

ON THE LAW OF THE ITERATED LOGARITHM FOR INFINITE DIMENSIONAL ORNSTEIN-UHLENBECK PROCESSES

QI-MAN SHAO

ABSTRACT Let $\{X_k(t), -\infty < t < \infty\}_{k=1}^{\infty}$ be independent Ornstein-Uhlenbeck processes and $X(t, n) = \sum_{i=1}^n X_i(t)$. In this paper the law of iterated logarithm for $X(t, n)$ is considered. The results obtained improve those of Csörgő and Lin(1988) and Schmuland(1987).

A real valued stationary Gaussian process $\{X(t), -\infty < t < \infty\}$ will be called an *Ornstein-Uhlenbeck process* with coefficients γ and λ ($\gamma \geq 0, \lambda > 0$) if $EX(t) = 0$ and $EX(s)X(t) = (\gamma/\lambda) \exp(-\lambda|t - s|)$. Let

$$\{Y(t), -\infty < t < \infty\} = \{X_k(t), -\infty < t < \infty\}_{k=1}^{\infty}$$

be a sequence of independent Ornstein-Uhlenbeck processes with coefficients γ_k and λ_k . The process $Y(\cdot)$ was first studied by Dawson(1972) as the stationary solution of the infinite array of stochastic differential equations:

$$dX_k(t) = -\lambda_k X_k(t) dt + (2\gamma_k)^{1/2} dW_k(t), \quad k = 1, 2, \dots,$$

where $\{W_k(t), -\infty < t < \infty\}_{k=1}^{\infty}$ are independent Wiener processes. The properties of $Y(\cdot)$ have been extensively studied in the literature. Since $EX_k^2(t) = \gamma_k/\lambda_k$, it is clear that for every fixed t , $Y(t)$ is almost surely in ℓ^2 if and only if $\sum_{k=1}^{\infty} \gamma_k/\lambda_k < \infty$. The continuity properties of $Y(\cdot)$ were investigated by Dawson(1972), Schmuland(1987), Iscoe and McDonald(1986), Fernique(1989), Csáki, Csörgő and Shao(1991). Csörgő and Lin(1988) studied $Y(\cdot)$ in terms of the path behaviour of the two-time parameter stochastic process $\{X(t, n), -\infty < t < \infty, n = 1, 2, \dots\}$, where $X(t, n) = \sum_{k=1}^n X_k(t)$, $X(t, 0) = 0$ for all $t \in R$ and established P. Lévy type moduli of continuity, large increment rates for the latter process and the following law of the iterated logarithm:

THEOREM A. Let $\lambda_N^* = \max_{1 \leq i \leq N} \lambda_i$, and $\sigma_N = \sigma(N) = \sum_{i=1}^N \gamma_i/\lambda_i$. Assume that

$$(1) \quad (\log \lambda_N^*) / \log \log N \longrightarrow 0, \text{ as } N \longrightarrow \infty,$$

and that the non-decreasing sequence $\{T_N\}$ satisfies

$$(2) \quad \log T_N / \log \log N \longrightarrow 0, \text{ as } N \longrightarrow \infty.$$

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Suppose also that for every $\epsilon > 0$ there exist $1 < \theta_1 < \theta_2$ such that

$$(3) \quad \limsup_{k \rightarrow \infty} \sigma(\theta_1^{k+1}) / \sigma(\theta_1^k) \leq 1 + \epsilon$$

and

$$(4) \quad \limsup_{k \rightarrow \infty} \sigma(\theta_2^k) / \sigma(\theta_2^{k+1}) \leq \epsilon.$$

Then, with $\beta_N^* = (2(\sum_{i=1}^N \gamma_i / \lambda_i) \log \log N)^{1/2}$, we have

$$\limsup_{N \rightarrow \infty} |X(T_N, N)| / \beta_N^* = \limsup_{N \rightarrow \infty} \max_{1 \leq n \leq N} \sup_{|t| \leq T_N} |X(t, n)| / \beta_N^* = 1 \text{ a.s.}$$

Schmuland(1987), using Dirichlet form-techniques, proved that if $\gamma_k / \lambda_k \equiv 1$ and $\sum_{i=1}^n \gamma_i / (2n \log \log n) \rightarrow 0$ as $n \rightarrow \infty$, then

$$(5) \quad P\left\{ \limsup_{n \rightarrow \infty} X(t, n) / (2n \log \log n)^{1/2} = 1 \text{ for all } t \in R \right\} = 1.$$

It is not difficult to see that (3) and (4), in fact, imply that there exists positive constants α_1, α_2, c_1 and c_2 such that

$$(6) \quad \sigma_n / n^{\alpha_1} \leq c_1 \sigma_m / m^{\alpha_1}$$

and

$$(7) \quad \sigma_n / n^{\alpha_2} \geq c_2 \sigma_m / m^{\alpha_2}$$

for each $1 \leq n \leq m$.

Unfortunately, conditions (1) and (2) in Theorem A are too restrictive to be satisfied even for $\lambda_k = k^\alpha$, or $\lambda_k = \log^\alpha(1 + k)$ ($\alpha > 0$), or $T_N = \log N$. The aim of this note is to relax the conditions of Theorem A and that of Schmuland(1987) as well.

Let $\{T_N, n \geq 1\}$ be a non-decreasing sequence of positive numbers. Put

$$\begin{aligned} \sigma_N &= \sigma(N) = \sum_{i=1}^N \gamma_i / \lambda_i, \quad \Gamma_N = \sum_{i=1}^N \gamma_i, \\ \beta_N &= \left(2\sigma_N (\log(\Gamma_N T_N / \sigma_N) + \log \log \sigma_N) \right)^{1/2}, \end{aligned}$$

where and in the sequel, $\log x = \ln(\max(x, e))$, \ln is the natural logarithm.

For $0 < \epsilon < 1$, define $\theta_n(\epsilon)$ as the solution of the equation

$$(8) \quad \sum_{i=1}^n \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i \theta_n(\epsilon)} = \epsilon \sigma_n.$$

THEOREM 1. *Assume that*

$$(9) \quad T_N \Gamma_N / \sigma_N + \sigma_N \rightarrow \infty, \text{ as } N \rightarrow \infty.$$

Then, we have

$$(10) \quad \limsup_{N \rightarrow \infty} \max_{1 \leq n \leq N} \sup_{|t| \leq T_N} |X(t, n)| / \beta_N \leq 1 \text{ a.s.}$$

THEOREM 2. Assume that (9) is satisfied and that there exists a positive constant C such that

$$(11) \quad \sigma_N \leq C\sigma_{N-1} \text{ for every } N \geq 1,$$

$$(12) \quad \log \frac{T_N \Gamma_N}{\sigma_N} \leq (1 + C\epsilon) \log \frac{T_N}{\theta_N(\epsilon)} + C\epsilon \log \log \sigma_N,$$

for every $0 < \epsilon < 1$ as $N \rightarrow \infty$. Then, we have

$$(13) \quad \limsup_{N \rightarrow \infty} \sup_{0 \leq t \leq T_N} |X(t, N)| / \beta_N = 1 \text{ a.s.}$$

$$(14) \quad \limsup_{N \rightarrow \infty} \max_{1 \leq n \leq N} \sup_{|t| \leq T_N} |X(t, n)| / \beta_N = 1 \text{ a.s.}$$

If, in addition, we also have

$$(15) \quad \log \log \sigma_N = o\left(\log \frac{T_N \Gamma_N}{\sigma_N}\right), \text{ as } N \rightarrow \infty.$$

Then

$$(16) \quad \lim_{N \rightarrow \infty} \sup_{0 \leq t \leq T_N} |X(t, N)| / \beta_N = 1 \text{ a.s.}$$

$$(17) \quad \lim_{N \rightarrow \infty} \max_{1 \leq n \leq N} \sup_{|t| \leq T_N} |X(t, n)| / \beta_N = 1 \text{ a.s.}$$

THEOREM 3. Assume that (11) is satisfied. Moreover, suppose that

$$(18) \quad \log(\Gamma_N / \sigma_N) = o(\log \log \sigma_N) \text{ as } N \rightarrow \infty,$$

and

$$(19) \quad \sigma_N \rightarrow \infty \text{ as } N \rightarrow \infty.$$

Then, we have

$$(20) \quad P\left\{\limsup_{N \rightarrow \infty} X(t, N) / (2\sigma_N \log \log \sigma_N)^{1/2} = 1 \text{ for all } t \in \mathcal{R}\right\} = 1.$$

Before stating our corollaries, we introduce the following notations:

$$\lambda_N^* = \max_{i \leq N} \lambda_i, \quad m_n(1, \epsilon) = \max \left\{ \ell : \sum_{i=\ell}^n \frac{\gamma_i}{\lambda_i} \geq (1 - \epsilon)\sigma_n \right\},$$

$$m_n(2, \epsilon) = \min \left\{ \ell : \sum_{i=1}^{\ell} \frac{\gamma_i}{\lambda_i} \geq (1 - \epsilon)\sigma_n \right\}, \quad \lambda'_N(\epsilon) = \max \left\{ \min_{m_n(1, \epsilon) \leq i \leq N} \lambda_i, \min_{1 \leq i \leq m_n(2, \epsilon)} \lambda_i \right\}.$$

A sequence $\{a_n\}$ is called *quasi-increasing* if there exists a positive constant C such that

$$a_k \leq Ca_n \text{ for each } k \leq n.$$

COROLLARY 1. Assume that (9) and (11) are satisfied and that there exists a positive constant C such that

$$\lambda_N^* \leq C(\lambda'_N(\epsilon))^{1+C\epsilon} \log^{C\epsilon} \sigma_N$$

for every $0 < \epsilon < 1$. Then, (13) and (14) hold. If we also have (15), then (16) and (17) are true.

COROLLARY 2. Assume that (11) is satisfied and that there is positive constants α and C such that $\sigma(n)/n^\alpha$ is quasi-increasing and $\lambda_{2k}^* \leq C \min_{k \leq l \leq 2k} \lambda_l$ for each $k \geq 1$. Then, (13) and (14) hold. If we also have (15), then (16) and (17) are true.

COROLLARY 3. Assume that (11) is satisfied and that

$$\log \frac{T_N \Gamma_N}{\sigma_N} = o(\log \log \sigma_N) \text{ as } N \rightarrow \infty.$$

Then, we have (13), (14) and

$$(21) \quad \limsup_{N \rightarrow \infty} X(T_N, N) / (2\sigma_N \log \log \sigma_N)^{1/2} = 1 \text{ a.s.}$$

COROLLARY 4. Assume that (9) and (11) are satisfied and that $\lambda_n \sigma_n^{1-\alpha}$ and $\sigma_n^{1/\alpha} / \lambda_n$ are quasi-increasing for some $0 < \alpha < 1$. Then, (13) and (14) hold. If we also have (15), then (16) and (17) are true.

The proof of theorems is based on the following lemmas.

LEMMA 1 (FERNIQUE(1964)). Let $G(t)$ be a Gaussian process on $[0, 1]$ with

$$E(G(t) - G(s))^2 \leq \Lambda^2(|t - s|)$$

where Λ is continuous, non-decreasing and satisfies $\int_1^\infty \Lambda(e^{-y^2}) dy < \infty$ and also $EG^2(t) \leq \Gamma^2$. Then, for every $x > 0$

$$P\left\{ \sup_{0 \leq t \leq 1} |G(t)| > x \left(\Gamma + 4 \int_1^\infty \Lambda(e^{-y^2}) dy \right) \right\} \leq d \int_x^\infty e^{-y^2/2} dy,$$

where d is an absolute constant.

LEMMA 2. For every $0 < \epsilon < 1$, there exists a constant $C = C(\epsilon)$ such that

$$(22) \quad P\left\{ \sup_{|t| \leq T} |X(t, n)| \geq x \sigma_n^{1/2} \right\} \leq C \left(1 + \frac{T \Gamma_n}{\sigma_n} \right) \exp\left(-\frac{(1-\epsilon)}{2} x^2 \right).$$

PROOF. Note that

$$(23) \quad EX^2(t, n) = \sigma_n$$

and

$$(24) \quad E(X(t, n) - X(s, n))^2 = 2 \sum_{i=1}^n \frac{\gamma_i}{\lambda_i} (1 - e^{-\lambda_i |t-s|}) \leq 2\Gamma_n |t-s|$$

for every t and s . Put

$$\delta = \frac{\epsilon}{32(1-\epsilon)^{1/2}}, \quad \theta = \frac{\delta^2 \sigma_n}{\Gamma_n}.$$

Then

$$(25) \quad P\left\{\sup_{|t| \leq T} |X(t, n)| \geq x\sigma_n^{1/2}\right\} \leq 2\left(\left[\frac{T}{\theta}\right] + 1\right)P\left\{\sup_{0 \leq t \leq \theta} |X(t, n)| \geq x\sigma_n^{1/2}\right\} \\ = 2\left(\left[\frac{T}{\theta}\right] + 1\right)P\left\{\sup_{0 \leq t \leq 1} |X(t\theta, n)| \geq x\sigma_n^{1/2}\right\}.$$

By (23) and (24), we have

$$(26) \quad \int_1^\infty (2\theta\Gamma_n e^{-y^2})^{1/2} dy \leq 4\delta\sigma_n^{1/2}.$$

Using the Fernique lemma, we find

$$(27) \quad P\left\{\sup_{0 \leq t \leq 1} |X(t\theta, n)| \geq x\sigma_n^{1/2}\right\} \\ \leq P\left\{\sup_{0 \leq t \leq 1} |X(t\theta, n)| \geq \frac{x}{1+16\delta}(\sigma_n^{1/2} + 4 \int_1^\infty (2\theta\Gamma_n e^{-y^2})^{1/2} dy)\right\} \\ \leq d \int_{\frac{x}{1+16\delta}}^\infty e^{-t^2/2} dt \\ \leq d \exp\left(-\frac{x^2}{2(1+16\delta)^2}\right) \\ \leq d \exp\left(-\frac{(1-\epsilon)x^2}{2}\right).$$

Now (22) follows from (27) and (25).

LEMMA 3. Let $0 < \epsilon < \frac{1}{2}$, $\theta_n(\epsilon)$ be the solution of the equation (8). Then, there is a positive $C(\epsilon)$ such that

$$(28) \quad P\left\{\sup_{0 \leq t \leq T} |X(t, n)| \leq x\sigma_n^{1/2}\right\} \leq \left(1 - C(\epsilon) \exp\left(-\frac{x^2}{2(1-2\epsilon)}\right)\right)^{T/\theta_n(\epsilon)}$$

for each $x > 0$.

PROOF. Let $\{W_i(t), 0 \leq t < \infty\}_{i=1}^\infty$ be independent standard Wiener processes. Noting that

$$\{X(t, n), 0 \leq t \leq T\} \text{ and } \left\{\sum_{i=1}^n \left(\frac{\gamma_i}{\lambda_i}\right)^{1/2} \frac{W_i(e^{2\lambda_i t})}{e^{\lambda_i t}}, 0 \leq t \leq T\right\}$$

have the same distribution, we have

$$(29) \quad P\left\{\sup_{0 \leq t \leq T} |X(t, n)| \leq x\sigma_n^{1/2}\right\} \leq P\left\{\sup_{0 \leq j \leq \lfloor \frac{T}{\theta_n} \rfloor} |X(j\theta_n, n)| \leq x\sigma_n^{1/2}\right\} \\ = P\left\{\max_{0 \leq j \leq \lfloor \frac{T}{\theta_n} \rfloor} \left|\sum_{i=1}^n \left(\frac{\gamma_i}{\lambda_i}\right)^{1/2} \frac{W_i(e^{2\lambda_i \theta_n})}{e^{\lambda_i \theta_n}}\right| \leq x\sigma_n^{1/2}\right\},$$

where $\theta_n = \theta_n(\epsilon)$. Set

$$U_j = \sum_{i=1}^n \left(\frac{\gamma_i}{\lambda_i}\right)^{1/2} \frac{W_i(e^{2j\lambda_i\theta_n})}{e^{j\lambda_i\theta_n}}, \quad V_j = \sum_{i=1}^n \left(\frac{\gamma_i}{\lambda_i}\right)^{1/2} \frac{W_i(e^{2(j-1)\lambda_i\theta_n})}{e^{j\lambda_i\theta_n}}.$$

It is easy to see that

$$U_j - V_j \sim N\left(0, \sum_{i=1}^n \frac{\gamma_i}{\lambda_i} (1 - e^{-2\lambda_i\theta_n})\right).$$

Whence

$$(30) \quad U_j - V_j \sim N(0, (1 - \epsilon)\sigma_n)$$

by the definition of θ_n . Thus, by (30), we obtain

(31)

$$\begin{aligned} & P\left\{\max_{0 \leq j \leq \lfloor \frac{T}{\theta_n} \rfloor} |U_j| \leq x\sigma_n^{1/2}\right\} \\ &= P\left\{\max_{0 \leq j < \lfloor \frac{T}{\theta_n} \rfloor} |U_j| \leq x\sigma_n^{1/2}, |U_{\lfloor \frac{T}{\theta_n} \rfloor} - V_{\lfloor \frac{T}{\theta_n} \rfloor} + V_{\lfloor \frac{T}{\theta_n} \rfloor}| \leq x\sigma_n^{1/2}\right\} \\ &= \int_{-\infty}^{\infty} P\{|U_{\lfloor \frac{T}{\theta_n} \rfloor} - V_{\lfloor \frac{T}{\theta_n} \rfloor} + y| \leq x\sigma_n^{1/2}\} dP\{V_{\lfloor \frac{T}{\theta_n} \rfloor} < y, \max_{0 \leq j < \lfloor \frac{T}{\theta_n} \rfloor} |U_j| \leq x\sigma_n^{1/2}\} \\ &= \int_{-\infty}^{\infty} \left(\Phi\left(\frac{x\sigma_n^{1/2} - y}{((1 - \epsilon)\sigma_n)^{1/2}}\right) - \Phi\left(\frac{-x\sigma_n^{1/2} - y}{((1 - \epsilon)\sigma_n)^{1/2}}\right)\right) dP \\ &\quad \left\{V_{\lfloor \frac{T}{\theta_n} \rfloor} < y, \max_{0 \leq j < \lfloor \frac{T}{\theta_n} \rfloor} |U_j| \leq x\sigma_n^{1/2}\right\} \\ &\leq \int_{-\infty}^{\infty} \left(\Phi\left(\frac{x}{(1 - \epsilon)^{1/2}}\right) - \Phi\left(\frac{-x}{(1 - \epsilon)^{1/2}}\right)\right) dP\{V_{\lfloor \frac{T}{\theta_n} \rfloor} < y, \max_{0 \leq j < \lfloor \frac{T}{\theta_n} \rfloor} |U_j| \leq x\sigma_n^{1/2}\} \\ &= \left(1 - \frac{2}{\sqrt{2\pi}} \int_{\frac{x}{(1-\epsilon)^{1/2}}}^{\infty} e^{-t^2/2} dt\right) P\left\{\max_{0 \leq j < \lfloor \frac{T}{\theta_n} \rfloor} |U_j| \leq x\sigma_n^{1/2}\right\} \\ &\leq (1 - C(\epsilon)e^{-\frac{x^2}{2(1-2\epsilon)}}) P\left\{\max_{0 \leq j < \lfloor \frac{T}{\theta_n} \rfloor} |U_j| \leq x\sigma_n^{1/2}\right\}, \end{aligned}$$

here we have used the following facts on the Wiener Process:

- i) $U_{\lfloor \frac{T}{\theta_n} \rfloor} - V_{\lfloor \frac{T}{\theta_n} \rfloor}$ and $\{V_{\lfloor \frac{T}{\theta_n} \rfloor}, U_j, 0 \leq j < \lfloor \frac{T}{\theta_n} \rfloor\}$ are independent,
- ii) $\Phi(x - y) - \Phi(-x - y) \leq \Phi(x) - \Phi(-x)$ for every $y \in R$ and $x \geq 0$,
- iii) for each $\delta > 0$, there is a $C(\delta) > 0$ such that

$$\int_x^{\infty} e^{-t^2/2} dt \geq C(\delta) \exp\left(-\frac{x^2(1 + \delta)}{2}\right) \text{ for every } x \geq 0.$$

By recurrence, we conclude from (29) and (31) that (28) holds true.

From (28) it is easy to see that

LEMMA 4. Let $0 < \epsilon < \frac{1}{2}$, $\theta_n(\epsilon)$ be the solution of the equation (8). Then, there is a positive $C(\epsilon)$ such that

$$(32) \quad P\left\{\sup_{0 \leq t \leq T} |X(t, n)| \geq x\sigma_n^{1/2}\right\} \geq C(\epsilon) \left(1 + \frac{T}{\theta_n}\right) \exp\left(-\frac{x^2}{2(1 - 2\epsilon)}\right)$$

for each $x \geq (2(1 - 2\epsilon) \log \frac{T}{\theta_n})^{1/2}$.

LEMMA 5. For each $0 < \epsilon < \frac{1}{2}$, there is a constant $C = C(\epsilon)$ such that

$$(33) \quad P\left\{ \max_{1 \leq n \leq N} \sup_{|t| \leq T} |X(t, n)| \geq x\sigma_N^{1/2} \right\} \leq C \left(1 + \frac{T\Gamma_N}{\sigma_N} \right) \exp\left(-\frac{(1 - 2\epsilon)x^2}{2} \right).$$

PROOF. (33) will follow from Lemma 2 and

$$(34) \quad P\left\{ \max_{1 \leq n \leq N} \sup_{|t| \leq T} |X(t, n)| \geq x\sigma_N^{1/2} \right\} \\ \leq 4 \left(1 + \frac{T\Gamma_N}{\sigma_N} \right) P\left\{ \sup_{|t| \leq \sigma_N/\Gamma_N} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2} \right\}$$

for every x sufficiently large. Let

$$B = \sigma_N/\Gamma_N, \quad E_1 = \left\{ \sup_{|t| \leq B} |X(t, 1)| \geq x\sigma_N^{1/2} \right\}, \\ E_i = \left\{ \max_{j < i} \sup_{|t| \leq B} |X(t, j)| < x\sigma_N^{1/2} \leq \sup_{|t| \leq B} |X(t, i)| \right\}, \quad i = 2, \dots, N.$$

Noting that

$$\left\{ \max_{1 \leq n \leq N} \sup_{|t| \leq B} |X(t, n)| \geq x\sigma_N^{1/2} \right\} = \bigcup_{n=1}^N E_n \subset \left\{ \sup_{|t| \leq B} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2} \right\} \\ \cup \bigcup_{n=1}^{N-1} \left(E_n \cap \left\{ \sup_{|t| \leq B} |X(t, N)| < x(1 - \epsilon)\sigma_N^{1/2} \right\} \right) \\ \subset \left\{ \sup_{|t| \leq B} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2} \right\} \\ \cup \bigcup_{n=1}^{N-1} \left(E_n \cap \left\{ \sup_{|t| \leq B} |X(t, N) - X(t, n)| \geq \epsilon x\sigma_N^{1/2} \right\} \right)$$

and that $\{X(t, N) - X(t, n), |t| \leq B\}$ and E_n are independent, we have

$$P\left\{ \max_{1 \leq n \leq N} \sup_{|t| \leq B} |X(t, n)| \geq x\sigma_N^{1/2} \right\} \\ \leq P\left\{ \sup_{|t| \leq B} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2} \right\} \\ + \sum_{n=1}^{N-1} P\left\{ \sup_{|t| \leq B} |X(t, N) - X(t, n)| \geq \epsilon x\sigma_N^{1/2} \right\} P(E_n) \\ \leq P\left\{ \sup_{|t| \leq B} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2} \right\} \\ + \sum_{n=1}^{N-1} d \left(1 + \frac{B \sum_{i=1+n}^N \gamma_i}{\sum_{i=1+n}^N \gamma_i / \lambda_i} \right) \exp\left(-\frac{\epsilon^2 x^2 \sigma_N}{4 \sum_{i=1+n}^N \gamma_i / \lambda_i} \right) P(E_n)$$

$$\begin{aligned} &\leq P\left\{\sup_{|t|\leq B} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2}\right\} + 2d \exp\left(-\frac{\epsilon^2 x^2}{4}\right) \sum_{n=1}^{N-1} P(E_n) \\ &\leq P\left\{\sup_{|t|\leq B} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2}\right\} + \frac{1}{2}P\left\{\max_{1\leq n\leq N} \sup_{|t|\leq B} |X(t, n)| \geq x\sigma_N^{1/2}\right\} \end{aligned}$$

provided $x \geq 4(\log(8d))/\epsilon$. In the last but second inequality we have used the fact that $f(y) = ye^{-ay}$ is decreasing on $[1/a, \infty)$ for each $a > 0$ fixed, and d is an absolute constant as in Lemma 2. The above inequality yields

$$P\left\{\max_{1\leq n\leq N} \sup_{|t|\leq B} |X(t, n)| \geq x\sigma_N^{1/2}\right\} \leq 2P\left\{\sup_{|t|\leq B} |X(t, N)| \geq x(1 - \epsilon)\sigma_N^{1/2}\right\}$$

for $x \geq 4(\log(8d))/\epsilon$, as desired.

PROOF OF THEOREM 1. It suffices to show that for each $0 < \epsilon < 1/8$

$$(35) \quad \limsup_{N \rightarrow \infty} \max_{1 \leq n \leq N} \sup_{|t| \leq T_N} |X(t, n)| / \beta_N \leq 1 + 8\epsilon \text{ a.s.}$$

For $k \geq 0$, put

$$\begin{aligned} H_k &= \{N : (1 + \epsilon)^k < \beta_N \leq (1 + \epsilon)^{k+1}\}, \\ M_k &= \max\{N : N \in H_k\}. \end{aligned}$$

Clearly, (9) implies that $\beta_N \rightarrow \infty$ as $N \rightarrow \infty$. So, we have

$$\begin{aligned} (36) \quad \limsup_{N \rightarrow \infty} \max_{1 \leq n \leq N} \sup_{|t| \leq T_N} |X(t, n)| / \beta_N &\leq \limsup_{k \rightarrow \infty} \max_{N \in H_k} \max_{1 \leq n \leq N} \sup_{|t| \leq T_N} |X(t, n)| / \beta_N \\ &\leq (1 + \epsilon) \limsup_{k \rightarrow \infty} \max_{1 \leq n \leq M_k} \sup_{|t| \leq T_{M_k}} |X(t, n)| / (1 + \epsilon)^{k+1}. \end{aligned}$$

From the definition of M_k , we find that

$$\frac{(1 + \epsilon)^{2k}}{2(\log(T_{M_k} \Gamma_{M_k} / \sigma_{M_k}) + \log \log \sigma_{M_k})} < \sigma_{M_k} \leq \frac{(1 + \epsilon)^{2(k+1)}}{2(\log(T_{M_k} \Gamma_{M_k} / \sigma_{M_k}) + \log \log \sigma_{M_k})}.$$

Whence, for each $k \geq 1$

$$(37) \quad \sigma_{M_k} \geq (1 + \epsilon)^k \text{ or } \left(\frac{T_{M_k} \Gamma_{M_k}}{\sigma_{M_k}} + e\right) \geq \exp\left(\frac{1}{4}(1 + \epsilon)^k\right).$$

Using Lemma 5, we deduce

$$\begin{aligned}
 (38) \quad & P\left\{ \max_{1 \leq n \leq M_k} \sup_{|t| \leq T_{M_k}} |X(t, n)| \geq (1 + \epsilon)^{k+1} (1 + \epsilon)^2 \right\} \\
 & \leq P\left\{ \max_{1 \leq n \leq M_k} \sup_{|t| \leq T_{M_k}} |X(t, n)| \geq \beta_{M_k} (1 + \epsilon)^2 \right\} \\
 & \leq C(\epsilon) \left(1 + \frac{T_{M_k} \Gamma_{M_k}}{\sigma_{M_k}} \right) \exp\left(-(1 + \epsilon) \left(\log \frac{T_{M_k} \Gamma_{M_k}}{\sigma_{M_k}} + \log \log \sigma_{M_k} \right) \right) \\
 & \leq C(\epsilon) \left(1 + \frac{T_{M_k} \Gamma_{M_k}}{\sigma_{M_k}} \right)^{-\epsilon} (\log \sigma_{M_k})^{-(1+\epsilon)} \\
 & \leq C(\epsilon) k^{-(1+\epsilon)}
 \end{aligned}$$

by (37). Now (35) follows from (36), (38) and the Borel-Cantelli lemma. This completes the proof of Theorem 1.

PROOF OF THEOREM 2. Noting that σ_N is non-decreasing, we have

$$\sigma_N \rightarrow \sigma \text{ as } N \rightarrow \infty,$$

where $0 < \sigma \leq \infty$. If $0 < \sigma < \infty$, then (9) implies $T_N \Gamma_N / \sigma_N \rightarrow \infty$ and hence (15) is satisfied. So we only need to consider two cases: one is $\sigma = \infty$, the other is (15) being satisfied. We formulate the proof below in two steps, which together with (10) will imply our statements.

STEP 1. Suppose $\sigma = \infty$, then, for each $0 < \epsilon < 1/(4C^2)$

$$(39) \quad \limsup_{N \rightarrow \infty} \sup_{0 \leq t \leq T_N} |X(t, N)| / \beta_N \geq 1 - \epsilon^{1/2} \text{ a.s.}$$

Let

$$N_1 = 1, N_{k+1} = \min \left\{ n : \sigma_n \geq \left(\frac{8C^2}{\epsilon^2} \right)^k \right\}, \quad k = 1, 2, \dots$$

From condition (11), we get

$$(40) \quad \left(\frac{8C^2}{\epsilon^2} \right)^k < \sigma_{N_{k+1}} \leq C \left(\frac{8C^2}{\epsilon^2} \right)^k.$$

Clearly, $\sigma = \infty$ implies $N_k \uparrow \infty$ as $k \rightarrow \infty$. Then

$$\begin{aligned}
 (41) \quad \limsup_{N \rightarrow \infty} \sup_{0 \leq t \leq T_N} |X(t, N)| / \beta_N & \geq \limsup_{k \rightarrow \infty} \sup_{0 \leq t \leq T_{N_k}} |X(t, N_k)| / \beta_{N_k} \\
 & \geq \limsup_{k \rightarrow \infty} \sup_{0 \leq t \leq T_{N_k}} |X(t, N_k) - X(t, N_{k-1})| / \beta_{N_k} \\
 & \quad - \limsup_{k \rightarrow \infty} \sup_{0 \leq t \leq T_{N_k}} |X(t, N_{k-1})| / \beta_{N_k}.
 \end{aligned}$$

Using Lemma 5 again, we have

$$\begin{aligned}
 &P\left\{\sup_{0 \leq t \leq T_{N_k}} |X(t, N_{k-1})|/\beta_{N_k} \geq \frac{\epsilon}{2}\right\} \\
 &\leq C(\epsilon)\left(1 + \frac{T_{N_k}\Gamma_{N_{k-1}}}{\sigma_{N_{k-1}}}\right)\exp\left(-\frac{\epsilon^2\sigma_{N_k}}{9\sigma_{N_{k-1}}}\left(\log\frac{T_{N_k}\Gamma_{N_k}}{\sigma_{N_k}} + \log\log\sigma_{N_k}\right)\right) \\
 &\leq C(\epsilon)\left(1 + \frac{T_{N_k}\Gamma_{N_{k-1}}}{\sigma_{N_{k-1}}}\right)\left(1 + \frac{T_{N_k}\Gamma_{N_k}}{\sigma_{N_k}}\right)^{-2}\log^{-2}\sigma_{N_k} \\
 &\leq C(\epsilon)k^{-2}
 \end{aligned}$$

by (40). This implies that

$$(42) \quad \limsup_{k \rightarrow \infty} \sup_{0 \leq t \leq T_{N_k}} |X(t, N_{k-1})|/\beta_{N_k} \leq \frac{\epsilon}{2} \text{ a.s.}$$

To estimate $|X(t, N_k) - X(t, N_{k-1})|/\beta_{N_k}$, we let $\theta_k^*(\epsilon)$ be the solution of the equation

$$\sum_{i=1+N_{k-1}}^{N_k} \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i\theta_k^*(\epsilon)} = \epsilon \sum_{i=1+N_{k-1}}^{N_k} \frac{\gamma_i}{\lambda_i}$$

and let

$$\beta'_k = \left(2\left(\sum_{i=1+N_{k-1}}^{N_k} \frac{\gamma_i}{\lambda_i}\right)\left(\log(T_{N_k}/\theta_k^*(\epsilon)) + \log\log\sum_{i=1+N_{k-1}}^{N_k} \frac{\gamma_i}{\lambda_i}\right)\right).$$

Then, in terms of (32), we obtain

$$\begin{aligned}
 &P\left\{\sup_{0 \leq t \leq T_{N_k}} |X(t, N_k) - X(t, N_{k-1})|/\beta'_k \geq (1 - 2\epsilon)^{1/2}\right\} \\
 &\geq C(\epsilon)\left(1 + \frac{T_{N_k}}{\theta_k^*(\epsilon)}\right)\exp(-(\beta'_k)^2/2) \\
 &\geq C(\epsilon)\log^{-1}\left(\sum_{i=1+N_{k-1}}^{N_k} \frac{\gamma_i}{\lambda_i}\right) \\
 &\geq C(\epsilon)k^{-1}
 \end{aligned}$$

by (40) again. Therefore, we have

$$(43) \quad \limsup_{k \rightarrow \infty} \sup_{0 \leq t \leq T_{N_k}} |X(t, N_k) - X(t, N_{k-1})|/\beta'_k \geq (1 - 2\epsilon)^{1/2} \text{ a.s.,}$$

since $\{\sup_{0 \leq t \leq T_{N_k}} |X(t, N_k) - X(t, N_{k-1})|, k \geq 1\}$ are independent random variables.

On the other hand, it follows from the definitions of $\theta_{N_k}(\epsilon/2)$ and θ_k^* that

$$\begin{aligned}
 \frac{1}{4}\epsilon\sigma_{N_k} &= \sum_{i=1}^{N_k} \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i\theta_{N_k}(\frac{\epsilon}{4})} \\
 &\geq \sum_{i=1+N_{k-1}}^{N_k} \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i\theta_{N_k}(\frac{\epsilon}{4})}
 \end{aligned}$$

and

$$\begin{aligned} \frac{1}{2}\epsilon\sigma_{N_k} &= \frac{\epsilon}{2}(\sigma_{N_k} - \sigma_{N_{k-1}} + \sigma_{N_{k-1}}) \\ &\leq \frac{\epsilon}{2}(\sigma_{N_k} - \sigma_{N_{k-1}}) + \frac{\epsilon}{4}\sigma_{N_k}. \end{aligned}$$

From the latter, we find that $\sigma_{N_k} \leq 4(\sigma_{N_k} - \sigma_{N_{k-1}})$. Hence

$$\sum_{t=1+N_{k-1}}^{N_k} \frac{\gamma_t}{\lambda_t} e^{-2\lambda_t\theta_{N_k}(\frac{\epsilon}{4})} \leq \sum_{t=1+N_{k-1}}^{N_k} \frac{\gamma_t}{\lambda_t} e^{-2\lambda_t\theta_k^*(\epsilon)},$$

which is equivalent to say that $\theta_{N_k}(\frac{\epsilon}{4}) \geq \theta_k^*(\epsilon)$. Combining the above results with the assumption (12), we finally conclude that

$$(44) \quad \limsup_{k \rightarrow \infty} \sup_{0 \leq t \leq T_{N_k}} |X(t, N_k) - X(t, N_{k-1})| / \beta_{N_k} \geq \frac{(1 - 2\epsilon)}{(1 + C\epsilon)^2} \text{ a.s.}$$

This proves (39) by (41), (42) and (44).

STEP 2. If, in addition, (15) is satisfied, then for each $0 < \epsilon < \frac{1}{8}$

$$(45) \quad \liminf_{N \rightarrow \infty} \sup_{0 \leq t \leq T_N} |X(t, N)| / \alpha_N \geq 1 - 4\epsilon \text{ a.s.,}$$

where $\alpha_N = (2\sigma_N \log \frac{T_N \Gamma_N}{\sigma_N})^{1/2}$.

Let $1 < \theta < 1 + \frac{\epsilon^2}{6}$. Define

$$\begin{aligned} A_k &= \{N : \theta^k \sigma_1 \leq \sigma_N < \theta^{k+1} \sigma_1\}, \quad k = 0, 1, \dots, \\ B_j &= \left\{N : \theta^j \leq \frac{T_N \Gamma_N}{\sigma_N} + 1 < \theta^{j+1}\right\}, \quad j = 0, 1, \dots, \\ L_{k,j} &= \min\{N : N \in A_k B_j\}, \quad L_{k,j}^* = \max\{N : N \in A_k B_j\}, \\ \Gamma_{k,j} &= \sum_{t=1+L_{k,j}}^{L_{k,j}^*} \gamma_t, \quad \sigma_{k,j} = \sum_{t=1+L_{k,j}}^{L_{k,j}^*} \frac{\gamma_t}{\lambda_t}. \end{aligned}$$

Clearly, (15) implies that $T_N \Gamma_N / \sigma_N \rightarrow \infty$ and that $A_k B_j = \emptyset$ if $k \geq \theta^{\epsilon j}$, when j is sufficiently large. Thus, we have

$$\begin{aligned} (46) \quad \liminf_{N \rightarrow \infty} \sup_{0 \leq t \leq T_N} |X(t, N)| / \alpha_N &\geq \liminf_{J \rightarrow \infty} \inf_{N \in B_j} \sup_{0 \leq t \leq T_N} |X(t, N)| / \alpha_N \\ &\geq \liminf_{J \rightarrow \infty} \inf_{0 \leq k \leq \theta^j} \inf_{N \in B_j A_k} \sup_{0 \leq t \leq T_N} |X(t, N)| / \alpha_N \\ &\geq \liminf_{J \rightarrow \infty} \inf_{0 \leq k \leq \theta^j} \inf_{N \in B_j A_k} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, N)|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \\ &\geq \liminf_{J \rightarrow \infty} \inf_{0 \leq k \leq \theta^j} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \\ &\quad - \limsup_{J \rightarrow \infty} \sup_{0 \leq k \leq \theta^j} \sup_{L_{k,j} \leq N \leq L_{k,j}^*} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, N) - X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}}. \end{aligned}$$

Similarly to (33), we can obtain that

$$P \left\{ \sup_{L_{k,j} \leq N \leq L_{k,j}^*} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, N) - X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \geq \epsilon \right\} \leq C(\epsilon) \left(1 + \frac{T_{L_{k,j}} \Gamma_{k,j}}{\sigma_{k,j}} \right) \exp \left(-\frac{\epsilon^2 \theta^{k+1} \log \theta^{j+1}}{2\sigma_{k,j}} \right).$$

Since xe^{-x} is decreasing on $[1, \infty)$ and $\sigma_{k,j} = \sigma_{L_{k,j}^*} - \sigma_{L_{k,j}} \leq (\theta - 1)\theta^k$, the above inequality is bounded by

$$C(\epsilon) \left(1 + \frac{T_{L_{k,j}^*} \Gamma_{L_{k,j}^*}}{\sigma_{L_{k,j}^*}} \right) \exp \left(-\frac{\epsilon^2 \log \theta^{j+1}}{2(\theta - 1)} \right) \leq C(\epsilon)\theta^{j+1} \exp(-3 \log \theta^{j+1}) \leq C(\epsilon)\theta^{-2j}$$

for every j sufficiently large. Therefore

$$P \left\{ \sup_{0 \leq k \leq \theta^j} \sup_{L_{k,j} \leq N \leq L_{k,j}^*} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, N) - X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \geq \epsilon \right\} \leq C(\epsilon)\theta^{-j},$$

which follows that

$$(47) \quad \limsup_{j \rightarrow \infty} \sup_{0 \leq k \leq \theta^j} \sup_{L_{k,j} \leq N \leq L_{k,j}^*} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, N) - X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \leq \epsilon \text{ a.s.}$$

On the other hand, using (28), we have

$$\begin{aligned} P \left\{ \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \leq \frac{1 - 2\epsilon}{\theta} \right\} &\leq P \left\{ \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, L_{k,j})|}{(2\sigma_{L_{k,j}} \log \theta^{j+1})^{1/2}} \leq 1 - 2\epsilon \right\} \\ &\leq \left(1 - C(\epsilon) \exp(- (1 - 2\epsilon) \log \theta^{j+1}) \right)^{T_{L_{k,j}} / \theta_{L_{k,j}}(\epsilon)} \\ &\leq \exp \left(-\frac{C(\epsilon) T_{L_{k,j}}}{\theta^{(1-2\epsilon)j} \theta_{L_{k,j}}(\epsilon)} \right) \\ &\leq \exp(-C(\epsilon)\theta^{\epsilon j}) \end{aligned}$$

by (12) and (15), for every sufficiently large j , and hence

$$P \left\{ \inf_{0 \leq k \leq \theta^j} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \leq \frac{1 - 2\epsilon}{\theta} \right\} \leq \theta^{\epsilon j} \exp(-C(\epsilon)\theta^{\epsilon j}) \leq \theta^{-j}$$

provided that j is sufficiently large, which implies immediately

$$(48) \quad \liminf_{j \rightarrow \infty} \inf_{0 \leq k \leq \theta^j} \sup_{0 \leq t \leq T_{L_{k,j}}} \frac{|X(t, L_{k,j})|}{(2\theta^{k+1} \log \theta^{j+1})^{1/2}} \geq \frac{1 - 2\epsilon}{\theta} \text{ a.s.}$$

by the Borel-Cantelli lemma.

Now (45) follows from (46)–(48). This completes the proof of Theorem 2.

PROOF OF THEOREM 3. It suffices to show that

$$(49) \quad \forall A > 0, \quad \limsup_{n \rightarrow \infty} \sup_{|t| \leq A} \frac{|X(t, n)|}{(2\sigma_n \log \log \sigma_n)^{1/2}} \leq 1 \text{ a.s.}$$

and

$$(50) \quad \forall \epsilon > 0, \forall A > 0, \quad \lim_{n \rightarrow \infty} P \left\{ \bigcup_{|t| \leq A} \bigcap_{i=n}^{\infty} \{X(t, i) < (1 - \epsilon)(2\sigma_i \log \log \sigma_i)^{1/2}\} \right\} = 0$$

hold true.

(49) follows from Theorem 1 and (18) immediately. We now prove (50). Let

$$0 < \epsilon < \frac{1}{4}, \quad n_k = \max\{n : \sigma_n \leq a^k\}$$

where $a > 1$ is a constant which will be specified later. Then

$$\frac{a^k}{C} \leq \sigma_{n_k} \leq a^k.$$

Clearly, (50) is implied by

$$(51) \quad \lim_{k \rightarrow \infty} P \left\{ \bigcup_{|t| \leq A} \bigcap_{i=k}^{\infty} \{X(t, n_i) < (1 - \epsilon)(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2}\} \right\} = 0.$$

Noting that

$$\begin{aligned} & \{X(t, n_i) < (1 - \epsilon)(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2}\} \\ & \subset \{X(t, n_{i-1}) < -\frac{\epsilon}{2}(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2}\} \\ & \cup \{X(t, n_i) - X(t, n_{i-1}) < (1 - \frac{\epsilon}{2})(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2}\}, \end{aligned}$$

we have

$$\begin{aligned} & \bigcup_{|t| \leq A} \bigcap_{i=k}^{\infty} \{X(t, n_i) < (1 - \epsilon)(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2}\} \\ & \subset \bigcup_{|t| \leq A} \bigcap_{i=k}^{\infty} \left\{ X(t, n_i) - X(t, n_{i-1}) < \left(1 - \frac{\epsilon}{2}\right)(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\} \\ & \cup \bigcup_{|t| \leq A} \bigcup_{i=k}^{\infty} \left\{ X(t, n_{i-1}) < -\frac{\epsilon}{2}(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\}. \end{aligned}$$

From Theorem 1 and (18) it follows that

$$P \left\{ \bigcup_{|t| \leq A} \bigcup_{i=k}^{\infty} \left\{ X(t, n_{i-1}) < -\frac{\epsilon}{2}(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\} \right\} \rightarrow 0 \text{ as } k \rightarrow \infty$$

provided $a > 8C/\epsilon^2$.

The rest we should do is to prove

$$(52) \quad P \left\{ \bigcup_{|t| \leq A} \bigcap_{i=k}^{\infty} \left\{ X(t, n_i) - X(t, n_{i-1}) < \left(1 - \frac{\epsilon}{2}\right) (2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\} \right\} \longrightarrow 0$$

as $k \rightarrow \infty$. Let $b := b_k = 1/(Ak^2)$. Then

$$\begin{aligned} (53) \quad & P \left\{ \bigcup_{|t| \leq A} \bigcap_{i=k}^{\infty} \left\{ X(t, n_i) - X(t, n_{i-1}) < \left(1 - \frac{\epsilon}{2}\right) (2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\} \right\} \\ & \leq P \left\{ \bigcup_{|t| \leq A} \bigcap_{i=k}^{2k} \left\{ X(t, n_i) - X(t, n_{i-1}) < \left(1 - \frac{\epsilon}{2}\right) (2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\} \right\} \\ & \leq 4k^2 P \left\{ \bigcup_{0 \leq t \leq b} \bigcap_{i=k}^{2k} \left\{ X(t, n_i) - X(t, n_{i-1}) < \left(1 - \frac{\epsilon}{2}\right) (2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\} \right\} \\ & \leq 4k^2 P \left\{ \bigcap_{i=k}^{2k} \left\{ X(b, n_i) - X(b, n_{i-1}) < \left(1 - \frac{\epsilon}{3}\right) (2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2} \right\} \right\} \\ & \quad + 4k^2 P \left\{ \bigcup_{0 \leq t \leq b} \bigcup_{i=k}^{2k} \left\{ \frac{X(t, n_i) - X(t, n_{i-1}) - X(b, n_i) + X(b, n_{i-1})}{(2\sigma_{n_i} \log \log \sigma_{n_i})^{1/2}} < -\frac{\epsilon}{6} \right\} \right\} \\ & := I_1(k) + I_2(k). \end{aligned}$$

Since $\{X(b, n_i) - X(b, n_{i-1}), k \leq i \leq 2k\}$ are independent, we have

$$\begin{aligned} (54) \quad I_1(k) & \leq 4k^2 \prod_{i=k}^{2k} \left(1 - C(\epsilon) \exp\left(-\left(1 - \frac{\epsilon}{6}\right) \log \log \sigma_{n_i}\right)\right) \\ & \leq 4k^2 \prod_{i=k}^{2k} \left(1 - C(\epsilon) i^{-1+\frac{\epsilon}{6}}\right) \\ & \leq 4k^2 \exp\left(-\sum_{i=k}^{2k} C(\epsilon) i^{-1+\frac{\epsilon}{6}}\right) \\ & \leq 4k^2 \exp\left(-C(\epsilon) k^{\epsilon/6}\right) \longrightarrow 0, \text{ as } k \rightarrow \infty. \end{aligned}$$

On the other hand, for $0 \leq t \leq b$ and $k \leq i \leq 2k$ we have

$$\begin{aligned} E \left(X(t, n_i) - X(t, n_{i-1}) - X(b, n_i) + X(b, n_{i-1}) \right)^2 & = 2 \sum_{j=1+n_{i-1}}^{n_i} \frac{\gamma_j}{\lambda_j} (1 - e^{-2\lambda_j(b-t)}) \\ & \leq 4(b-t) \sum_{j=1+n_{i-1}}^{n_i} \gamma_j \\ & \leq \frac{4\Gamma_{n_i}}{A\sigma_{n_i}} k^{-2} \sigma_{n_i} \\ & \leq 4k^{-2} \sigma_{n_i} (\log \sigma_{n_i}) / A \\ & \leq 8\sigma_{n_i} (\log a) / (Ak) \\ & \leq \epsilon \sigma_{n_i} / 48 \end{aligned}$$

provided that k is large enough.

Consequently, using the Fernique lemma again, we get

$$\begin{aligned} I_2(k) &\leq C(\epsilon)k^3 \max_{k \leq i \leq 2k} \left(1 + \frac{b\Gamma_{n_i}}{\sigma_{n_i}}\right) \exp(-4 \log \log \sigma_{n_k}) \\ &\leq C(\epsilon)k^3 \exp(-4 \log \log \sigma_{n_k}) \\ &\leq C(\epsilon)k^{-1} \longrightarrow 0 \text{ as } k \rightarrow \infty. \end{aligned}$$

This proves (52) by (53) and (54), as desired. The proof of Theorem 3 is completed.

PROOF OF COROLLARY 1. It is easy to see that

$$\begin{aligned} \frac{T_N \Gamma_N}{\sigma_N} &\leq \lambda_N^*, \\ 2\epsilon\sigma_n &= \sum_{i=1}^n \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i \theta_n(2\epsilon)} \\ &\leq \sum_{i=1}^{m_n(1,\epsilon)-1} \frac{\gamma_i}{\lambda_i} + \sum_{i=m_n(1,\epsilon)}^n \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i \theta_n(2\epsilon)} \\ &\leq \epsilon\sigma_n + \sum_{i=m_n(1,\epsilon)}^n \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i \theta_n(2\epsilon)} \\ &\leq \epsilon\sigma_n + \left(\sum_{i=m_n(1,\epsilon)}^n \frac{\gamma_i}{\lambda_i}\right) \exp(-2 \min_{m_n(1,\epsilon) \leq i \leq n} \{\lambda_i\} \theta_n(2\epsilon)) \\ &\leq \epsilon\sigma_n + \sigma_n \exp(-2 \min_{m_n(1,\epsilon) \leq i \leq n} \{\lambda_i\} \theta_n(2\epsilon)). \end{aligned}$$

The latter implies that

$$\frac{1}{\theta_n(2\epsilon)} \geq 2 \min_{m_n(1,\epsilon) \leq i \leq n} \{\lambda_i\} / \log(1/\epsilon).$$

Similarly, we have

$$\frac{1}{\theta_n(2\epsilon)} \geq 2 \min_{1 \leq i \leq m_n(2,\epsilon)} \{\lambda_i\} / \log(1/\epsilon).$$

Consequently, we obtain

$$\frac{1}{\theta_n(2\epsilon)} \geq 2\lambda'_n(\epsilon) / \log(1/\epsilon)$$

This indicates that the condition (12) is satisfied. The corollary now follows from Theorems 2 and 3.

PROOF OF COROLLARY 2. Since σ_n/n^α is quasi-increasing, there exists a positive constant C such that

$$(55) \quad \sigma_\ell / \ell^\alpha \leq C\sigma_n / n^\alpha$$

for each $\ell \leq n$. From (55) we can find that for every $0 < \epsilon < \frac{1}{4}$

$$\sigma_\ell \leq \epsilon\sigma_n \text{ for each } \ell \leq \left(\frac{\epsilon}{C}\right)^{1/\alpha} n$$

and hence

$$(56) \quad m_n(1, \epsilon) \geq \left(\frac{\epsilon}{C}\right)^{1/\alpha} n.$$

On the other hand, it is easy to find that from the assumption $\lambda_{2k}^* \leq C \min_{k \leq l \leq 2k} \lambda_l$, for each $0 < \epsilon < \frac{1}{4}$, there exists a constant $C(\epsilon)$ such that

$$(57) \quad \lambda_n^* \leq C(\epsilon) \min_{\epsilon n \leq l \leq n} \lambda_l.$$

Thus, the assumption of Corollary 1 is satisfied by (56) and (57) and hence the corollary holds.

The proof of Corollary 3 is trivial and so is omitted here.

PROOF OF COROLLARY 4. By the assumption of quasi-increasing, there is a positive constant C such that for each $k \leq n$

$$\lambda_k \sigma_k^{1-\alpha} \leq C \lambda_n \sigma_n^{1-\alpha}$$

and

$$\sigma_k^{1/\alpha} / \lambda_k \leq C \sigma_n^{1/\alpha} / \lambda_n.$$

Then

$$\begin{aligned} \frac{\Gamma_n}{\sigma_n} &= \left(\sum_{i=1}^n \frac{\gamma_i}{\lambda_i} \lambda_i\right) / \sigma_n \\ &\leq \left(\sum_{i=1}^n \frac{(\sigma_i - \sigma_{i-1})}{\sigma_i^{1-\alpha}} \lambda_n \sigma_n^{1-\alpha}\right) / \sigma_n \\ &\leq C \lambda_n / \alpha, \end{aligned}$$

and

$$\begin{aligned} 2\epsilon \sigma_n &= \sum_{i=1}^n \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i \theta_n(2\epsilon)} \\ &\leq \epsilon \sigma_n + \sum_{i=m_n(1,\epsilon)}^n \frac{\gamma_i}{\lambda_i} e^{-2\lambda_i \theta_n(2\epsilon)} \\ &\leq \epsilon \sigma_n + \sum_{i=m_n(1,\epsilon)}^n \frac{\gamma_i}{\lambda_i} \exp\left(-\frac{2\lambda_n \sigma_i^{1/\alpha} \theta_n(2\epsilon)}{C \sigma_n^{1/\alpha}}\right) \\ &\leq \epsilon \sigma_n + \sum_{i=m_n(1,\epsilon)}^n \frac{\gamma_i}{\lambda_i} \exp\left(-\frac{2\lambda_n \epsilon^{1/\alpha} \theta_n(2\epsilon)}{C}\right) \\ &\leq \epsilon \sigma_n + \sigma_n \exp\left(-\frac{2\lambda_n \epsilon^{1/\alpha} \theta_n(2\epsilon)}{C}\right). \end{aligned}$$

Therefore, we have

$$\frac{1}{\theta_n(2\epsilon)} \geq \frac{2\lambda_n \epsilon^{1/\alpha}}{C \log(1/\epsilon)}.$$

This proves that condition (12) is also satisfied and hence the corollary follows from Theorems 2 and 3.

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Department of Mathematics
Hangzhou University
Hangzhou, Zhejiang
People's Republic of China

Present address
Department of Mathematics
National University of Singapore
Singapore 0511