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# On the asymptotic equidistribution of sums of independent identically distributed random variables

by

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**ABSTRACT.** — For a sum  $S_n$  of  $n$  I.I.D. random variables the idea of approximate equidistribution is made precise by introducing a notion of asymptotic translation invariance. The distribution of  $S_n$  is shown to be asymptotically translation invariant in this sense iff  $S_1$  is nonlattice. Some ramifications of this result are given.

*Key words* : Sums of I.I.D. random variables, asymptotic equidistribution, asymptotic translation invariance.

**RÉSUMÉ.** — On introduit, pour une somme  $S_n$  de  $n$  variables aléatoires indépendantes équidistribuées, une notion d'invariance asymptotique par translation, qui permet de rendre précise l'idée d'équidistribution approximative. On montre que la loi de  $S_n$  est asymptotiquement invariante par translation en ce sens si, et seulement si, la loi de  $S_1$  est non arithmétique. On donne quelques extensions de ce résultat.

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## 1. INTRODUCTION

Let  $T_1, T_2, \dots$  be a sequence of independent random variables with a common distribution  $P^{T_n} = P^{T_1}$  and let  $S_n = T_1 + \dots + T_n$ . Intuitively, if

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$E(T_1)=0$  and  $T_1 \neq 0$ , the mass of the probability measure  $P^{S_n}$  is expected to be approximately “equidistributed”, as  $n$  becomes large. If  $E(T_1)>0$ , one is inclined to think of something like an “approach to uniformity at infinity”. An old result of this kind is due to Robbins (1953). If  $T_1$  is not concentrated on a lattice,

$$n^{-1} \sum_{i=1}^n h(S_i) \rightarrow \lim_{T \rightarrow \infty} (2T)^{-1} \int_{-T}^T h(x) dx, \quad \text{as } n \rightarrow \infty \quad (1.1)$$

for all almost periodic functions  $h$  (i. e. if  $h$  is the uniform limit of trigonometric polynomials). For a partial sharpening of this result see Theorem 3 of Stadje (1985). In the case when  $T_1$  is not concentrated on a lattice,  $E(T_1)=0$  and  $0 < \sigma^2 := \text{Var}(T_1) < \infty$ , the expectation of asymptotic equidistribution can also be justified by the limiting relation

$$\sigma(2\pi n)^{1/2} P(S_n \in I) \rightarrow \lambda(I), \quad \text{as } n \rightarrow \infty \quad (1.2)$$

which is valid for all bounded intervals  $I \subset \mathbb{R}$ , where  $\lambda$  denotes the Lebesgue measure (Shepp (1964), Stone (1965, 1967), Breiman (1968), chapt. 10).

One might try to interpret approximate uniformity of  $P^{S_n}$  by stating that  $P(S_n \in I)$  asymptotically only depends on the length of  $I$ . Since  $\lim_{n \rightarrow \infty} P(S_n \in I) = 0$  for every bounded interval  $I$ , this idea should be made

reasonable by examining the speed of convergence of  $P(S_n \in I)$ . This is done in (1.2) stating  $a_n P^{S_n}$  approaches the Lebesgue measure, where  $a_n = \sigma(2\pi n)^{1/2}$ .

In this paper another approach to the idea of equidistribution of  $P^{S_n}$  is developed. The essential property of an “equidistribution” is the invariance under translations. To measure the degree of translation invariance of a probability measure  $Q$  on  $\mathbb{R}$ , we introduce, for  $a \in \mathbb{R}$  and  $t > s > 0$ , the quantities

$$d(a, t, Q) := \sum_{i=-\infty}^{\infty} |Q((a+it, a+(i+1)t]) - Q((a+(i-1)t, a+it])| \quad (1.3)$$

$$D(t, Q) := \sup_{a \in \mathbb{R}} d(a, t, Q) \quad (1.4)$$

$$\tilde{D}(s, t, Q) := \sup_{s \leq u \leq t} D(u, Q). \quad (1.5)$$

We call a sequence  $(Q_n)_{n \geq 1}$  of probability measures *asymptotically translation invariant* (ATI), if  $\lim_{n \rightarrow \infty} \tilde{D}(s, t, Q_n) = 0$  for all  $t > s > 0$ . The main

theorem of this paper states that  $(P^{S_n})_{n \geq 1}$  is ATI if, and only if,  $P^{T_1}$  is not concentrated on a lattice. No moment conditions are needed for this equivalence. Let  $D_n(t) := \tilde{D}(t^{-1}, t, P^{S_n})$ ,  $t > 1$ . Regarding the speed of convergence of  $D_n(t)$  we remark that

$$\liminf_{n \rightarrow \infty} n^{1/2} D_n(t) > 0 \quad \text{for all } t > 1, \quad \text{if } E(T_1^2) < \infty. \quad (1.6)$$

To see (1.6), let without loss of generality  $E(T_1) = 0$  and  $E(T_1^2) = 1$ . Then, by Chebyshev's inequality,

$$P(|S_n| < n^{1/2}) \geq 1 - n^{-1}. \tag{1.7}$$

The interval  $(-n^{1/2}, n^{1/2})$  can be covered by  $[2n^{1/2}/t] + 1$  half-open intervals of length  $t$ . One of these intervals, say  $I$ , obviously satisfies

$$P(S_n \in I) \geq (1 - n^{-1}) / ([2n^{1/2}/t] + 1) \geq \frac{1}{2} \frac{1}{2n^{1/2}t + 1}, \quad \text{if } n \geq 2. \tag{1.8}$$

Choose  $a \in [0, t]$  and  $i_0 \in \mathbb{Z}$  such that  $I = (a + i_0 t, a + (i_0 + 1)t]$ .

Then

$$\begin{aligned} d(a, t, P^{S_n}) &\geq \sum_{i=-\infty}^{i_0} [P(S_n \in (a + it, a + (i + 1)t]) \\ &\quad - P(S_n \in (a + (i - 1)t, a + it))] \\ &= P(S_n \in (a + i_0 t, a + (i_0 + 1)t]) \geq \frac{1}{4t + 2} n^{-1/2}, \quad n \geq 2 \end{aligned} \tag{1.9}$$

(1.6) follows from (1.9).

In order to derive a converse result to (1.6), we need a further notion. A distribution  $Q$  on  $\mathbb{R}$  is called strongly nonlattice, if its characteristic function  $\varphi$  satisfies

$$\limsup_{|\zeta| \rightarrow \infty} |\varphi(\zeta)| < 1. \tag{1.10}$$

The second main result of this note is that if  $P^{T_1}$  is strongly nonlattice,

$$\limsup_{n \rightarrow \infty} n^{1/2} D_n(t) < \infty \quad \text{for all } t > 1. \tag{1.11}$$

## 2. THE MAIN THEOREM

We shall prove

**THEOREM 1.** — *The following two statements are equivalent.*

(a)  $P^{T_1}$  is nonlattice.

(b)  $(P^{S_n})_{n \geq 1}$  is ATI.

*Proof.* — (a)  $\Rightarrow$  (b). Assume first that  $E|T_1|^3 < \infty$  and  $E(T_1) = 0$ . Let  $\sigma^2 = E(T_1^2)$ ,  $\mu_3 = E(T_1^3)$  and denote the distribution function of  $S_n$  by  $F_n$ . Then

$$\begin{aligned} d(a, t, P^{S_n}) &= \sum_{i=-\infty}^{\infty} |F_n(a + (i + 1)t) - 2F_n(a + it) \\ &\quad + F_n(a + (i - 1)t)|. \end{aligned} \tag{2.1}$$

Since  $P^{T_1}$  is nonlattice, a well-known expansion for distribution functions yields

$$F_n(n^{1/2} \sigma x) = \Phi(x) + \frac{\mu_3}{6 \sigma^3 n^{1/2}} (1 - x^2) \varphi(x) + \varepsilon_n(x) n^{-1/2} \tag{2.2}$$

for all  $x \in \mathbb{R}$ , where

$$\varepsilon_n := \sup_{x \in \mathbb{R}} |\varepsilon_n(x)| \rightarrow 0, \quad \text{as } n \rightarrow \infty \tag{2.3}$$

and  $\Phi$  and  $\varphi$  are the distribution function and density of  $N(0, 1)$  (see e. g. Feller (1971), p. 539). It is easy to check that for each  $j \in \mathbb{N}$  and  $a \in [0, t]$

$$\begin{aligned} \sum_{i > j} |F_n(a + (i + 1)t) - 2F_n(a + it) + F_n(a + (i - 1)t)| \\ \leq 1 - F_n(a + (j + 1)t) + 1 - F_n(a + jt) \leq 2P(S_n > jt) \end{aligned} \tag{2.4}$$

and

$$\begin{aligned} \sum_{i < -j} |F_n(a + (i + 1)t) - 2F_n(a + it) + F_n(a + (i - 1)t)| \\ \leq F_n(a - jt) + F_n(a - (j + 1)t) \leq 2P(S_n \leq -(j - 1)t). \end{aligned} \tag{2.5}$$

Inserting (2.2)-(2.5) into (2.1) we obtain, for  $a \in [0, t]$ ,

$$\begin{aligned} d(a, t, P^{S_n}) \leq 2P(|S_n| \geq (j - 1)t) + (2j + 1)\varepsilon_n n^{-1/2} \\ + \sum_{i = -\infty}^{\infty} \left| \Phi\left(\frac{a + (i + 1)t}{\sigma n^{1/2}}\right) - 2\Phi\left(\frac{a + it}{\sigma n^{1/2}}\right) + \Phi\left(\frac{a + (i - 1)t}{\sigma n^{1/2}}\right) \right| \\ + \frac{|\mu_3|}{6 \sigma^3 n^{1/2}} \sum_{i = -\infty}^{\infty} |(1 - x_{i+1}^2) \varphi(x_{i+1}) \\ - 2(1 - x_i^2) \varphi(x_i) + (1 - x_{i-1}^2) \varphi(x_{i-1})| \end{aligned} \tag{2.6}$$

where  $x_i := (a + it)/\sigma n^{1/2}$ . By Chebyshev's inequality,

$$2P(|S_n| \geq (j - 1)t) + (2j + 1)\varepsilon_n n^{-1/2} \leq \frac{2\sigma^2 n}{t^2(j - 1)^2} + (2j + 1)\varepsilon_n n^{-1/2}. \tag{2.7}$$

The smallest order of magnitude of the righthand side of (2.7) is attained for  $j = j_n$  being equal to the integer part of  $n^{1/2} \varepsilon_n^{-1/3}$ ; in this case

$$\begin{aligned} 2P(|S_n| \geq (j_n - 1)t) + (2j_n + 1)\varepsilon_n n^{-1/2} \\ = (1 + t^{-2}) O(\varepsilon_n^{2/3}), \quad \text{as } n \rightarrow \infty. \end{aligned} \tag{2.8}$$

Next we estimate the two series in (2.6). Let  $X$  be a standard normal random variable. Then the first sum at the righthand side of (2.6) is equal

to

$$\begin{aligned} & \sum_{i=-\infty}^{\infty} \left| \mathbb{P}(\sigma n^{1/2} X \in (a+it, a+(i+1)t)) \right. \\ & \quad \left. - \mathbb{P}(\sigma n^{1/2} X - t \in (a+it, a+(i+1)t)) \right| \\ & \leq (\sigma n^{1/2})^{-1} \int_{-\infty}^{\infty} \left| \varphi(x/\sigma n^{1/2}) - \varphi((x+t)/\sigma n^{1/2}) \right| dx \\ & = 2(\sigma^{1/2})^{-1} \left[ \int_{-\infty}^{-t/2} \varphi((x+t)/\sigma n^{1/2}) dx - \int_{-\infty}^{-t/2} \varphi(x/\sigma n^{1/2}) dx \right] \\ & = 2 \int_{-t/2 \sigma n^{1/2}}^{t/2 \sigma n^{1/2}} \varphi(u) du = t O(n^{-1/2}), \quad \text{as } n \rightarrow \infty. \end{aligned} \tag{2.9}$$

To estimate the last sum at the right side of (2.6), note that the function  $(1-x^2)\exp(-x^2/2)$  has four points of inflexion. Regarding the sequence

$$a_i := (1-x_{i+1}^2)\varphi(x_{i+1}) - (1-x_i^2)\varphi(x_i), \quad i \in \mathbb{Z}$$

this implies that  $(a_i - a_{i-1})_{i \in \mathbb{Z}}$  changes signs at most four times. Using its telescoping form the sum in question can be bounded from above as follows:

$$\sum_{i=-\infty}^{\infty} |a_i - a_{i-1}| \leq 8 \sup_{-\infty < i < \infty} |a_i|. \tag{2.10}$$

Further, by the mean value theorem,

$$|a_i| \leq |x_{i+1} - x_i| \sup_{x \in \mathbb{R}} \left| \frac{d}{dx} (1-x^2)\varphi(x) \right| = K t/\sigma n^{1/2} \tag{2.11}$$

for some constant K. Inserting (2.8)-(2.11) into (2.6) we arrive at

$$d(a, t, P^{S_n}) = (1+t^{-2}) O(\varepsilon_n^{2/3}) + t O(n^{-1/2}), \quad \text{as } n \rightarrow \infty, \tag{2.12}$$

so that

$$D(t, P^{S_n}) = (1+t^{-2}) O(\varepsilon_n^{2/3}) + t O(n^{-1/2}), \quad \text{as } n \rightarrow \infty. \tag{2.13}$$

To establish the assertion without moment conditions we first remark that  $D(t, Q)$  is translation invariant in the sense that

$$D(t, Q) = D(t, Q * \varepsilon_x) \quad \text{for all } x \in \mathbb{R}, \quad t > 0, \tag{2.14}$$

where  $*$  denotes convolution and  $\varepsilon_x$  is the point mass at  $x$ . Thus, (2.13) holds, if  $E|T_1|^3 < \infty$  (without the assumption  $E(T_1) = 0$ ). Further, for probability measures Q and R we have

$$D(t, Q * R) \leq D(t, Q). \tag{2.15}$$

(2.15) is proved as follows:

$$\begin{aligned}
 D(t, Q \star R) &= \sup_a \sum_{i=-\infty}^{\infty} \left| (Q \star R)((a+it, a+(i+1)t]) \right. \\
 &\quad \left. - (Q \star R)((a+(i-1)t, a+it]) \right| \\
 &\leq \sup_a \sum_{i=-\infty}^{\infty} \int_{-\infty}^{\infty} \left| Q((a+it-x, a+(i+1)t-x)) \right. \\
 &\quad \left. - Q((a+(i-1)t-x, a+it-x)) \right| dR(x) \\
 &\leq \int_{-\infty}^{\infty} \sup_a \sum_{i=-\infty}^{\infty} \left| Q((a+it-x, a+(i+1)t-x)) \right. \\
 &\quad \left. - Q((a+(i-1)t-x, a+it-x)) \right| dR(x) \\
 &= \int_{-\infty}^{\infty} D(t, Q) dR(x) = D(t, Q). \quad (2.16)
 \end{aligned}$$

Next suppose that  $P^{T_1} = \alpha Q + (1-\alpha) R$  for some  $\alpha \in (0, 1]$  and probability measures  $Q$  and  $R$  such that  $Q$  satisfies, for some constants  $K_1, K_2$ ,

$$D(t, Q^{*n}) \leq (1+t^{-2}) K_1 \varepsilon_n^{2/3} + t K_2 n^{-1/2}, \quad \text{as } n \rightarrow \infty \quad (2.17)$$

( $Q^{*n}$  is the  $n$ -fold convolution of  $Q$  with itself). Then

$$\begin{aligned}
 D(t, P^{S_n}) &= \sup_a \sum_{i=-\infty}^{\infty} \left| \sum_{l=0}^n \binom{n}{l} \alpha^l (1-\alpha)^{n-l} \{ Q^{*l} \star R^{*(n-l)}((a+it, \right. \\
 &\quad \left. a+(i+1)t]) - Q^{*l} \star R^{*(n-l)}((a+(i-1)t, a+it]) \} \right| \\
 &\leq \sum_{l=0}^n \binom{n}{l} \alpha^l (1-\alpha)^{n-l} D(t, Q^{*l} \star R^{*(n-l)}) \\
 &\leq \sum_{l=0}^n \binom{n}{l} \alpha^l (1-\alpha)^{n-l} D(t, Q^{*l}). \quad (2.18)
 \end{aligned}$$

Let  $\delta_n := \sup_{l>n} \varepsilon_l$ . Then  $\delta_n \downarrow 0$  and, by Chebyshev's inequality for the binomial distribution and (2.17), it follows that, for arbitrary  $\varepsilon \in (0, \alpha)$ ,

$$\begin{aligned}
 \sum_{l=0}^n \binom{n}{l} \alpha^l (1-\alpha)^{n-l} D(t, Q^{*l}) &\leq 2 \sum_{l \leq (\alpha-\varepsilon)n} \binom{n}{l} \alpha^l (1-\alpha)^{n-l} \\
 &\quad + (1+t^{-2}) K_1 \sup_{l > (\alpha-\varepsilon)n} \varepsilon_l + t K_2 ((\alpha-\varepsilon)n)^{-1/2} \\
 &\leq 2\alpha(1-\alpha)\varepsilon^{-2}n^{-1} + K[(1+t^{-2})\delta_{(\alpha-\varepsilon)n} + t n^{-1/2}] \quad (2.19)
 \end{aligned}$$

where  $K = \max(K_1, K_2)$ . (2.19) implies that  $D_n(t) \rightarrow 0$ , as  $n \rightarrow \infty$ , for all  $t > 1$ .

Now we choose a function  $f$  on  $\mathbb{R}$  such that  $0 < f(x) < 1$  for all  $x \in \mathbb{R}$  and

$$\int_{-\infty}^{\infty} |x|^3 f(x) dP^{T_1}(x) < \infty.$$

Let  $\alpha := \int_{-\infty}^{\infty} f(x) dP^{T_1}(x)$  and define the probability measures  $Q$  and  $R$  by  $dQ := \alpha^{-1} f dP^{T_1}$ ,  $dR := (1-\alpha)^{-1} (1-f) dP^{T_1}$ . Then the third moment of  $Q$  is finite and  $Q$  is nonlattice so that  $Q$  satisfies (2.17). Since  $P^{T_1} = \alpha Q + (1-\alpha) R$ , it follows that  $D_n(t) \rightarrow 0$ .

(b)  $\Rightarrow$  (a). Let  $P^{T_1}$  have span  $\lambda > 0$ . If  $t \in (0, \lambda/2)$ , at most one of the successive intervals  $(a+(i-1)t, a+it]$  and  $(a+it, a+(i+1)t]$  contains a multiple of  $\lambda$ . Therefore it is obvious that  $d(a, t, P^{S_n}) = 2$  for all  $a \in \mathbb{R}$ ,  $t \in (0, \lambda/2)$  and  $n \in \mathbb{N}$ .

### 3. THE STRONGLY NONLATTICE CASE

Concerning the speed of convergence of  $D_n(t)$  we shall now prove

THEOREM 2. — *If  $P^{T_1}$  is strongly nonlattice,*

$$\limsup_{n \rightarrow \infty} n^{1/2} D_n(t) < \infty \quad \text{for all } t > 1. \tag{3.1}$$

*Proof.* — Let  $\eta := \limsup_{|\zeta| \rightarrow \infty} |\varphi(\zeta)| < 1$ . We can decompose

$P^{T_1} = \alpha Q + (1-\alpha) R$ , where  $\alpha \in (0, 1]$  and  $Q$  and  $R$  are probability measures such that  $Q$  is strongly nonlattice and concentrated on a bounded interval. If  $P^{T_1}$  itself is concentrated on a bounded interval, this is trivial. Otherwise let  $\alpha_N := P(T_1 \in [-N, N])$ , where  $N$  is large enough to ensure  $0 < \alpha_N < 1$ . Define, for Borel sets  $B$ ,

$$\begin{aligned} Q_N(B) &:= \alpha_N^{-1} P(T_1 \in B \cap [-N, N]) \\ R_N(B) &:= (1-\alpha_N)^{-1} P(T_1 \in B \setminus [-N, N]). \end{aligned}$$

Then the characteristic functions  $\tilde{\varphi}_N$  and  $\tilde{\tilde{\varphi}}_N$  of  $Q_N$  and  $R_N$  satisfy  $\tilde{\varphi}_N = \alpha_N^{-1} (\varphi - (1-\alpha_N) \tilde{\tilde{\varphi}}_N)$  so that

$$\limsup_{|\zeta| \rightarrow \infty} |\tilde{\varphi}_N(\zeta)| \leq \alpha_N^{-1} (\eta + 1 - \alpha_N), \tag{3.2}$$

and the righthand side of (3.2) is smaller than 1 for sufficiently large  $N$ , because  $\alpha_N \uparrow 1$ .

We proceed by proving the assertion for  $Q$  instead of  $P^{T_1}$ . Obviously we may assume that  $\int x dQ(x) = 0$ . Let  $F_n$  be the distribution function of



$Q^{*n}$  and  $\sigma^2 := \int x^2 dQ(x)$ . Since  $Q$  is strongly nonlattice and possesses moments of all orders, a well-known expansion yields, for every  $r \geq 3$ ,

$$F_n(n^{1/2} \sigma x) - \Phi(x) - \varphi(x) \sum_{k=3}^r n^{-(k/2)+1} R_k(x) = o(n^{-(r/2)+1}) \quad (3.3)$$

uniformly in  $x$ , where  $R_k$  is a polynomial depending only on the first  $r$  moments of  $Q$  (see, e. g., Feller (1971), p. 541). Letting  $r=5$  and proceeding as in (2.4)-(2.6) we obtain, for arbitrary  $j$ ,

$$\begin{aligned} d(a, t, Q^{*n}) \leq & 2 Q^{*n}(\mathbb{R} \setminus [-(j-1)t, (j-1)t]) \\ & + (2j+1) o(n^{-3/2}) + \sum_{i=-j}^j \left| \Phi\left(\frac{a+(i+1)t}{\sigma n^{1/2}}\right) \right. \\ & \quad \left. - 2\Phi\left(\frac{a+it}{\sigma n^{1/2}}\right) + \Phi\left(\frac{a+(i-1)t}{\sigma n^{1/2}}\right) \right| \\ & + \sum_{k=3}^5 n^{-(k/2)+1} \sum_{i=-j}^j \left| \varphi(x_{i+1}) R_k(x_{i+1}) - 2\varphi(x_i) R_k(x_i) \right. \\ & \quad \left. + \varphi(x_{i-1}) R_k(x_{i-1}) \right|. \quad (3.4) \end{aligned}$$

Here again  $x_i = (a+it)/\sigma n^{1/2}$ . Since each function  $\varphi(x) R_k(x)$  has a bounded derivative and only a finity number of points of inflexion, the same reasoning as in the proof of Theorem 1 (for  $R(x) = 1-x^2$ ) shows that, for  $k=3, 4, 5$ ,

$$\begin{aligned} \sum_{i=-\infty}^{\infty} \left| \varphi(x_{i+1}) R_k(x_{i+1}) - \varphi(x_i) R_k(x_i) \right| \\ \leq L \sup_{-\infty < i < \infty} \left| \varphi(x_{i+1}) R_k(x_{i+1}) - \varphi(x_i) R_k(x_i) \right| \\ \leq \tilde{L} \sup_{-\infty < i < \infty} |x_{i+1} - x_i| = \tilde{L} t / \sigma n^{1/2}, \quad (3.5) \end{aligned}$$

where  $L$  and  $\tilde{L}$  are appropriate constants. Thus the last term at the right side of (3.4) is  $t O(n^{-1/2})$ . Further using (2.9) for the remaining sum in (3.4) and Chebyshev's inequality we arrive at

$$\begin{aligned} d(a, t, Q^{*n}) \leq & 2 Q^{*n}(\mathbb{R} \setminus [-(j-1)t, (j-1)t]) \\ & + (2j+1) o(n^{-3/2}) + t O(n^{-1/2}) \\ & = t^{-2} O(n/j^2) + (2j+1) o(n^{-3/2}) + t O(n^{-1/2}). \quad (3.6) \end{aligned}$$

Choosing  $j = j_n = n^{5/6}$ , (3.6) implies that

$$d(a, t, Q^{*n}) = t^{-2} O(n^{-2/3}) + t O(n^{-1/2}). \quad (3.7)$$

Now arguing similarly as in (2.18) and (2.19),

$$D(t, P_n^S) \leq \sum_{l=0}^n \binom{n}{l} \alpha^l (1-\alpha)^{n-l} D(t, Q^{*l}) \\ \leq 2\alpha(1-\alpha)\varepsilon^{-2}n^{-1} + K[t^{-2}O(n^{-2/3}) + tO(n^{-1/2})], \quad (3.8)$$

where  $\varepsilon > 0$  and  $K$  are constants. It follows that  $D_n(t) = O(n^{-1/2})$  for each  $t > 1$ , as claimed.

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