



Convergence rates of limit distribution of maxima of lognormal samples

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ABSTRACT

Let M_n denote the partial maximum of an independent and identically distributed lognormal random sequence. In this paper, we derive the exact uniform convergence rate of the distribution of the normalized maximum $(M_n - b_n)/a_n$ to its extreme value distribution, where the constants a_n and b_n are chosen by an optimal way.
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1. Introduction and main results

Let $(\xi_n, n \geq 1)$ be a sequence of independent and identically distributed standardized normal random variables with distribution function $\Phi(x)$. Let $M_n = \max(\xi_k, 1 \leq k \leq n)$ denote the partial maximum. It is well-known that

$$\lim_{n \rightarrow \infty} P(M_n \leq a_n x + b_n) = \lim_{n \rightarrow \infty} \Phi^n(a_n x + b_n) = \exp(-e^{-x}) = \Lambda(x) \tag{1.1}$$

with the normalized constants a_n and b_n given by

$$a_n = (2 \log n)^{-1/2}, \quad b_n = a_n^{-1} - \frac{a_n}{2} (\log \log n + \log 4\pi).$$

The pointwise convergence rate of (1.1) is

$$\Phi^n(a_n x + b_n) - \Lambda(x) \sim \frac{e^{-x} \exp(-e^{-x}) (\log \log n)^2}{16 \log n} \tag{1.2}$$

for large n . For more details, see [1,2]. Hall [1] investigated this problem further and proved

$$\frac{C_1}{\log n} < \sup_{x \in \mathbb{R}} |P(M_n \leq a_n x + b_n) - \Lambda(x)| < \frac{C_2}{\log n} \tag{1.3}$$

for some absolute constants $0 < C_1 < C_2$ if we choose a_n and b_n by the following equations:

$$2\pi b_n^2 \exp(b_n^2) = n^2, \quad a_n = b_n^{-1}. \tag{1.4}$$

For the extreme value distributions and their associated uniform convergence rates of some given distributions, see [3,4] respectively for exponential and mixed exponential distributions, [5,6] for the general error distribution, [7,8] for the short-tailed symmetric distribution.

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It is popular in theoretical analysis and wide applications that the normal distribution is carried over to logarithmic normal one. Let $\xi \sim N(0, 1)$, we say that X follows the logarithmic normal distribution if X is defined by $X = e^{\xi}$, abbreviated as *lognormal distribution*. In this paper, we are interested in the uniform convergence rate of extremes from samples with common distribution F following the lognormal distribution. The probability density function of the lognormal distribution is given by

$$f(x) = \frac{x^{-1}}{\sqrt{2\pi}} \exp\left(-\frac{(\log x)^2}{2}\right), \quad x > 0.$$

In the sequel, let $(X_n, n \geq 1)$ be a sequence of independent random variables with common distribution F following the lognormal distribution, and let M_n denote its partial maximum. For the limiting distribution of the maximum of lognormal, by Lemma 2 in Section 2 and the arguments similar to the proof of Theorem 1.5.3 in [2], we can derive

$$\lim_{n \rightarrow \infty} P(M_n \leq \alpha_n x + \beta_n) = \lim_{n \rightarrow \infty} F^n(\alpha_n x + \beta_n) = \Lambda(x)$$

with normalized constants α_n and β_n given by

$$\alpha_n = \frac{\exp((2 \log n)^{1/2})}{(2 \log n)^{1/2}}, \tag{1.5}$$

and

$$\beta_n = (\exp((2 \log n)^{1/2})) \left(1 - \frac{\log 4\pi + \log \log n}{2(2 \log n)^{1/2}}\right). \tag{1.6}$$

As mentioned by Hall [1] and Leadbetter et al. [2] that the choices of normalized constants may influence the convergence rate of extremes, and the distribution tail representation may help us find the optimal normalized constants. For the distributional tail representation of lognormal distribution, Lemma 2(ii) in Section 2 shows that

$$1 - F(x) = c(x) \exp\left(-\int_e^x \frac{g(t)}{f(t)} dt\right)$$

for sufficiently large x , where $c(x) \rightarrow (2\pi e)^{-1/2}$ as $x \rightarrow \infty$, $f(x) = x/\log x$ and $g(x) = 1 + (\log x)^{-2}$. Noting that $f'(x) \rightarrow 0$ and $g(x) \rightarrow 1$ as $x \rightarrow \infty$. By Proposition 1.1(a) and Corollary 1.7 of [9], we can choose the norming constants a_n and b_n in such a way that b_n is the solution of the equation

$$2\pi (\log b_n)^2 \exp((\log b_n)^2) = n^2 \tag{1.7}$$

and a_n satisfies

$$a_n = f(b_n) = \frac{b_n}{\log b_n}, \tag{1.8}$$

then

$$\lim_{n \rightarrow \infty} P(M_n \leq a_n x + b_n) = \lim_{n \rightarrow \infty} F^n(a_n x + b_n) = \Lambda(x). \tag{1.9}$$

The aim of this paper is to prove that the uniform convergence rate of (1.9) is proportional to $1/(\log n)^{1/2}$. However, for $F^n(\alpha_n x + \beta_n)$ the convergence rate is no better than $(\log \log n)^2/(\log n)^{1/2}$ even though $\alpha_n/a_n \rightarrow 1$ and $(\beta_n - b_n)/a_n \rightarrow 0$ as $n \rightarrow \infty$. The main results are stated as follows:

Theorem 1. Let $\{X_n, n \geq 1\}$ denote a sequence of independent identically distributed random variables with common distribution F which is lognormal distribution.

(i) For norming constants a_n and b_n given respectively by (1.7) and (1.8), then there exist absolute constants $0 < C_1 < C_2$ such that

$$\frac{C_1}{(\log n)^{1/2}} < \sup_{x \in \mathbb{R}} |F^n(a_n x + b_n) - \Lambda(x)| < \frac{C_2}{(\log n)^{1/2}}$$

for $n \geq 2$.

(ii) For norming constants α_n and β_n respectively given by (1.4) and (1.6), we have

$$F^n(\alpha_n x + \beta_n) - \Lambda(x) \sim -\frac{e^{-x} \exp(-e^{-x}) (\log \log n)^2}{8 (2 \log n)^{1/2}}$$

for large n .

2. Auxiliary lemmas

In this section we provide several key properties of lognormal distribution which are needed for the proofs of our main results. The first one is the distributional tail decomposition of lognormal distribution, which is stated as follows.

Lemma 1. *Let F denote the lognormal distribution function. For $x > 1$, we have*

$$1 - F(x) = \frac{1}{\sqrt{2\pi}} (\log x)^{-1} \exp\left(-\frac{(\log x)^2}{2}\right) - \gamma(x) \tag{2.1}$$

$$= \frac{1}{\sqrt{2\pi}} (\log x)^{-1} \exp\left(-\frac{(\log x)^2}{2}\right) (1 - (\log x)^{-2}) + s(x), \tag{2.2}$$

where

$$0 < \gamma(x) = \frac{1}{\sqrt{2\pi}} \int_x^\infty (\log t)^{-2} \exp\left(-\frac{(\log t)^2}{2}\right) \frac{1}{t} dt < \frac{1}{\sqrt{2\pi}} (\log x)^{-3} \exp\left(-\frac{(\log x)^2}{2}\right) \tag{2.3}$$

and

$$0 < s(x) = \frac{3}{\sqrt{2\pi}} \int_x^\infty (\log t)^{-4} \exp\left(-\frac{(\log t)^2}{2}\right) \frac{1}{t} dt < \frac{3}{\sqrt{2\pi}} (\log x)^{-5} \exp\left(-\frac{(\log x)^2}{2}\right). \tag{2.4}$$

Proof. By integration by parts we have

$$\begin{aligned} 1 - F(x) &= \frac{1}{\sqrt{2\pi}} (\log x)^{-1} \exp\left(-\frac{(\log x)^2}{2}\right) - \frac{1}{\sqrt{2\pi}} \int_x^\infty (\log t)^{-2} \exp\left(-\frac{(\log t)^2}{2}\right) \frac{1}{t} dt \\ &= \frac{1}{\sqrt{2\pi}} (\log x)^{-1} \exp\left(-\frac{(\log x)^2}{2}\right) - \gamma(x), \end{aligned}$$

which is (2.1). Similarly,

$$\gamma(x) = \frac{1}{\sqrt{2\pi}} (\log x)^{-3} \exp\left(-\frac{(\log x)^2}{2}\right) - s(x).$$

Putting it into (2.1), we obtain (2.2), where

$$\begin{aligned} s(x) &= \frac{3}{\sqrt{2\pi}} \int_x^\infty (\log t)^{-4} \exp\left(-\frac{(\log t)^2}{2}\right) \frac{1}{t} dt \\ &= \frac{3}{\sqrt{2\pi}} (\log x)^{-5} \exp\left(-\frac{(\log x)^2}{2}\right) - \frac{15}{\sqrt{2\pi}} \int_x^\infty (\log t)^{-6} \exp\left(-\frac{(\log t)^2}{2}\right) \frac{1}{t} dt \\ &< \frac{3}{\sqrt{2\pi}} (\log x)^{-5} \exp\left(-\frac{(\log x)^2}{2}\right). \end{aligned}$$

The proof is complete. \square

By Lemma 1, we can derive the Mills ratio and the distributional tail representation of lognormal distribution, which are helpful, just as mentioned in Section 1, to find the optimal normalized constants a_n and b_n given by (1.7) and (1.8). The following result is about the Mills-type ratio and distributional tail representation of lognormal distribution, which is a special case of tail behavior and Mills ratio of logarithmic general error distribution studied by Liao and Peng [10].

Lemma 2. *Let $F(x)$ denote the distribution of lognormal random variable with density $F'(x)$, then*

(i) *for $x > 1$ we have*

$$\frac{x F'(x)}{\log x} \left(1 - \frac{1}{(\log x)^2}\right) < 1 - F(x) < \frac{x}{\log x} F'(x)$$

and

$$\frac{1 - F(x)}{F'(x)} \sim \frac{x}{\log x} \tag{2.5}$$

as $x \rightarrow \infty$;

(ii) for large x we have

$$1 - F(x) = c(x) \exp\left(-\int_e^x \frac{g(t)}{f(t)} dt\right) \tag{2.6}$$

with $c(x) \rightarrow (2\pi e)^{-1/2}$, $g(x) = 1 + (\log x)^{-2} \rightarrow 1$ as $x \rightarrow \infty$ and the auxiliary function $f(x) = x(\log x)^{-1}$ satisfies $f'(x) \rightarrow 0$ as $x \rightarrow \infty$.

Let the norming constants a_n and b_n are defined by (1.7) and (1.8) respectively. Let

$$a_n^* = a_n r_n, \quad b_n^* = b_n + \delta_n a_n \tag{2.7}$$

with $r_n \rightarrow 1$ and $\delta_n \rightarrow 0$ as $n \rightarrow \infty$. Then $a_n^*/a_n \rightarrow 1$, and $(b_n^* - b_n)/a_n \rightarrow 0$ which implies $F^n(a_n^*x + b_n^*) \rightarrow \Lambda(x)$. Since $a_n^*x + b_n^* \rightarrow \infty$, for large n we have the following expansion.

Lemma 3. Let a_n^* and b_n^* be defined by (2.7). For fixed $x \in R$ and sufficiently large n ,

$$F^n(a_n^*x + b_n^*) - \Lambda(x) = \Lambda(x)e^{-x} \left((r_n - 1)x + \delta_n - \frac{1}{2}a_n b_n^{-1} (r_n x + \delta_n)^2 + O\left((r_n - 1)^2 + \delta_n^2 + (a_n b_n^{-1})^2\right) \right). \tag{2.8}$$

Proof. Note by (1.7) that $\log b_n \sim (2 \log n)^{1/2}$, which implies that

$$a_n b_n^{-1} = \frac{1}{\log b_n} \sim (2 \log n)^{-1/2} \rightarrow 0$$

as $n \rightarrow \infty$ by virtue of (1.8). Notice

$$\log(1 + a_n b_n^{-1} (r_n x + \delta_n)) = a_n b_n^{-1} (r_n x + \delta_n) - \frac{1}{2} (a_n b_n^{-1})^2 (r_n x + \delta_n)^2 + O\left((a_n b_n^{-1})^3\right),$$

implying

$$(a_n b_n^{-1})^{-1} \log(1 + a_n b_n^{-1} (r_n x + \delta_n)) = r_n x + \delta_n - \frac{1}{2} a_n b_n^{-1} (r_n x + \delta_n)^2 + O\left((a_n b_n^{-1})^2\right)$$

and

$$(\log(1 + a_n b_n^{-1} (r_n x + \delta_n)))^2 = (a_n b_n^{-1})^2 (r_n x + \delta_n)^2 + O\left((a_n b_n^{-1})^3\right),$$

so we have

$$\begin{aligned} & \exp\left(- (a_n b_n^{-1})^{-1} \log(1 + a_n b_n^{-1} (r_n x + \delta_n)) - \frac{1}{2} (\log(1 + a_n b_n^{-1} (r_n x + \delta_n)))^2\right) \\ &= e^{-x} \left(1 - (r_n - 1)x - \delta_n + \frac{1}{2} a_n b_n^{-1} (r_n x + \delta_n)^2 + O\left((a_n b_n^{-1})^2 + \delta_n^2 + (r_n - 1)^2\right) \right) \end{aligned}$$

and

$$(1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1} (r_n x + \delta_n)))^{-1} = 1 - (a_n b_n^{-1})^2 (r_n x + \delta_n) + O\left((a_n b_n^{-1})^3\right).$$

Hence,

$$\begin{aligned} & (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1} (r_n x + \delta_n)))^{-1} \\ & \times \exp\left(- (a_n b_n^{-1})^{-1} \log(1 + a_n b_n^{-1} (r_n x + \delta_n)) - \frac{1}{2} (\log(1 + a_n b_n^{-1} (r_n x + \delta_n)))^2\right) \\ &= \left(1 - (a_n b_n^{-1})^2 (r_n x + \delta_n) + O\left((a_n b_n^{-1})^3\right) \right) e^{-x} \\ & \times \left(1 - (r_n - 1)x - \delta_n + \frac{1}{2} a_n b_n^{-1} (r_n x + \delta_n)^2 + O\left((a_n b_n^{-1})^2 + \delta_n^2 + (r_n - 1)^2\right) \right) \\ &= e^{-x} \left(1 - (r_n - 1)x - \delta_n + \frac{1}{2} a_n b_n^{-1} (r_n x + \delta_n)^2 + O\left((a_n b_n^{-1})^2 + \delta_n^2 + (r_n - 1)^2\right) \right). \end{aligned}$$

Thus for large n we have

$$\begin{aligned} & \frac{1}{\sqrt{2\pi}} (\log (a_n^*x + b_n^*))^{-1} \exp \left(-\frac{1}{2} (\log (a_n^*x + b_n^*))^2 \right) \\ &= \frac{1}{\sqrt{2\pi}} (\log b_n (1 + a_n b_n^{-1} (r_n x + \delta_n)))^{-1} \exp \left(-\frac{1}{2} (\log b_n (1 + a_n b_n^{-1} (r_n x + \delta_n)))^2 \right) \\ &= n^{-1} (1 + a_n b_n^{-1} \log (1 + a_n b_n^{-1} (r_n x + \delta_n)))^{-1} \\ &\quad \times \exp \left(- (a_n b_n^{-1})^{-1} \log (1 + a_n b_n^{-1} (r_n x + \delta_n)) - \frac{1}{2} (\log (1 + a_n b_n^{-1} (r_n x + \delta_n)))^2 \right) \\ &= n^{-1} e^{-x} \left(1 - (r_n - 1)x - \delta_n + \frac{1}{2} a_n b_n^{-1} (r_n x + \delta_n)^2 + O \left((a_n b_n^{-1})^2 + \delta_n^2 + (r_n - 1)^2 \right) \right). \end{aligned} \tag{2.9}$$

Similarly, one can get

$$(\log (a_n^*x + b_n^*))^{-2} = (a_n b_n^{-1})^2 + O \left((a_n b_n^{-1})^4 \right) \tag{2.10}$$

and

$$s(x) = O \left(n^{-1} (a_n b_n^{-1})^4 \right) \tag{2.11}$$

by virtue of (2.4).

Using (2.9)–(2.11) together with (2.2) and (2.4) we have

$$\begin{aligned} & F^n (a_n^*x + b_n^*) - \Lambda(x) \\ &= \left(1 - n^{-1} e^{-x} \left(1 - (r_n - 1)x - \delta_n + \frac{1}{2} a_n b_n^{-1} (r_n x + \delta_n)^2 + O \left((r_n - 1)^2 + \delta_n^2 + (a_n b_n^{-1})^2 \right) \right) \right. \\ &\quad \left. \times \left(1 - (a_n b_n^{-1})^2 + O \left((a_n b_n^{-1})^4 \right) \right) + O \left(n^{-1} (a_n b_n^{-1})^4 \right) \right)^n - \Lambda(x) \\ &= \Lambda(x) e^{-x} \left((r_n - 1)x + \delta_n - \frac{1}{2} a_n b_n^{-1} (r_n x + \delta_n)^2 + O \left((r_n - 1)^2 + \delta_n^{-1} + (a_n b_n^{-1})^2 \right) \right), \end{aligned}$$

which completes the proof. \square

3. Proofs of the main results

The aim of this section is to prove the main results.

Proof of Theorem 1. (i) Letting $r_n = 1, \delta_n = 0$ in (2.7) and noting $a_n b_n^{-1} \sim 1 / (2 \log n)^{1/2}$, by Lemma 3 we can prove that there exists an absolute constant $C_1 > 0$ such that

$$\sup_{x \in \mathbb{R}} |F^n(a_n x + b_n) - \Lambda(x)| > \frac{C_1}{(\log n)^{1/2}}$$

for $n \geq 2$. In order to obtain the upper bound, we need to prove the following.

$$\sup_{d_n \leq x < \infty} |F^n(a_n x + b_n) - \Lambda(x)| < \mathbb{D}_1 a_n b_n^{-1}, \tag{3.1}$$

$$\sup_{0 \leq x < d_n} |F^n(a_n x + b_n) - \Lambda(x)| < \mathbb{D}_2 a_n b_n^{-1}, \tag{3.2}$$

$$\sup_{-c_n \leq x < 0} |F^n(a_n + b_n) - \Lambda(x)| < \mathbb{D}_3 a_n b_n^{-1} \tag{3.3}$$

$$\sup_{-\infty < x < -c_n} |F^n(a_n x + b_n) - \Lambda(x)| < \mathbb{D}_4 a_n b_n^{-1} \tag{3.4}$$

for $n \geq n_0$ as there must exist absolute constant $C_2 > C_1$ such that $n_0 = \sup(k : C_2 / (\log k)^{1/2} \geq 1)$, where $\mathbb{D}_i > 0$ for $i = 1, 2, 3, 4$ are absolute constants, and c_n and d_n are respectively defined by

$$c_n =: \log \log \log b_n > 0, \quad d_n =: -\log \log \frac{\log b_n}{\log b_n - 1} > 0$$

for $n \geq n_0$. Note that $x \geq -c_n$ implies

$$b_n + a_n x \geq b_n - a_n c_n = b_n \left(1 - \frac{\log \log \log b_n}{\log b_n} \right) > 0, \quad n \geq n_0.$$

For the rest of the proof, let $C_j, j = 1, 2, \dots, 12$ stand for the absolute positive constants.

For $x \geq -c_n$, let $\Psi_n(x) = 1 - F(a_n x + b_n)$, then

$$n \log F(a_n x + b_n) = -n\Psi_n(x) - R_n(x), \tag{3.5}$$

where

$$0 < R_n(x) = -n (\Psi_n(x) + \log(1 - \Psi_n(x))) < \frac{n\Psi_n^2(x)}{2(1 - \Psi_n(x))}$$

since

$$-x - \frac{x^2}{2(1-x)} < \log(1-x) < -x \quad \text{for } 0 < x < 1. \tag{3.6}$$

We deduce from (2.1) that for $x \geq -c_n$,

$$\begin{aligned} \Psi_n(x) &\leq \Psi_n(-c_n) = 1 - F(b_n - a_n c_n) \\ &< n^{-1} (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-1} \exp\left(- (a_n b_n^{-1})^{-1} \log(1 - a_n b_n^{-1} c_n)\right) \\ &< n^{-1} (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-1} \exp\left(c_n + \frac{a_n b_n^{-1} c_n^2}{2(1 - a_n b_n^{-1} c_n)}\right) \\ &< \frac{\log(2 \log n)}{2n} \left(1 + \frac{\log\left(1 - \frac{\log \log \log b_n}{\log b_n}\right)}{\log b_n} \right)^{-1} \exp\left(\frac{\frac{(\log \log \log b_n)^2}{\log b_n}}{2\left(1 - \frac{\log \log \log b_n}{\log b_n}\right)}\right) \\ &< C_1 < 1 \end{aligned} \tag{3.7}$$

for $n \geq n_0$, implying

$$\inf_{x \geq -c_n} (1 - \Psi_n(x)) > 1 - C_1 > 0,$$

so,

$$\begin{aligned} 0 < R_n(x) &\leq \frac{n\Psi_n^2(x)}{2(1 - \Psi_n(x))} \\ &< \frac{a_n b_n^{-1}}{2\sqrt{2\pi}(1 - C_1)} \left(1 + \frac{\log\left(1 - \frac{\log \log \log b_n}{\log b_n}\right)}{\log b_n} \right)^{-2} \frac{(\log \log b_n)^2}{\exp\left(\frac{1}{2}(\log b_n)^2\right)} \exp\left(\frac{\frac{(\log \log \log b_n)^2}{\log b_n}}{\left(1 - \frac{\log \log \log b_n}{\log b_n}\right)}\right) \\ &< C_2 a_n b_n^{-1} \end{aligned}$$

for $n \geq n_0$. Thus

$$|\exp(-R_n(x)) - 1| < R_n(x) < C_2 a_n b_n^{-1} \tag{3.8}$$

as $1 - e^{-x} < x$ for $x > 0$.

Let $A_n(x) = \exp(-n\Psi_n(x) + e^{-x})$ and $B_n(x) = \exp(-R_n(x))$, by (3.8) we have

$$\begin{aligned} |F^n(a_n x + b_n) - \Lambda(x)| &\leq \Lambda(x) |A_n(x) - 1| + |B_n(x) - 1| \\ &< \Lambda(x) |A_n(x) - 1| + C_2 a_n b_n^{-1}. \end{aligned} \tag{3.9}$$

From (2.1) and (2.3) it follows that

$$-n\Psi_n(x) + e^{-x} = \left(1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1} x) \right)^{-1} e^{-x} C_n(x) \tag{3.10}$$

with

$$C_n(x) = -D_n(x)E_n(x) + F_n(x), \tag{3.11}$$

where $D_n(x)$, $E_n(x)$ and $F_n(x)$ are respectively defined by

$$\begin{aligned} D_n(x) &= \exp\left(x - \frac{1}{2}(\log(1 + a_n b_n^{-1}x))^2 - (a_n b_n^{-1})^{-1} \log(1 + a_n b_n^{-1}x)\right), \\ E_n(x) &= 1 - (a_n b_n^{-1})^2 (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1}x))^{-2} \kappa_n, \\ F_n(x) &= 1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1}x), \end{aligned}$$

where $0 < \kappa_n < 1$.

Noting that $1 + x < e^x$, $x \in \mathbb{R}$ and $\log(1 + x) < x$, $x > -1$, for $x \geq -c_n$ we have

$$\begin{aligned} D_n(x) &> 1 + x - \frac{1}{2}(\log(1 + a_n b_n^{-1}x))^2 - (a_n b_n^{-1})^{-1} \log(1 + a_n b_n^{-1}x) \\ &> 1 - \frac{1}{2}(\log(1 + a_n b_n^{-1}x))^2. \end{aligned} \tag{3.12}$$

First we prove (3.3). Noting that $\log(1 + a_n b_n^{-1}x) < 0$ in the case of $-c_n \leq x < 0$ and $(\log(1 + x))^2 < \left(x - \frac{x^2}{2(1+x)}\right)^2$ as $-1 < x < 0$, by (3.12) we have

$$C_n(x) < a_n b_n^{-1} \left(\frac{1}{2} a_n b_n^{-1} \left(x - \frac{a_n b_n^{-1} x^2}{2(1 + a_n b_n^{-1}x)} \right)^2 + a_n b_n^{-1} (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1}x))^{-2} \right) \tag{3.13}$$

for $-c_n \leq x < 0$. On the other hand, since $e^x - 1 < x e^x$, $x \in \mathbb{R}$ and by (3.6) we have

$$D_n(x) < 1 + \frac{a_n b_n^{-1} x^2}{2(1 + a_n b_n^{-1}x)} \exp\left(\frac{a_n b_n^{-1} x^2}{2(1 + a_n b_n^{-1}x)}\right),$$

so,

$$C_n(x) > a_n b_n^{-1} \left(-\frac{x^2}{2(1 + a_n b_n^{-1}x)} \exp\left(\frac{a_n b_n^{-1} x^2}{2(1 + a_n b_n^{-1}x)}\right) + a_n b_n^{-1} x - \frac{(a_n b_n^{-1}x)^2}{2(1 + a_n b_n^{-1}x)} \right). \tag{3.14}$$

Combining (3.13) with (3.14), for $-c_n \leq x < 0$ we have

$$\begin{aligned} |C_n(x)| &< a_n b_n^{-1} \left(\frac{1}{2} a_n b_n^{-1} \left(x - \frac{a_n b_n^{-1} x^2}{2(1 + a_n b_n^{-1}x)} \right)^2 + a_n b_n^{-1} (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1}x))^{-2} \right. \\ &\quad \left. + \frac{x^2}{2(1 + a_n b_n^{-1}x)} \exp\left(\frac{a_n b_n^{-1} x^2}{2(1 + a_n b_n^{-1}x)}\right) - a_n b_n^{-1} x + \frac{(a_n b_n^{-1}x)^2}{2(1 + a_n b_n^{-1}x)} \right) \\ &< a_n b_n^{-1} \left(\frac{1}{8} a_n b_n^{-1} c_n^2 \left(1 + \frac{1}{1 - a_n b_n^{-1} c_n} \right)^2 + a_n b_n^{-1} (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-2} \right. \\ &\quad \left. + \frac{x^2}{2(1 - a_n b_n^{-1} c_n)} \exp\left(\frac{a_n b_n^{-1} c_n^2}{2(1 - a_n b_n^{-1} c_n)}\right) + a_n b_n^{-1} c_n + \frac{(a_n b_n^{-1} c_n)^2}{2(1 - a_n b_n^{-1} c_n)} \right) \\ &< a_n b_n^{-1} (C_3 + C_4 x^2) \end{aligned}$$

for $n \geq n_0$. So for $-c_n \leq x < 0$ we have

$$\begin{aligned} |-n\Psi_n(x) + e^{-x}| &= (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1}x))^{-1} e^{-x} |C_n(x)| \\ &< (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1}x))^{-1} e^{-x} a_n b_n^{-1} (C_3 + C_4 x^2) \\ &< (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-1} e^{-x} a_n b_n^{-1} (C_3 + C_4 x^2) \\ &< (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-1} e^{c_n} a_n b_n^{-1} (C_3 + C_4 c_n^2) \end{aligned}$$

$$\begin{aligned}
 &= (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-1} \left(\mathbb{C}_3 \frac{\log \log b_n}{\log b_n} + \mathbb{C}_4 \frac{\log \log b_n (\log \log \log b_n)^2}{\log b_n} \right) \\
 &< \left(1 + \frac{\log \left(1 - \frac{\log \log \log b_n}{\log b_n} \right)}{\log b_n} \right)^{-1} \left(\mathbb{C}_3 \frac{\log \log b_n}{\log b_n} + \mathbb{C}_4 \frac{(\log \log b_n)^3}{\log b_n} \right) \\
 &< \mathbb{C}_5
 \end{aligned}$$

for $n \geq n_0$. Hence noting $|e^x - 1| < |x|(e^x + 1)$, $x \in \mathbb{R}$, we have

$$\begin{aligned}
 \Lambda(x) |A_n(x) - 1| &= \Lambda(x) |\exp(-n\Psi_n(x) + e^{-x}) - 1| \\
 &< \Lambda(x) |-n\Psi_n(x) + e^{-x}| (\exp(-n\Psi_n(x) + e^{-x}) + 1) \\
 &< (e^{\mathbb{C}_5} + 1) \Lambda(x) (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-1} e^{-x} a_n b_n^{-1} (\mathbb{C}_3 + \mathbb{C}_4 x^2) \\
 &< (e^{\mathbb{C}_5} + 1) e^{-1} a_n b_n^{-1} (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-1} (\mathbb{C}_3 + \mathbb{C}_4 x^2) \exp\left(-\frac{x^2}{2}\right) \\
 &< \mathbb{C}_6 a_n b_n^{-1}
 \end{aligned}$$

for $n \geq n_0$, since $\max_{-c_n \leq x \leq 0} (\mathbb{C}_3 + \mathbb{C}_4 x^2) e^{-x^2/2} \leq \mathbb{C}_3 + \mathbb{C}_4$. Combining above inequality with (3.9) we have

$$\sup_{-c_n \leq x < 0} |F^n(a_n x + b_n) - \Lambda(x)| < (\mathbb{C}_2 + \mathbb{C}_6) a_n b_n^{-1}$$

for $n > n_0$.

Second, consider the case in which $0 \leq x < d_n$. By (3.6), (3.10) and (3.12) we can derive

$$\mathbb{C}_n(x) < a_n b_n^{-1} \left(\frac{1}{2} a_n b_n^{-1} x^2 + a_n b_n^{-1} (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1} x))^{-2} + a_n b_n^{-1} x \right). \tag{3.15}$$

On the other hand, note that $x - x^2/2 < \log(1 + x)$, $x > 0$ and $e^x - 1 < x e^x$, $x \in \mathbb{R}$, implies

$$\begin{aligned}
 D_n(x) &< \exp\left(x - (a_n b_n^{-1})^{-1} \log(1 + a_n b_n^{-1} x)\right) \\
 &< 1 + \frac{a_n b_n^{-1} x^2}{2} \exp\left(\frac{a_n b_n^{-1} x^2}{2}\right),
 \end{aligned}$$

so,

$$\begin{aligned}
 \mathbb{C}_n(x) &> -1 - \frac{a_n b_n^{-1} x^2}{2} \exp\left(\frac{a_n b_n^{-1} x^2}{2}\right) + 1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1} x) \\
 &> -\frac{a_n b_n^{-1} x^2}{2} \exp\left(\frac{a_n b_n^{-1} x^2}{2}\right)
 \end{aligned} \tag{3.16}$$

as $x \geq 0$. Combining (3.15) with (3.16), for $0 \leq x < d_n$ we can get

$$|\mathbb{C}_n(x)| < a_n b_n^{-1} (\mathbb{C}_7 + \mathbb{C}_8 x^2)$$

for $n \geq n_0$, implying

$$\begin{aligned}
 |-n\Psi_n(x) + e^{-x}| &< (1 + a_n b_n^{-1} \log(1 + a_n b_n^{-1} x))^{-1} e^{-x} a_n b_n^{-1} (\mathbb{C}_7 + \mathbb{C}_8 x^2) \\
 &< e^{-x} a_n b_n^{-1} (\mathbb{C}_7 + \mathbb{C}_8 x^2) \\
 &< a_n b_n^{-1} (\mathbb{C}_7 + 4\mathbb{C}_8 e^{-2}) \\
 &< \mathbb{C}_9.
 \end{aligned}$$

Hence by $|e^x - 1| < |x|(e^x + 1)$, $x \in \mathbb{R}$,

$$\begin{aligned}
 \Lambda(x) |A_n(x) - 1| &= \Lambda(x) |\exp(-n\Psi_n(x) + e^{-x}) - 1| \\
 &< \Lambda(x) |-n\Psi_n(x) + e^{-x}| (\exp(-n\Psi_n(x) + e^{-x}) + 1) \\
 &< (e^{\mathbb{C}_9} + 1) \Lambda(x) e^{-x} a_n b_n^{-1} (\mathbb{C}_7 + \mathbb{C}_8 x^2) \\
 &= (e^{\mathbb{C}_9} + 1) a_n b_n^{-1} (\mathbb{C}_7 + \mathbb{C}_8 x^2) \exp(-e^{-x} - x)
 \end{aligned}$$

$$\begin{aligned} &< (e^{C_9} + 1) a_n b_n^{-1} (C_7 + C_8 x^2) \exp\left(-1 - \frac{1}{2} x^2\right) \\ &\leq (e^{C_9} + 1) e^{-1} (C_7 + 2C_8 e^{-1}) a_n b_n^{-1} \\ &= C_{10} a_n b_n^{-1} \end{aligned}$$

for $n \geq n_0$ since $0 \leq x < d_n$. Combining with (3.9) to get (3.2).

Third we prove (3.1). Obviously,

$$\sup_{x \geq d_n} (1 - \Lambda(x)) \leq 1 - \Lambda(d_n) = a_n b_n^{-1}, \tag{3.17}$$

and from (3.5) we deduce that

$$1 - F^n(a_n d_n + b_n) < n \Psi_n(d_n) + R_n(d_n). \tag{3.18}$$

By (2.1) we have

$$\begin{aligned} n \Psi_n(d_n) &< \frac{n}{\sqrt{2\pi}} (\log(a_n d_n + b_n))^{-1} \exp\left(-\frac{1}{2} (\log(a_n d_n + b_n))^2\right) \\ &< \exp\left(- (a_n b_n^{-1})^{-1} \log(1 + a_n b_n^{-1} d_n)\right) \\ &< \exp\left(-d_n + \frac{a_n b_n^{-1} d_n^2}{2}\right) \\ &= a_n b_n^{-1} (\log b_n) \left(\log \frac{\log b_n}{\log b_n - 1}\right) \exp\left(\frac{\left(\log \log \frac{\log b_n}{\log b_n - 1}\right)^2}{2 \log b_n}\right) \\ &< C_{11} a_n b_n^{-1} \end{aligned} \tag{3.19}$$

for $n \geq n_0$ since the continuous function $x \log(x/(x-1)) \exp\left(\frac{(\log \log \frac{x}{x-1})^2}{2x}\right)$ is decreasing as $x > 2$. Note that $R_n(x) < C_2 a_n b_n^{-1}$ by (3.7). Combining with (3.17), (3.18) and (3.19), we derive that

$$\begin{aligned} \sup_{x \geq d_n} |F^n(a_n x + b_n) - \Lambda(x)| &\leq 1 - F^n(a_n d_n + b_n) + 1 - \Lambda(d_n) \\ &< (C_{11} + C_2 + 1) a_n b_n^{-1} \end{aligned}$$

for $n \geq n_0$ which completes the proof of (3.1).

Finally, consider the case in which $-\infty < x \leq -c_n$. In this case,

$$\begin{aligned} \sup_{x \leq -c_n} F^n(a_n x + b_n) &\leq F^n(b_n - a_n c_n) \\ &\leq \exp\left(-e^{c_n} \left(1 - \frac{1}{2} (\log(1 - a_n b_n^{-1} c_n))^2 - (a_n b_n^{-1})^2 (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-2}\right)\right) \\ &\leq a_n b_n^{-1} \exp\left(\frac{(\log \log b_n) c_n^2}{2 (\log b_n)^2} + \frac{1}{2 (1 - a_n b_n^{-1} c_n)} \frac{(\log \log b_n) c_n^3}{(\log b_n)^3}\right. \\ &\quad \left. + \frac{1}{8 (1 - a_n b_n^{-1} c_n)^2} \frac{(\log \log b_n) c_n^4}{(\log b_n)^4} + \frac{\log \log b_n}{(\log b_n)^2} (1 + a_n b_n^{-1} \log(1 - a_n b_n^{-1} c_n))^{-2}\right) \\ &< C_{12} a_n b_n^{-1} \end{aligned}$$

for $n \geq n_0$. Thus we have

$$\begin{aligned} \sup_{x < -c_n} |F^n(a_n x + b_n) - \Lambda(x)| &\leq F^n(b_n - a_n c_n) + \Lambda(-c_n) \\ &< (C_{12} + 1) a_n b_n^{-1} \end{aligned}$$

for $n \geq n_0$. This completes the proof of (3.4). The proof of Theorem 1(i) is complete.

(ii) Setting

$$\begin{aligned} b_n &= \beta_n + \theta_n \\ &= \left(\exp((2 \log n)^{1/2}) \right) \left(1 - \frac{B_n}{2(2 \log n)^{1/2}} + \frac{\theta_n}{\exp((2 \log n)^{1/2})} \right), \end{aligned} \quad (3.20)$$

where β_n is defined by (1.6), $B_n = \log 4\pi + \log \log n$ and $\theta_n = \exp((2 \log n)^{1/2}) o(1/(\log n)^{1/2})$. Note that

$$\log(1+x) = x - \frac{x^2}{2} + O(x^3) \quad \text{as } x \rightarrow 0.$$

Taking logarithms of (3.20), we have

$$\begin{aligned} \log b_n &= (2 \log n)^{1/2} \left(1 - \frac{B_n}{4 \log n} + \frac{\theta_n}{(2 \log n)^{1/2} \exp((2 \log n)^{1/2})} \right. \\ &\quad \left. - \frac{1}{2(2 \log n)^{1/2}} \left(\frac{B_n}{2(2 \log n)^{1/2}} - \frac{\theta_n}{\exp((2 \log n)^{1/2})} \right)^2 + O\left(\frac{B_n^3}{(\log n)^2}\right) \right). \end{aligned}$$

Hence,

$$\begin{aligned} (\log b_n)^2 &= 2 \log n - B_n + \frac{B_n^2}{8 \log n} - \frac{B_n \theta_n}{(2 \log n)^{1/2} \exp((2 \log n)^{1/2})} + \frac{(1 - (2 \log n)^{1/2}) \theta_n^2}{\exp(2(2 \log n)^{1/2})} \\ &\quad + \frac{(2(2 \log n)^{1/2} + B_n) \theta_n}{\exp((2 \log n)^{1/2})} - \frac{B_n^2}{4(2 \log n)^{1/2}} + O\left(\frac{B_n^3}{\log n}\right) \end{aligned}$$

and

$$\begin{aligned} \log \log b_n &= \frac{1}{2} \log 2 + \frac{1}{2} \log \log n - \frac{B_n}{4 \log n} + \frac{\theta_n}{(2 \log n)^{1/2} \exp((2 \log n)^{1/2})} - \frac{B_n^2}{32(\log n)^2} \\ &\quad - \frac{\theta_n^2}{4(\log n) \exp(2(2 \log n)^{1/2})} + \frac{B_n \theta_n}{2(2 \log n)^{3/2} \exp((2 \log n)^{1/2})} + O\left(\frac{B_n^2}{(\log n)^{3/2}}\right). \end{aligned}$$

Putting above expansions of $(\log b_n)^2$ and $\log \log b_n$ into

$$\log 2\pi + 2 \log \log b_n + (\log b_n)^2 = 2 \log n, \quad (3.21)$$

we have

$$\theta_n \sim \frac{\exp((2 \log n)^{1/2}) B_n^2}{16 \log n}.$$

Hence,

$$b_n = \exp((2 \log n)^{1/2}) \left(1 - \frac{B_n}{2(2 \log n)^{1/2}} + \frac{B_n^2}{16 \log n} + o\left(\frac{(\log \log n)^2}{\log n}\right) \right).$$

Now for α_n and β_n defined by (1.4) and (1.6), one can check that

$$r_n - 1 = \frac{\alpha_n}{a_n} - 1 \sim \frac{\log \log n}{2(2 \log n)^{1/2}},$$

and

$$\delta_n = \frac{\beta_n - b_n}{a_n} \sim -\frac{(\log \log n)^2}{8(2 \log n)^{1/2}}$$

for large n . Hence by Lemma 3 we can derive

$$F^n(\alpha_n x + \beta_n) - \Lambda(x) \sim -\frac{e^{-x} \Lambda(x)}{8} \frac{(\log \log n)^2}{(2 \log n)^{1/2}}$$

for large n , which completes the proof of Theorem 1(ii). \square

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