

THE SPEED OF CONVERGENCE OF A MARTINGALE

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ABSTRACT

Let $X_n, n \geq 0$, be a martingale with respect to the σ -fields \mathcal{F}_n and let $B_n^2 = \sum_{i=0}^n E\{(X_i - X_{i-1})^2 | \mathcal{F}_{i-1}\}$. It is known that if $B_1^2 < \infty$ on some set Ω_0 then $X_\infty = \lim X_n$ exists and is finite a.e. on Ω_0 . We show that under suitable conditions there exists a constant $\eta < \infty$ for which $\limsup B_n^{-1} \{\log \log B_n^2\}^{-\frac{1}{2}} |X_\infty - X_{n-1}| \leq \sqrt{2(\eta + 1)}$. If "the fluctuations of B_n are small" (in the sense of the Corollary) then $\eta = 0$ and the usual upper bound of a law of the iterated logarithm results. This upper bound is not necessarily achieved, though.

1. Introduction and statement of results

Throughout $X_n, n \geq 0$, is a fixed square integrable martingale with respect to the increasing family of σ -fields $\{\mathcal{F}_n\}_{n \geq 0}$. We denote the corresponding martingale difference sequence by

$$Y_n = X_n - X_{n-1}, \quad n \geq 1,$$

and set

$$v_n^2 = E\{Y_n^2 | \mathcal{F}_{n-1}\}, \quad B_n^2 = \sum_{i=1}^n v_i^2.$$

Stout's law of the iterated logarithm ([11], see also [12], corol. 5.4.2 and p. 303) states that

$$(1.1) \quad \limsup_{n \rightarrow \infty} \left\{ \left(\sum_{i=1}^n v_i^2 \right) \log \log \left(\sum_{i=1}^n v_i^2 \right) \right\}^{-\frac{1}{2}} X_n = \sqrt{2},$$

if $B_1^2 = \infty$ and some additional hypotheses hold. Since martingales can be imbedded in Brownian motion this result is closely related to the law of the iterated logarithm for a Wiener process $\{W(t)\}_{t \geq 0}$:

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$$(1.2) \quad \limsup_{t \rightarrow \infty} (t \log \log t)^{-\frac{1}{2}} W(t) = \sqrt{2} \quad \text{w.p.1.}$$

(Compare the proof the law of the iterated logarithm for i.i.d. random variables in [2], theorem 13.25 and corollary 12.33.) It is well known that Brownian motion also satisfies a law of the iterated logarithm for small times, i.e. for each fixed s

$$(1.3) \quad \limsup_{t \rightarrow s} \{|t-s| \log \log |t-s|^{-1}\}^{-1/2} (W(t) - W(s)) = \sqrt{2} \quad \text{w.p.1}$$

(see [2], theorem (12.29)). The theorem below can be viewed as a partial analogue of (1.3) for martingales for which $B_1^2 < \infty$ on a set of positive probability.

If the increments Y_n are actually independent, better results have already been proven by Chow and Teicher [3] and Barbour [1]. The martingale case seems to have been considered first by Heyde [6]. Heyde's conditions imply all our conditions except for (1.5), after a suitable truncation of the Y_n 's (see "Special Case" below). Example (i) below describes a situation where Heyde's theorem does not apply, but where one still has an iterated logarithm bound. Nevertheless, one will normally try to apply Heyde's theorem first; his conditions are less cumbersome and his conclusions considerably stronger than ours.

THEOREM. *Let $X_n, Y_n, \mathcal{F}_n, v_n^2$ and B_n^2 be as above. Assume that there exist constants $0 < \eta_1 < 1, \eta_2 > 0$, a sequence of $\{\mathcal{F}_n\}$ stopping times T_n which increase to ∞ w.p.1, a sequence of random variables a_n and a set $\Omega_0 \in \vee \mathcal{F}_n$ such that the following conditions (1.4)–(1.8) hold a.e. on Ω_0 :*

$$(1.4) \quad B_1^2(\omega) < \infty,$$

$$(1.5) \quad Y_n(\omega) = o(B_n(\omega) \{\log \log B_n^2(\omega)\}^{-\frac{1}{2}}),$$

$$(1.6) \quad a_n \text{ is } \mathcal{F}_{T_n} \text{ measurable, } a_n \geq 0 \text{ and } \liminf_{n \rightarrow \infty} \frac{1}{n} \log \frac{1}{a_n} > 0,$$

$$(1.7) \quad a_n^2(\omega) \leq B_{T_n}^2(\omega) \leq a_n^2(\omega) \exp(n^{\eta_1}) \text{ eventually}$$

and

$$(1.8) \quad B_{T_{n+1}}^2(\omega) \geq n^{-\eta_2} B_{T_n}^2(\omega) \text{ eventually.}$$

Then $X_\infty = \lim X_n$ exists a.e. on Ω_0 and^{*}

^{*} We interpret the left hand side of (1.9) as zero when both numerator and denominator vanish for large n . Since $Y_l = 0$ for all $l \geq n$ a.e. on the set $\{B_n = 0\}$, (1.9) is trivial on the set $\Omega_1 = \{\omega : B_n(\omega) = 0 \text{ for some } n\}$. Thus we can remove Ω_1 from Ω_0 and in the proof in section 2 we tacitly assume $\Omega_0 \cap \Omega_1 = \emptyset$.

$$(1.9) \quad \limsup_{n \rightarrow \infty} \frac{|X_\infty - X_{n-1}|}{B_n \{\log \log B_n^{-2}\}^{\frac{1}{2}}} = \limsup_{n \rightarrow \infty} \frac{\left| \sum_n Y_l \right|}{B_n \{\log \log B_n^{-2}\}^{\frac{1}{2}}} \\ \leq \sqrt{2(\eta_1 + \eta_2 + 1)} \text{ a.e. on } \Omega_0.$$

COROLLARY. If (1.4)–(1.8) hold a.e. on Ω_0 for all $\eta_1, \eta_2 > 0$, then

$$(1.10) \quad \limsup_{n \rightarrow \infty} \frac{|X_\infty - X_{n-1}|}{B_n \{\log \log B_n^{-2}\}^{\frac{1}{2}}} \leq \sqrt{2}$$

a.e. on Ω_0 .

SPECIAL CASE. Assume that for some sequence of stopping times $T_n \uparrow \infty$ and constants $b_n > 0$, $K > 0$ we have

$$(1.11) \quad P \left\{ b_n^2 \leq B_{T_n}^2 - B_{T_{n+1}}^2 = \sum_{T_n \leq l < T_{n+1}} v_l^2 \leq K b_n^2 \right\} = 1,$$

and as $n \rightarrow \infty$

$$(1.12) \quad \max_{T_n < l \leq T_{n+1}} |Y_l| = o \left\{ \sum_{n+1}^\infty b_l^2 \right\}^{\frac{1}{2}} \left\{ \log \log \left(\sum_{n+1}^\infty b_l^2 \right)^{-1} \right\}^{-\frac{1}{2}} \text{ w.p.1.}$$

If in addition for all $\eta > 0$

$$(1.13) \quad \sum_1^\infty b_n^2 < \infty, \quad \liminf \frac{1}{n} \log \left\{ \sum_n^\infty b_l^2 \right\}^{-1} > 0,$$

$$\text{and } \sum_{n+1}^\infty b_l^2 \geq n^{-\eta} \sum_n^\infty b_l^2 \text{ eventually,}$$

then (1.10) holds w.p.1. (Merely take $a_n^2 = \sum_n^\infty b_l^2$.)

This case often applies with $K = 1$ and some b_n^2 if v_n^2 is non-random. In the case where the Y_n are independent Chow and Teicher [3, theorem 2] have proven (1.10) with equality, even without the condition (1.12). Heyde [6] also obtains equality in (1.10), but we already pointed out that his conditions imply ours after a truncation with the exception of (1.5). Indeed, in [6] theorem 1(b)

$$s_n^2 = E \left\{ \sum_{l=n}^\infty Y_l^2 \right\} < \infty,$$

and b(iii) in [6] implies that for each $\varepsilon > 0$

$$(1.14) \quad |Y_l| < \varepsilon s_l \text{ eventually and}$$

$$\frac{1}{s_n} \sum_{l=n}^\infty E \{ Y_l I[|Y_l| < \varepsilon s_l \mid \mathcal{F}_{l-1}] \} \rightarrow 0 \text{ w.p.1}$$

(note that we call X_n what [6] calls Y_n and vice versa). Under the conditions of [6] we can now take

$$a_n = 2^{-n}, \quad T_n = \sup\{k : s_k^2 \geq 2^{-n}\}.$$

Then for

$$\tilde{Y}_l(\varepsilon) = Y_l I[|Y_l| < \varepsilon s_l] - E\{Y_l I[|Y_l| < \varepsilon s_l] \mid \mathcal{F}_{l-1}\}$$

one has

$$(1.15) \quad \begin{aligned} & P \left\{ \sum_{T_n \leq l < T_{n+1}} [\tilde{Y}_l^2(\varepsilon) - E\{\tilde{Y}_l^2(\varepsilon) \mid \mathcal{F}_{l-1}\}] \geq \varepsilon' 2^{-n} \right\} \\ & \leq (\varepsilon')^{-2} 2^{2n} \sum_{T_n \leq l < T_{n+1}} E\{\tilde{Y}_l^4(\varepsilon)\}. \end{aligned}$$

By b(iv) of [6] the sum of (1.15) over n converges. This together with b(i) of [6] in the a.s. version and (1.14) imply

$$2^{n+1} \sum_{T_n \leq l < T_{n+1}} E\{\tilde{Y}_l^2(\varepsilon) \mid \mathcal{F}_{l-1}\} \rightarrow \eta \quad \text{w.p.1}$$

(η as in [6]). Thus, for any sequence $\delta_n \downarrow 0$

$$\delta_n 2^{-n-1} \leq \sum_{T_n \leq l < T_{n+1}} E\{\tilde{Y}_l^2(\varepsilon) \mid \mathcal{F}_{l-1}\} \leq \delta_n^{-1} 2^{-n-1}$$

eventually.

Thus under the conditions of theorem 1(b) in [6] a simple variant of (1.11) and (1.13) apply to the $\tilde{Y}_l(\varepsilon)$.

EXAMPLES. (i) To illustrate the relation with [6] further we consider the Polya urn example of [6], sect. 3. At stage n there are b_n (r_n) black (red) balls in an urn. A ball is drawn at random and replaced and then a random number c_n of balls of the color drawn are added to the urn. If \mathcal{F}_n is an increasing sequence of σ -fields such that b_n , r_n and c_n are \mathcal{F}_n measurable, then $X_n = b_n/(b_n + r_n)$ is a martingale. Heyde considers only the case where $c_n = c$, a constant. It is not hard to vary c_n in such a way that b(i) of [6] fails. In this example

$$|Y_n| \leq K_n, \quad \text{and}$$

$$v_n^2 = X_{n-1}(1 - X_{n-1})K_n^2 \sim X_\infty(1 - X_\infty)K_n^2,$$

where

$$K_n = \frac{c_{n-1}}{b_{n-1} + r_{n-1} + c_{n-1}}$$

(see [6], formula (22)). If $1 \leq c_n \leq C$ w.p.1, then

$$\frac{1}{b_0 + c_0 + nC} \leq K_n \leq \frac{C}{b_0 + c_0 + n}.$$

On the set

$$(1.16) \quad 0 < X_\infty < 1$$

one has eventually for any $\varepsilon > 0$

$$\frac{1}{2} X_\infty (1 - X_\infty) C^{-2} n^{-1} \leq B_n^2 \leq (1 + \varepsilon) X_\infty (1 - X_\infty) C^2 n^{-1}.$$

Taking $T_n = 2^n$, $a_n^2 = n^{-1} 2^{-n}$ we now easily obtain

$$\limsup_{n \rightarrow \infty} (n / \log \log n)^{1/2} |X_n - X_\infty| \leq C(2X_\infty(1 - X_\infty))^{1/2}$$

a.e. on the set (1.16). Note that

$$E\{(X_\infty - X_{n-1})^2 \mid \mathcal{F}_{n-1}\} = E\{B_n^2 \mid \mathcal{F}_{n-1}\} \leq 2C^2 n^{-1}$$

from which one easily deduces that (1.16) always has a positive probability when $b_0 > 0$, $r_0 > 0$. With some more work one actually can show that (1.16) holds w.p.1.

(ii) Another example is obtained by taking $X_n = W(\tau_n)$, where $W(\cdot)$ is a Wiener process as above and τ_n an increasing sequence of stopping times for $W(\cdot)$. Assume that

$$(1.17) \quad c_n^2 \leq \tau_n - \tau_{n-1} \leq K' c_n^2 \quad \text{w.p.1}$$

for some constants K' , $c_n > 0$ satisfying

$$(1.18) \quad \sum_1^\infty c_n^2 < \infty,$$

$$c_n^2 \log c_n^{-1} = o \left\{ \sum_{i=n}^\infty c_i^2 / \left(\log \log \left(\sum_{i=n}^\infty c_i^2 \right)^{-1} \right) \right\}.$$

Then

$$\limsup_{n \rightarrow \infty} \frac{|W(\tau) - W(\tau_n)|}{\{(\tau - \tau_n) \log \log (\tau - \tau_n)^{-1}\}^{1/2}} \leq \sqrt{2K'} \quad \text{w.p.1.}$$

Indeed $c_n^2 \leq v_n^2 \leq K' c_n^2$ by (1.17) and since

$$\sum_{l=n+1}^\infty c_l^2 \sim \sum_{l=n}^\infty c_l^2$$

we can take

$$T_n = \sup \left\{ k : \sum_k c_i^2 \geq (2K')^{-2n} \right\}.$$

Then

$$(2K')^{-2n} \leq B_{T_n}^2 \leq (2K')^{-2n+1}$$

eventually, so that (1.11) holds with $b_n^2 = (2K')^{-2n}(1 - 1/2K')$, $K = 2K'/(2K' - 1)$. Also (1.12) holds, since by the uniform Hölder condition for the Wiener process

$$\begin{aligned} |Y_i| &= |W(\tau_i) - W(\tau_{i-1})| = O\{(\tau_i - \tau_{i-1}) \log(\tau_i - \tau_{i-1})^{-1}\}^{1/2} \\ &= O\{c_i (\log c_i^{-1})^{1/2}\}. \end{aligned}$$

This result gains in interest when it is compared with the results of Orey and Taylor in [10]: there is w.p.1, an everywhere dense set $S \subset [0, \infty)$ of Hausdorff dimension one, such that for all $s \in S$,

$$\limsup_{t \uparrow s} \frac{|W(s) - W(t)|}{\{|s - t| \log \log |s - t|^{-1}\}^{1/2}} = \infty.$$

Apparently if (1.17), (1.18) hold, either $\tau \notin S$ or if $\tau \in S$, τ_n misses the points where $|W(\tau) - W(\tau_n)|$ is much bigger than $\{(\tau - \tau_n) \log \log (\tau - \tau_n)^{-1}\}^{1/2}$.

COUNTEREXAMPLES. The following examples show some of the inherent limitations on weakening the hypotheses or proving a converse of our theorem.

(i) *This example shows that (1.9) is essentially sharp.* Let $0 < \beta < 2\gamma$, $\sigma_k^2 \downarrow 0$ and L_k be constants such that L_k is an integer and

$$(1.19) \quad \sigma_k^2 L_k^2 = (k!)^{-\gamma}, \quad k^{-\gamma-1} L_k \rightarrow 1.$$

Now take $X_0 = T_0 = 0$ and define Y_l and T_k such that $T_k - 1$ is a stopping time as follows: On the set $\{T_k - 1 \leq l < T_{k+1} - 1\} \in \mathcal{F}_l$, the conditional distribution of Y_{l+1} , given \mathcal{F}_l will be normal with mean zero and variance σ_k^2 . Moreover,

$$\begin{aligned} (1.20) \quad T_{k+1} - 1 &= \inf \left\{ n \geq T_k + L_k k^{1/2} (\log k)^{1/2(\beta+3)} : \sum_{n=L_k+1}^n Y_l \right. \\ &\quad \left. \geq \{(\beta+2)L_k \sigma_k^2 \log \log (L_k \sigma_k^2)^{-1}\}^{1/2} \right\}, \end{aligned}$$

provided there exists an $n \leq T_k + 2L_k k^{1/2} (\log k)^{1/2(\beta+3)}$ with this property. Otherwise take

$$(1.21) \quad T_{k+1} = T_k + [2L_k k^{1/2} (\log k)^{1/2(\beta+3)}] + 1.$$

In other words, after $T_k - 1$ we add a large number of independent normal

random variables with mean zero and variance σ_k^2 . We try to add just so many of these random variables that the sum of the last L_k of them is exceptionally large.

By definition $v_l^2 = \sigma_k^2$ for $T_k \leq l < T_{k+1}$ so that

$$(1.22) \quad B_{T_k}^2 = \sum_{j=0}^{\infty} (T_{k+j+1} - T_{k+j}) \sigma_{k+j}^2.$$

Also, by definition

$$(1.23) \quad L_k k^{\frac{1}{2}\beta} (\log k)^{\frac{1}{2}(\beta+3)} \leq T_{k+1} - T_k \leq 2L_k k^{\frac{1}{2}\beta} (\log k)^{\frac{1}{2}(\beta+3)} + 1.$$

(1.22), (1.23) and (1.19) show that

$$\begin{aligned} k^{\frac{1}{2}\beta-\gamma-1} (\log k)^{\frac{1}{2}(\beta+3)} (k!)^{-\gamma} &\sim L_k k^{\frac{1}{2}\beta} (\log k)^{\frac{1}{2}(\beta+3)} \sigma_k^2 \leq B_{T_k}^2 \\ &\leq 3k^{\frac{1}{2}\beta-\gamma-1} (\log k)^{\frac{1}{2}(\beta+3)} (k!)^{-\gamma} \end{aligned}$$

from which (1.4), (1.6)–(1.8) with any $\eta_1 > 0$, $\eta_2 > \gamma$ and

$$a_n^2 = \frac{1}{2} n^{\frac{1}{2}\beta-\gamma-1} (\log n)^{\frac{1}{2}(\beta+3)} (n!)^{-\gamma}$$

immediately follow. (1.5) is also easy and thus, by (1.9)

$$(1.24) \quad \limsup_{n \rightarrow \infty} \frac{|X_n - X_{n-1}|}{B_n \{\log \log B_n^{-2}\}^{\frac{1}{2}}} \leq \sqrt{2\gamma + 2}.$$

On the other hand, simple probability estimates show

$$(1.25) \quad P\{X_{T_{n+1}-1} - X_{T_n+1-L_n-1} \geq \{(\beta+2)L_n \sigma_n^2 \log \log (L_n \sigma_n^2)^{-1}\}^{\frac{1}{2}} \mid \mathcal{F}_{T_n-1}\} \geq K n^{-1}$$

for some constant $K = K(\beta, \gamma) > 0$. From (1.25), (1.24), (1.19) and the generalized Borel–Cantelli lemma ([9], corol. VII.2.6) it can be shown that

$$\begin{aligned} (1.26) \quad \limsup_{n \rightarrow \infty} \frac{|X_n - X_{n-1}|}{B_n \{\log \log B_n^{-2}\}^{\frac{1}{2}}} &\leq \limsup_{n \rightarrow \infty} \frac{|X_{T_{n+1}-1} - X_{T_n+1-L_n-1}|}{\{L_n \sigma_n^2 \log \log (L_n \sigma_n^2)^{-1}\}^{\frac{1}{2}}} \\ &\leq \sqrt{\beta + 2} \quad \text{w.p.1.} \end{aligned}$$

Since β can be chosen as close to 2γ as desired, (1.24) and (1.26) substantiate our claim.

(ii) *Even if (1.4)–(1.8) hold for any $\eta_1, \eta_2 > 0$ for almost all ω we may have*

$$(1.27) \quad \limsup_{n \rightarrow \infty} \frac{|X_n - X_{n-1}|}{B_n} \leq C_0 \quad \text{w.p.1,}$$

where C_0 is a certain finite constant. We shall not give the lengthy construction of such an example here. It relies on Kahane's solution [8] of Dvoretzky's

conjecture ([4]) about times where the law of the iterated logarithm fails for a Brownian motion $W(\cdot)$. Kahane shows that there exists a constant $C_1 < \infty$ and w.p.1 an everywhere dense set S such that for all $s \in S$

$$(1.28) \quad \limsup_{t \uparrow s} |s - t|^{-\frac{1}{2}} |W(t) - W(s)| \leq C_1.$$

(Kahane even proves (1.28) for the \limsup as $t \rightarrow s$, irrespective of $t < s$ or $t > s$.) Kahane does not construct any s satisfying (1.28) as an (accessible) stopping time. To obtain an example satisfying (1.4)–(1.8) for all $\eta_1, \eta_2 > 0$, as well as (1.27), we constructed a sequence of stopping times τ_n for $W(\cdot)$ such that τ_n strictly increases to a limit s which satisfies (1.28), and such that for all $\eta > 0$ one has eventually

$$s - \tau_{n+1} \geq n^{-\eta} (s - \tau_n).$$

Thus, in general, it is not necessary that

$$(1.29) \quad \limsup_{n \rightarrow \infty} \frac{|X_\infty - X_{n-1}|}{B_n f(n)} > 0$$

for any sequence of normalizing constants $f(n) \uparrow \infty$. We have no simple modification of (1.4)–(1.8) which guarantees (1.29). Note, however, that [6] has a conclusion which is stronger than (1.29).

2. Proof of theorem

It is known (see [9], proposition VII. 2.3 c) that (1.4) implies the convergence to a finite limit of X_n a.e. on Ω_0 . Thus we only have to prove (1.9). For typographical convenience we shall often write $B(S)$ instead of B_S . Fix $\theta > 1$ and $0 < \varepsilon < 1$ and introduce the following stopping times^{*}:

$$(2.1) \quad S(n, r, 0) = T_n,$$

$$S(n, r, k+1) = T_{n+1} \wedge \inf \left\{ t > S(n, r, k) : \sum_{l=S(n, r, k)}^t v_l^2 \geq \varepsilon a_n^2 \theta^r n^{-\eta_2} \right\},$$

$$r = 0, 1, \dots, [(\log \theta)^{-1} n^{\eta_1}], \quad k = 0, 1, \dots, 2[\theta \varepsilon^{-1} n^{\eta_2}] + 2.$$

Set $\eta = \eta_1 + \eta_2 + 1$ and assume that

$$(2.2) \quad \sum_m^\infty Y_l(\omega) \geq \{\sqrt{2\eta}(1-\varepsilon)^{-1} + 4\sqrt{\varepsilon\eta\theta}\} B_m(\omega) \left\{ \log \log \frac{1}{B_m^2(\omega)} \right\}^{\frac{1}{2}}$$

^{*} $a \wedge b$ denotes the minimum of a and b .

for some $\omega \in \Omega_0$ and

$$(2.3) \quad T_n \leq m < T_{n+1}.$$

Assume further that for this ω and n (1.4) as well as the inequalities in (1.7) and (1.8) hold. Then

$$(2.4) \quad B_{T_{n+1}}^2(\omega) \geq n^{-\eta_2} B_{T_n}^2(\omega),$$

and there exists an $0 \leq r \leq \{\log \theta\}^{-1} n^{\eta_1}$ for which

$$(2.5) \quad a_n^2 \theta^r \leq B^2(T_n) < a_n^2 \theta^{r+1}.$$

Moreover, if $S(n, r, j+2) < T_{n+1}$, then

$$\begin{aligned} \sum_{S(n, r, j) \leq l < S(n, r, j+2)} v_l^2 &\geq \sum_{S(n, r, j) \leq l \leq S(n, r, j+1)} v_l^2 \\ &\geq \varepsilon a_n^2 \theta^r n^{-\eta_2}, \end{aligned}$$

so that either

$$(2.6) \quad S(n, r, 2j) \geq T_{n+1}$$

or

$$(2.7) \quad a_n^2 \theta^{r+1} > B^2(T_n) \geq \sum_{T_n \leq l \leq S(n, r, 2j)} v_l^2 \geq j \varepsilon a_n^2 \theta^r n^{-\eta_2}.$$

Clearly (2.7) is impossible for $j > \theta \varepsilon^{-1} n^{\eta_2}$ and (2.6) must hold for any such j . Thus, by (2.3) there must exist a $k \leq 2\theta \varepsilon^{-1} n^{\eta_2} + 2$ with

$$(2.8) \quad S(n, r, k) \leq m < S(n, r, k+1).$$

Of course (2.3) and (2.8) together imply $T_n \leq S(n, r, k) < T_{n+1}$ and hence, by (2.4) and (2.5)

$$(2.9) \quad a_n^2 \theta^{r+1} > B^2(T_n) \geq B^2(S(n, r, k)) \geq B^2(T_{n+1}) \geq B^2(T_n) n^{-\eta_2} \geq a_n^2 \theta^r n^{-\eta_2}.$$

Consequently, there must exist an integer

$$s \in [0, (\log \theta)^{-1} \eta_2 \log n] + 2$$

for which

$$(2.10) \quad a_n^2 \theta^{r+s} n^{-\eta_2} \leq B^2(S(n, r, k)) < a_n^2 \theta^{r+s+1} n^{-\eta_2}.$$

Now, by virtue of (1.5), there exists a (deterministic) sequence $\delta_v \downarrow 0$ such that a.e. on Ω_0

$$(2.11) \quad |Y_l(\omega)| \leq \delta_\nu B_l(\omega) \left\{ \log \log \frac{1}{B_l^2(\omega)} \right\}^{-\frac{1}{2}} \quad \text{for all } l \geq T_\nu$$

and ν sufficiently large. Let δ_ν be fixed in this way and define for each fixed n, r, s^*

$$u(x) = x^{\frac{1}{2}} \left\{ \log \log \frac{1}{x} \right\}^{-\frac{1}{2}}, \quad 0 < x < 1,$$

$$Y_l^+ = Y_l^+(\omega, n, r, s) = Y_l(\omega) I[Y_l(\omega) \leq \delta_\nu u(a_n^2 \theta^{r+s+1} n^{-\eta_2})],$$

$$Y_l^- = Y_l^-(\omega, n, r, s) = Y_l(\omega) I[Y_l(\omega) \geq -\delta_\nu u(a_n^2 \theta^{r+s+1} n^{-\eta_2})].$$

Then for almost all $\omega \in \Omega_0$ for which (2.2), (2.3), (2.8) and (2.10) hold and large n , at least one of the events (2.12)–(2.14) must occur:

$$(2.12) \quad E(n) = \{\omega \in \Omega_0 : (2.11) \text{ fails for } \nu = n\},$$

There exists a $S(n, r, k) < t < S(n, r, k+1)$ with

$$(2.13)$$

$$\sum_{S(n, r, k) < l < t} Y_l^- \leq -2\{\varepsilon \eta a_n^2 \theta^{r+s+1} n^{-\eta_2} \log n\}^{\frac{1}{2}},$$

$$\sum_{S(n, r, k)+1}^{\infty} Y_l^+ = \sum_{S(n, r, k) < l < m} Y_l^- + \sum_m^{\infty} Y_l - Y_m I[m = S(n, r, k)]$$

$$(2.14) \quad \geq \{\sqrt{2\eta}(1-\varepsilon)^{-1} + 4\sqrt{\varepsilon\eta\theta} - \delta_n\} B_m \left\{ \log \log \frac{1}{B_m^2} \right\}^{\frac{1}{2}} \\ - 2\{\varepsilon \eta a_n^2 \theta^{r+s+1} n^{-\eta_2} \log n\}^{\frac{1}{2}}.$$

Note now that for almost all $\omega \in \Omega_0^{**}$

$$(2.15) \quad \lim_{n \rightarrow \infty} (\log n)^{-1} \log \log B_{T_n}^{-2} = 1.$$

Indeed, by (1.6) and (1.7) for large n

$$\log \log B_{T_n}^{-2} \geq \log \log a_n^{-2} \exp - n^{\eta_1} \geq \log n + O(1),$$

and if (1.7) and (1.8) hold for $n \geq n_0(\omega)$, then

$$B_{T_n}^2 \geq (n-1)^{-\eta_2} B_{T_{n-1}}^2 \geq \cdots \geq \prod_{l=n_0}^{n-1} l^{-\eta_2} B_{n_0}^2(\omega) \geq \{(n-1)!\}^{-\eta_2} \{(n_0-1)!\}^{\eta_2} a_{n_0}^2$$

* $I[\]$ denotes the indicator of the event between square brackets.

** As explained in an earlier footnote we assume without loss of generality that $\Omega_0 \cap \Omega_1 = \emptyset$.

and

$$\limsup (\log n)^{-1} \log \log B_{T_n}^{-2} \leq \lim (\log n)^{-1} \log \log \{(n-1)!\}^{\eta_2} = 1.$$

Thus, for large enough n we will have

$$(1 - \varepsilon) \log n \leq \log \log B_{T_n}^{-2} \leq \log \log B_{T_{n+1}}^{-2} \leq (1 + \varepsilon) \log n$$

and also

$$(2.16) \quad (1 - \varepsilon) \log n \leq \log \log B^{-2}(S(n, r, k)) \leq \log \log B_m^{-2} \leq (1 + \varepsilon) \log n.$$

For later use we point out that (2.15) and (1.7) imply that also a.e. on Ω_0

$$(2.17) \quad \lim_{n \rightarrow \infty} (\log n)^{-1} \log \log a_n^{-2} = 1.$$

Finally, we introduce

$$A_i^2(\omega) = A_i^2(\omega, n, r, k) = \sum_{S(n, r, k)+1}^i v_i^2(\omega),$$

and observe that if

$$(2.18) \quad S(n, r, k) < t < S(n, r, k+1),$$

then, by virtue of (2.1),

$$(2.19) \quad A_i^2(\omega, n, r, k) < \varepsilon a_n^2 \theta^{r+s+1} n^{-\eta_2}.$$

It follows that the event (2.13) is contained in the event

$$\begin{aligned} \Gamma(n, r, k, s) = & \left\{ \omega : \sum_{S(n, r, k) < t \leq i} Y_i^-(\omega, n, r, s) \right. \\ & \leq -\{\varepsilon \eta a_n^2 \theta^{r+s+1} n^{-\eta_2} \log n\}^{\frac{1}{2}} - \{\varepsilon a_n^2 \theta^{r+s+1} n^{-\eta_2}\}^{-\frac{1}{2}} (\eta \log n)^{\frac{1}{2}} A_i^2(\omega) \\ & \left. \text{for some } t > S(n, r, k) \right\} \end{aligned}$$

Similarly (2.8)–(2.10) and (2.1) imply

$$B_m^2 \geq B^2(S(n, r, k)) - \varepsilon a_n^2 \theta^{r+s+1} n^{-\eta_2} \geq (1 - \varepsilon) B^2(S(n, r, k)) \geq (1 - \varepsilon) a_n^2 \theta^{r+s+1} n^{-\eta_2}$$

and (2.10) implies

$$A_i^2(\omega, n, r, k) \leq B^2(S(n, r, k)) \leq a_n^2 \theta^{r+s+1} n^{-\eta_2}.$$

Therefore, if (2.8)–(2.10) and (2.16) hold, the event (2.14) is contained in

$$\begin{aligned} \Delta(n, r, k, s) = & \left\{ \omega : \sum_{S(n, r, k)+1}^t Y_i^+(\omega, n, r, s) \geq \{\sqrt{2\eta} + \sqrt{\varepsilon\eta}\} B(S(n, r, k)) (\log n)^{\frac{1}{2}} \right. \\ & \geq \frac{1}{2} (\sqrt{2\eta} + \sqrt{\varepsilon\eta}) \{a_n^2 \theta^{r+s} n^{-\eta_2} \log n\}^{\frac{1}{2}} \\ & \quad \left. + \frac{1}{2} (\sqrt{2\eta} + \sqrt{\varepsilon\eta}) \{a_n^2 \theta^{r+s+1} n^{-\eta_2}\}^{-\frac{1}{2}} (\log n)^{\frac{1}{2}} A_t^2(\omega) \right. \\ & \quad \left. \text{for some } t > S(n, r, k) \right\}. \end{aligned}$$

It follows from the above arguments that up to a null set

$$\begin{aligned} & \{\omega \in \Omega_0 : (2.2) \text{ occurs for infinitely many } m\} \\ (2.20) \quad & \subset \{E(n) \text{ i.o.}\} \cup \left\{ \bigcup_{r, k, s} \Gamma(n, r, k, s) \text{ occurs for infinitely many } n \right\} \\ & \cup \left\{ \bigcup_{r, k, s} \Delta(n, r, k, s) \text{ occurs for infinitely many } n \right\}. \end{aligned}$$

To conclude the proof we show that

$$(2.21) \quad \sum_n \sum_{r, k, s} (P\{\Gamma(n, r, k, s)\} + P\{\Delta(n, r, k, s)\}) < \infty$$

a.e. on Ω_0 , whenever

$$(2.22) \quad \frac{1}{2} (\sqrt{2\eta} + \sqrt{\varepsilon\eta})^2 (1 + \varepsilon)^{-1} \theta^{-\frac{1}{2}} > \eta.$$

Since $P\{E(n) \text{ i.o.}\} = 0$ by the choice of δ_ν (see (2.11)), (2.20), (2.21) and the Borel-Cantelli lemma imply

$$\limsup_{m \rightarrow \infty} B_m^{-1} \left\{ \log \log \frac{1}{B_m^2} \right\}^{-\frac{1}{2}} \sum_m Y_l \leq \sqrt{2\eta} (1 - \varepsilon)^{-1} + 4\sqrt{\varepsilon\eta\theta}$$

whenever (2.22) holds. Replacing Y_l by $-Y_l$ yields the same inequality for $-\sum Y_l$, so that (1.9) follows when we let $\theta \downarrow 1$, $\varepsilon \downarrow 0$. It remains to prove (2.21). We shall restrict ourselves to estimating $P\{\Delta(n, r, k, s)\}$, the estimate for $P\{\Gamma(n, r, k, s)\}$ being almost the same. For fixed n, r, k, s ,

$$E\{Y_l^+ \mid \mathcal{F}_{l-1}\} \leq E\{Y_l \mid \mathcal{F}_{l-1}\} = 0 \quad \text{and}$$

$$E\{(Y_l^+)^2 \mid \mathcal{F}_{l-1}\} \leq v_l^2.$$

Thus, $\sum_{i=1}^t Y_{S(n, r, k)+i}^+$, $t = 0, 1, \dots$, is a supermartingale with respect to the σ -fields $\mathcal{G}_t = \mathcal{F}_{S(n, r, k)+t}$, and if we write Z_t for $Y_{S(n, r, k)+t}^+$, then

$$E\{Z_t^2 \mid \mathcal{G}_{t-1}\} \leq v_{S(n, r, k)+t}^2.$$

Moreover

$$Z_i \leq c \equiv \delta_n \{a_n^2 \theta^{r+s+1} n^{-\eta_2}\}^{\frac{1}{2}} \{(1-\varepsilon) \log n\}^{-\frac{1}{2}}$$

as soon as

$$(2.23) \quad \log \log \{a_n^2 \theta^{r+s+1} n^{-\eta_2}\}^{-1} \geq (1-\varepsilon) \log n$$

(see the definition of Y_i^+). Now set

$$a = \frac{1}{2}(\sqrt{2\eta} + \sqrt{\varepsilon\eta}) \{a_n^2 \theta^{r+s} n^{-\eta_2} \log n\}^{\frac{1}{2}},$$

$$\lambda = (\sqrt{2\eta} + \sqrt{\varepsilon\eta}) (1+\varepsilon)^{-1} \{a_n^2 \theta^{r+s+1} n^{-\eta_2}\}^{-\frac{1}{2}} (\log n)^{\frac{1}{2}}$$

and

$$\Phi_c(\lambda) = c^{-2} \{\exp(\lambda c) - 1 - \lambda c\}.$$

Then

$$(2.24) \quad \begin{aligned} P\{\Delta(n, r, k, s)\} &\leq P\left\{\sum_{s(n, r, k)+1}^i Y_i^+ \geq a + \frac{\Phi_c(\lambda)}{\lambda} A_i^2 \text{ for some } i > S(n, r, k)\right\} \\ &\leq P\left\{\bigcup_{i \geq 1} \left\{\sum_{i=1}^i Z_i \geq a + \frac{\Phi_c(\lambda)}{\lambda} \sum_{i=1}^i E\{Z_i^2 | \mathcal{G}_{i-1}\}\right\}\right\} \end{aligned}$$

as soon as

$$(2.25) \quad \lambda^{-1} \Phi_c(\lambda) \leq \frac{1}{2}(1+\varepsilon)\lambda.$$

(2.23) holds for large n by virtue of (2.17) and the restrictions on r, s , while (2.25) holds for large n because $\lambda c \rightarrow 0$. Finally by [9], pp. 154–155 or [12] pp. 299–302, the last member of (2.24) is bounded by

$$\exp - \lambda a = \exp - \frac{1}{2}(\sqrt{2\eta} + \sqrt{\varepsilon\eta})^2 (1+\varepsilon)^{-1} \theta^{-\frac{1}{2}} \log n.$$

Since the triple (r, k, s) runs through at most

$$\{(\log \theta)^{-1} n^{\eta_1} + 1\} \cdot \{2\theta \varepsilon^{-1} n^{\eta_2} + 3\} \cdot \{(\log \theta)^{-1} \eta_2 \log n + 3\}$$

values, we see that for almost all $\omega \in \Omega_0$

$$\sum_{r, k, s} P\{\Delta(n, r, k, s)\} = O(n^{\eta_1 + \eta_2} \log n \exp - \frac{1}{2}(\sqrt{2\eta} + \sqrt{\varepsilon\eta})^2 (1+\varepsilon)^{-1} \theta^{-\frac{1}{2}} \log n).$$

Thus, if (2.22) holds we have indeed

$$\sum_n \sum_{r, k, s} P\{\Delta(n, r, k, s)\} < \infty.$$

□

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We were led to this investigation by the discussion in Section 1.3 and the proof of proposition 3 in [5]. Our theorem can be used to give a simple alternative proof of proposition 3 in [5] (essentially the same observation is made in [7]). We are grateful to Professor Williams for giving us a copy of [5] before publication.

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