Y_1, Y_2, \dots, Y_m$  be independent random elements of  $(D, d)$  satisfying the CLT in  $(D, d)$ . Then

$$Y = \sum_{i=1}^m \delta_i Y_i$$

satisfies the CLT in  $(D, d)$ .

**Proof.** Let

$$\zeta_n = \sum_{k=1}^n \sum_{i=1}^m \delta_{k,i} Y_{k,i}, \quad n \geq 1,$$

where  $\{Y_{k,i}, k \geq 1\}$ ,  $i = 1, 2, \dots, m$ , and  $\{\delta_k = (\delta_{k,1}, \delta_{k,2}, \dots, \delta_{k,m}), k \geq 1\}$  are independent sequences of independent copies of  $Y_i$ ,  $i = 1, 2, \dots, m$ , and of

<sup>(1)</sup> Theorem 1 was suggested by Prof. C. Ryll-Nardzewski.

$\delta = (\delta_1, \delta_2, \dots, \delta_m)$ , respectively. Notice that  $\xi_n$  has the same distribution as the random element

$$\xi_n = \sum_{i=1}^m \sum_{k=1}^{v_{n,i}} Y_{k,i},$$

where  $v_n = (v_{n,1}, v_{n,2}, \dots, v_{n,m})$  is a random vector in  $R^m$  with multinomial distribution, i.e.

$$P\{v_n = (k_1, k_2, \dots, k_m)\} = \frac{n!}{k_1! k_2! \dots k_m!} p_1^{k_1} p_2^{k_2} \dots p_m^{k_m},$$

and  $v_n$  is independent of  $Y_{n,i}, i = 1, 2, \dots, m, n \geq 1$ .

Write

$$A_i(t) = E Y_{n,i}(t), \quad t \geq 0, \quad A = (A_1, A_2, \dots, A_m), \quad \bar{Y}_{n,i} = Y_{n,i} - A_i,$$

$$\tilde{Y}_{n,i}(s) = \frac{1}{\sqrt{n}} \sum_{j=1}^{[ns]} \bar{Y}_{j,i}, \quad \tilde{v}_{n,i}(s) = \frac{sv_{n,i}}{n}, \quad s \geq 0,$$

$$\tilde{Y}_n = (\tilde{Y}_{n,1}, \tilde{Y}_{n,2}, \dots, \tilde{Y}_{n,m}), \quad \tilde{v}_n = (\tilde{v}_{n,1}, \tilde{v}_{n,2}, \dots, \tilde{v}_{n,m}),$$

$$V_{n,i} = \frac{1}{\sqrt{n}} (v_{n,i} - np_i), \quad V_n = (V_{n,1}, V_{n,2}, \dots, V_{n,m}).$$

Note that  $\tilde{Y}_{n,i}, \tilde{v}_{n,i}, \tilde{Y}_n, \tilde{v}_n$  are random elements of  $(\tilde{D}, \tilde{d}), (D_0, d), (\tilde{D}^m, \tilde{d}^m)$  and  $(D_0^m, d^m)$ , respectively, and  $A_i \in D, A \in D^m$ .

Using Theorem 1 from [2] we infer that  $\{\tilde{Y}_{n,i}\}$  converges in distribution in  $(\tilde{D}, \tilde{d})$  to a Gaussian random element  $\mathcal{W}_i$  of  $(\tilde{D}, \tilde{d})$ . Furthermore,  $\mathcal{W}_i$  is a homogeneous random element of  $(\tilde{D}, \tilde{d})$  with independent increments having continuous paths with probability one and such that  $\mathcal{W}_i(0) = 0$  and  $\mathcal{W}_i(1)$  has the same distribution as the limit in distribution of

$$\left\{ \frac{1}{\sqrt{n}} \sum_{j=1}^n \bar{Y}_{j,i} \right\}.$$

Hence  $\tilde{Y}_n \xrightarrow{D} \mathcal{W} = (\mathcal{W}_1, \mathcal{W}_2, \dots, \mathcal{W}_m)$  in  $(\tilde{D}^m, \tilde{d}^m)$ . By the law of large numbers and the Central Limit Theorem for multinomial distribution we have

$$v_n \rightarrow v \text{ a.e.} \quad \text{and} \quad V_n \xrightarrow{D} N = (N_1, N_2, \dots, N_m),$$

where  $v(t) = (p_1 t, p_2 t, \dots, p_m t)$  and  $N$  is a Gaussian random vector in  $R^m$  the expected value of which is the zero vector. From the separability of the metric spaces  $\tilde{D}^m, D_0^m$  and  $R^m$  it follows that  $(\tilde{Y}_n, \tilde{v}_n, V_n)$  is a random element of  $\tilde{D}^m \times D_0^m \times R^m$  with the product topology. Hence and from properties of  $\tau$  we obtain

$$(1) \quad (\tau(\tilde{Y}_n, v_n), V_n) \xrightarrow{D} (\tau(\mathcal{W}, v), N)$$

in  $\tilde{D}^m \times R^m$  with the product topology. Since the mapping  $+$  is continuous on  $\tilde{C} \times D$  (see [6]), so (1) yields

$$(2) \quad \sum_{i=1}^m \tau(\tilde{Y}_{n,i}, \tilde{v}_{n,i})(1) + \sum_{i=1}^m V_{n,i} A_i \xrightarrow{D} \sum_{i=1}^m \sqrt{p_i} \mathcal{W}_i(1) + \sum_{i=1}^m N_i A_i$$

in  $(D, d)$ . The left-hand side of (2) is equal to the random element

$$\frac{1}{\sqrt{n}} \sum_{i=1}^m \sum_{j=1}^{v_{n,i}} \bar{Y}_{j,i} + \frac{1}{\sqrt{n}} \sum_{i=1}^m (v_{n,i} - np_i) A_i$$

which at  $t$  is equal to

$$\frac{1}{\sqrt{n}} (\xi_n(t) - E\xi_n(t)).$$

Hence we obtain the assertion of Theorem 1.

Note that the arguments above prove Theorem 1 also for  $m = \infty$ .

Remark 1. For  $m = 2$  the covariance function  $r$  of the limiting Gaussian process is of the form

$$r(s, t) = pr_1(s, t) + (1-p)r_2(s, t) + \eta(s)\eta(t)p(1-p), \quad 0 \leq s \leq t < \infty,$$

where  $r_1$  and  $r_2$  are the covariance functions of  $Y_1$  and  $Y_2$ , respectively, and  $\eta(t) = E Y_1(t) - E Y_2(t)$ ,  $t \geq 0$ .

**3. Breakdown processes.** Let  $\{u_n, n \geq 0\}$ ,  $\{v_n, n \geq 1\}$ ,  $\{u'_n, n \geq 1\}$  and  $\{v'_n, n \geq 0\}$  be independent sequences of positive random variables which have no atoms at zero and are defined on the same probability space. Assume that  $u_0$  and  $v'_0$  have distribution functions (d.f.'s)  $G_0$  and  $F_0$ , respectively,  $u_n, u'_n$  ( $n \geq 1$ ) have d.f.  $G$  and  $v_n, v'_n$  ( $n \geq 1$ ) have d.f.  $F$ .

Let

$$Z_0 = u_0, \quad Z_n = Z_{n-1} + v_n + u_n, \quad n \geq 1,$$

$$Z'_0 = 0, \quad Z'_n = Z'_{n-1} + v'_{n-1} + u'_n, \quad n \geq 1.$$

Define the process  $X_0$  setting  $X_0(\omega, t) = 1$  if there exists an  $n \geq 0$  such that  $Z_n(\omega) \leq t < Z_n(\omega) + v_{n+1}(\omega)$  and setting  $X_0(\omega, t) = 0$  if  $t \leq u_0(\omega)$  or if there exists an  $n \geq 0$  such that  $Z_n(\omega) + v_{n+1}(\omega) \leq t < Z_{n+1}(\omega)$ . Analogically, define the process  $X_1$  setting  $X_1(\omega, t) = 1$  if there exists an  $n \geq 0$  such that  $Z'_n(\omega) \leq t < Z'_n(\omega) + v'_n(\omega)$  and setting  $X_1(\omega, t) = 0$  if there exists an  $n \geq 0$  such that  $Z'_n(\omega) + v'_n(\omega) \leq t < Z'_{n+1}(\omega)$ . Hence  $X_0$  and  $X_1$  are random elements of  $(D, d)$ .

Let  $\delta$  be a random variable taking on the values 1 and 0 with probability  $p$  and  $1-p$ , respectively,  $0 < p < 1$ . Define the process  $X$  setting  $X = \delta X_1 + (1-\delta) X_0$ . In reliability theory,  $X_0$ ,  $X_1$  and  $X$  are known as breakdown processes (see [3], Section 7).

For each  $s \geq 0$  define  $\gamma(s)$  as the time to the nearest change (to the right of  $s$ ) of a state of the process  $X$ , i.e.

$$\gamma(s) = \inf \{t - s, t > s, X(s) \neq X(t)\}.$$

Similarly, for each  $s \geq 0$  define  $\gamma_0(s)$  and  $\gamma_1(s)$  for  $X_0$  and  $X_1$ , respectively. All those random variables have no atoms at zero.

Let

$$\begin{aligned} H' &= \sum_{n=0}^{\infty} F^{*n} * G^{*n}, & H_{0,0} &= G_0 * F * H', & H_{0,1} &= G_0 * H', \\ H_{1,0} &= F_0 * H', & H_{1,1} &= F_0 * G * H', & H_0 &= (1-p)H_{0,0} + pH_{1,0}, \\ & & H_1 &= (1-p)H_{0,1} + pH_{1,1}, & H &= H_0 + H_1, \end{aligned}$$

where  $*$  denotes the convolution operation of d.f.'s. All the functions are non-negative and non-decreasing.

LEMMA 1. For  $0 \leq s \leq t < \infty$ ,  $y \geq 0$  and  $i = 0, 1$  we have

$$\begin{aligned} P\{\gamma_i(s) \leq y\} &= \int_s^{s+y} (1-G(s+y-u)) dH_{i,0}(u) + \int_s^{s+y} (1-F(s+y-u)) dH_{i,1}(u), \\ P\{\gamma(s) \leq y\} &= \int_s^{s+y} (1-G(s+y-u)) dH_0(u) + \int_s^{s+y} (1-F(s+y-u)) dH_1(u). \end{aligned}$$

Proof. Put  $p_{i,j}(s, y) = P\{X_i(s) = j, \gamma_i(s) > y\}$  for  $i, j = 0, 1$ . Note that

$$\begin{aligned} p_{0,1}(s, y) &= P\{\exists_{n \geq 0} Z_n \leq s < Z_n + v_{n+1}, \gamma_0(s) > y\} \\ &= \sum_{n \geq 0} P\{Z_n \leq s < Z_n + v_{n+1}, Z_n + v_{n+1} - s > y\} \\ &= \sum_{n \geq 0} \int_0^s P\{u \leq s < u + v_{n+1}, u + v_{n+1} - s > y\} dP\{Z_n \leq u\} \\ &= \int_0^s 1 - F(s+y-u) dH_{0,1}(u). \end{aligned}$$

Now

$$\begin{aligned} p_{0,0}(s, y) &= P\{s < Z_0, Z_0 - s > y\} + \\ &\quad + P\{\exists_{n \geq 0} Z_n + v_n \leq s < Z_n + v_{n+1} + u_{n+1}, Z_n + v_{n+1} + u_{n+1} - s > y\} \\ &= 1 - G_0(s+y) + \sum_{n \geq 0} \int_0^s (1-G(s+y-u)) dP\{Z_n + v_{n+1} \leq u\} \\ &= 1 - G_0(s+y) + \int_0^s (1-G(s+y-u)) dH_{0,0}(u). \end{aligned}$$

Analogically we obtain

$$p_{1,0}(s, y) = \int_0^s (1 - G(s + y - u)) dH_{1,0}(u),$$

$$p_{1,1}(s, y) = 1 - F_0(s + y) + \int_0^s (1 - F(s + y - u)) dH_{1,1}(u).$$

Notice that

$$(3) \quad P\{\gamma_0(s) > y\} = \int_0^s (1 - F(s + y - u)) dH_{0,1}(u) + \int_0^s (1 - G(s + y - u)) dH_{0,0}(u) + 1 - G_0(s + y),$$

$$(4) \quad P\{\gamma_1(s) > y\} = \int_0^s (1 - F(s + y - u)) dH_{1,1}(u) + \int_0^s (1 - G(s + y - u)) dH_{1,0}(u) + 1 - F_0(s + y).$$

Setting  $y = 0$  in (3) and (4) and using the fact that

$$P\{\gamma_0(s) > 0\} = P\{\gamma_1(s) > 0\} = 1$$

we obtain

$$\int_0^s (1 - F(s - u)) dH_{i,1}(u) + \int_0^s (1 - G(s - u)) dH_{i,0}(u) = \begin{cases} G_0(s) & \text{if } i = 0, \\ F_0(s) & \text{if } i = 1. \end{cases}$$

Now, putting the so-calculated  $G_0(s + y)$  and  $F_0(s + y)$  in (3) and (4), respectively, we obtain the assertion of Lemma 1 for  $i = 0, 1$ . The second assertion follows from the first one and from the definition of  $X$ .

**COROLLARY 1.** For  $0 \leq s \leq t < \infty$  and  $i = 0, 1$  we have

$$P\{\gamma_i(s) \leq t - s\} \leq H_i(t) - H_i(s), \quad P\{\gamma(s) \leq t - s\} \leq H(t) - H(s).$$

Define a function  $\tilde{F}$  by

$$\tilde{F}(t) = \max \{G_0(t), F_0(t), G(t), F(t)\}, \quad t \geq 0.$$

**LEMMA 2.** For  $0 \leq t_1 \leq t \leq t_2$  and  $i = 0, 1$  we have

$$(5) \quad P\{X_i(t_1) = X_i(t_2) \neq X_i(t)\} \leq P\{\gamma_i(t_1) < t - t_1\} \tilde{F}(t_2 - t_1),$$

$$(6) \quad P\{X(t_1) = X(t_2) \neq X(t)\} \leq P\{\gamma(t_1) < t - t_1\} \tilde{F}(t_2 - t_1).$$

**Proof.** We prove (5) for  $i = 0$ . The other assertions are proved in a similar way.

Define random variables  $s_n$  and  $S_n$  by  $s_{2n} = u_n$ ,  $s_{2n+1} = v_{n+1}$  and  $S_n = s_0 + s_1 + \dots + s_{n-1}$ ,  $n \geq 1$ . Let

$$N(t) = \begin{cases} 0 & \text{if } s_0 > t, \\ \max \{k: S_{k-1} \leq t\} & \text{if } s_0 \leq t. \end{cases}$$

If  $X_0(t_1) = X_0(t_2) \neq X(t)$  for  $t_1 < t < t_2$ , then  $N(t_1) < N(t) < N(t_2)$ . Thus

$$(7) \quad P\{X_0(t_1) = X_0(t_2) \neq X_0(t)\} \leq P\{S_{N(t_1)} < t, S_{N(t)} < t_2, N(t) - N(t_1) > 0\} \\ = \int_0^{t-t_1} P\{S_{N(t)} < t_2, N(t) - N(t_1) > 0 \mid S_{N(t_1)} - t_1 = u\} dP\{S_{N(t_1)} - t_1 \leq u\}.$$

Since  $S_{N(t_1)} - t_1 = \gamma_0(t_1)$ , the right-hand side of (7) is equal to

$$\int_0^{t-t_1} P\{S_{N(t)} < t_2, N(t) - N(t_1) > 0 \mid \gamma_0(t_1) = u\} dP\{\gamma_0(t_1) \leq u\} \\ = \int_0^{t-t_1} P\{S_{N(t)} - S_{N(t_1)} < t_2 - t_1 - u, N(t) - N(t_1) > 0 \mid \gamma_0(t_1) = u\} dP\{\gamma_0(t_1) \leq u\}.$$

Note that the integrand does not exceed

$$(8) \quad P\{S_{N(t)} - S_{N(t_1)} < t_2 - t_1, N(t) - N(t_1) > 0 \mid \gamma_0(t_1) = u\} \\ = \sum_{k=1}^{\infty} \sum_{k_1=0}^{k-1} P\{S_k - S_{k_1} < t_2 - t_1 \mid N(t) = k, N(t_1) = k_1, \gamma_0(t_1) = u\} \times \\ \times P\{N(t) = k, N(t_1) = k_1 \mid \gamma_0(t_1) = u\} \\ \leq \sum_{k=1}^{\infty} \sum_{k_1=0}^{k-1} \tilde{F}(t_2 - t_1) P\{N(t) = k, N(t_1) = k_1 \mid \gamma_0(t_1) = u\} \leq \tilde{F}(t_2 - t_1).$$

Hence in view of (7) and (8) we obtain (5) for  $i = 0$ .

**THEOREM 2.** If

$$\lim_{t \rightarrow 0} \tilde{F}(t) t^{-\alpha} < \infty \quad \text{for some } \alpha > 0,$$

then for each  $c > 0$  there exists a number  $b$  such that for  $t_1 \leq t \leq t_2$  and  $i = 0, 1$  the following inequalities hold:

$$(9) \quad P\{X_i(t_1) = X_i(t_2) \neq X_i(t)\} \leq b(\tilde{H}_i(t_2) - \tilde{H}_i(t_1))^{2\beta_1},$$

$$(10) \quad P\{X(t_1) = X(t_2) \neq X(t)\} \leq b(\tilde{H}(t_2) - \tilde{H}(t_1))^{2\beta_1},$$

where  $\beta_1 = \min(1, \alpha)$ ,  $\tilde{H}_i(t) = H_i(t) + t$ ,  $\tilde{H}(t) = H(t) + t$ .

**Proof.** There exist numbers  $u$  and  $b_1$ ,  $u \leq c$ , such that  $\tilde{F}(t) \leq b_1 t^\alpha$  for  $t \leq u$ . Hence for  $c$  there exists a constant  $b_2$  such that  $\tilde{F}(t) \leq b_2 t^\alpha$  for

$t \leq c$ . By Lemma 2 and Corollary 1, for  $0 \leq t_1 \leq t \leq t_2 \leq c$  we have

$$(11) \quad P\{X_i(t_1) = X_i(t_2) \neq X_i(t)\} \leq b_2(H_i(t_2) - H_i(t_1))(t_2 - t_1)^\alpha.$$

Now there exists a number  $b$  such that the right-hand side of (11) does not exceed

$$b(H_i(t_2) - H_i(t_1))(t_2 - t_1)^{\beta_1} \leq b(\tilde{H}_i(t_2) - H_i(t_1))^{2\beta_1}.$$

This completes the proof of Theorem 2.

**4. CLT for breakdown processes.** From the definition of  $X = \delta X_1 + (1 - \delta)X_0$  and Theorem 1 we get

**COROLLARY 2.** *If  $X_1$  and  $X_0$  satisfy the CLT in  $(D, d)$ , then  $X$  satisfies the CLT in  $(D, d)$ .*

The following technical lemma will be used in the sequel:

**LEMMA 3.** *If  $u$  and  $v$  are random variables such that  $|u| \leq 2$ ,  $|v| \leq 2$ ,  $Eu^2 \leq a$ ,  $Ev^2 \leq a$  and  $Eu^2v^2 \leq a^{2\alpha}$  for some  $\alpha > 0$ , then*

$$E(u - Eu)^2(v - Ev)^2 \leq Aa^\beta,$$

where  $\beta = \min(2\alpha, 3/2)$ ,  $A = 90 B^{\beta_0 - \beta}$ ,  $\beta_0 = 2 \max(1, \alpha)$ , and  $B$  is such that  $B > 1$  and  $16a/B < 1$ .

**Proof.** Calculating  $(u - Eu)^2(v - Ev)^2$ , taking the expectation and next using the Schwarz inequality, we see that  $E(u - Eu)^2(v - Ev)^2$  does not exceed

$$(12) \quad Eu^2v^2 + 3Eu^2Ev^2 + 2Ev^2(Eu^4)^{1/2} + 2Eu^2(Ev^4)^{1/2} + 4(Eu^2Ev^2Eu^4Ev^4)^{1/2}.$$

By assumptions we infer that (12) does not exceed

$$\begin{aligned} a^{2\alpha} + 7a^2 + 16a^{3/2} &\leq B^{\beta_0} \left[ \left(\frac{a}{B}\right)^2 + \left(\frac{7a}{B}\right)^2 + \left(\frac{16a}{B}\right)^{3/2} \right] \\ &\leq B^{\beta_0} \left[ \left(\frac{a}{B}\right)^\beta + \left(\frac{7a}{B}\right)^\beta + \left(\frac{16a}{B}\right)^\beta \right] \leq Aa^\beta B^{\beta_0 - \beta}. \end{aligned}$$

This completes the proof.

**THEOREM 3.** *If*

$$\lim_{t \rightarrow 0} \tilde{F}(t)t^{-\alpha} < \infty \quad \text{for some } \alpha > \frac{1}{2}$$

and  $H_0$  and  $H_1$  are continuous, then  $X_0$  and  $X_1$  satisfy the CLT in  $(D, t)$ .

**Proof.** Note that the following equalities are true for  $0 \leq t_1 \leq t \leq t_2 < \infty$  and  $i = 0, 1$ :

$$(13) \quad E(X_i(t) - X_i(t_1))^2 = P\{X_i(t_1) \neq X_i(t)\} \leq P\{\gamma_i(t_1) < t - t_1\},$$

$$E(X_i(t) - X_i(t_1))^2(X_i(t_2) - X_i(t))^2 = P\{X_i(t_1) = X_i(t_2) \neq X_i(t)\}.$$



By (13) and Corollary 1 we have

$$E(X_i(t) - EX_i(t) - X_i(t_1) + EX_i(t_1))^2 \leq E(X_i(t) - X_i(t_1))^2 \leq H_i(t) - H_i(t_1).$$

Now Lemma 3 yields, for  $0 \leq t_1 \leq t \leq t_2 \leq c$ ,

$$E(X_i(t) - EX_i(t) - X_i(t_1) + EX_i(t_1))^2 (X_i(t_2) - EX_i(t_2) - X_i(t) + EX_i(t))^2 \leq A(H_i(t_2) + t_2 - H_i(t_1) - t_1)^\beta,$$

where  $A$  depends on  $c$  and  $\beta = \min(2\alpha, 3/2)$ . Now using Theorem 2 from [4] we infer that  $X_i$  ( $i = 0, 1$ ) satisfy the CLT in  $D[0, c]$ ,  $c > 0$ , with the Skorohod topology. Hence, in view of Theorem 3' in [5], we conclude that  $X_i$  ( $i = 0, 1$ ) satisfy the CLT in  $(D, d)$ . This completes the proof of Theorem 3.

Using the arguments analogical to those in [1] (see p. 151-153) it can be shown that if  $F_0, G_0, F$  and  $G$  are continuous, then  $X$  does not satisfy the CLT in  $D$  endowed with the topology of uniform convergence on compacta in  $[0, \infty)$ . It is a consequence of the fact that

$$\frac{1}{\sqrt{n}}(X_1 + X_2 + \dots + X_n),$$

with  $X_i$  being independent copies of  $X$ , is a random element of  $(D, d)$  but not a random element of  $D$  with the topology of uniform convergence on compacta in  $[0, \infty)$ .

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Institute of Mathematics, Wrocław University  
pl. Grunwaldzki 2/4, 50-384 Wrocław, Poland

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