

PRECISE LARGE DEVIATIONS FOR CONSISTENTLY VARYING-TAILED DISTRIBUTIONS IN THE COMPOUND RENEWAL RISK MODEL

D.G. Konstantinides and F. Loukissas

Department of Statistics and Actuarial-Financial Mathematics, University of the Aegean,
GR-83200 Karlovassi, Samos, Greece

(e-mail: konstant@aegean.gr; floukissas@aegean.gr)

Received October 28, 2009

Abstract. We investigate precise large deviations for heavy-tailed random sums. We prove a general asymptotic relation in the compound renewal risk model for consistently varying-tailed distributions. This model was introduced in [Q. Tang, C. Su, T. Jiang, and J.S. Zang, Large deviation for heavy-tailed random sums in compound renewal model, *Stat. Probab. Lett.*, 52:91–100, 2001] as a more practical risk model. The proof is based on the inequality found in [D. Fuk and S.V. Nagaev, Probability for sums of independent random variables, *Theory Probab. Appl.*, 16:600–675, 1971].

MSC: 60F10, 60F05, 60K05

Keywords: heavy tails, random sums, precise large deviations, renewal model, compound renewal model

1 INTRODUCTION

1.1 Heavy-tailed distributions

In this paper, we are interesting in probabilities of precise large deviations in the compound renewal risk model. Let us consider a sequence $\{X_n, n \geq 1\}$ of i.i.d. r.v.'s with common distribution function F and finite mean μ . We write $F(x) = \mathbf{P}(X \leq x)$ and denote the tail by $\bar{F} = 1 - F$. We suppose that the sequence is independent of a nonnegative renewal counting process $\{N(t), t \geq 0\}$ representing the claim arrival process. We denote $\lambda(t) = \mathbf{E}N(t) < \infty$ for all $0 \leq t < \infty$ and assume that $\lambda(t) \rightarrow \infty$ as $t \rightarrow \infty$. We say that X (or its distribution F) is heavy tailed if it has no exponential moments. We recall some interesting subclasses of subexponential distributions.

- The class $\mathcal{R}_{-\alpha}$: A distribution function F with support on $[0, \infty)$ belongs to $\mathcal{R}_{-\alpha}$ if

$$\lim_{x \rightarrow \infty} \frac{\bar{F}(xy)}{\bar{F}(x)} = y^{-\alpha}, \quad y > 1,$$

for some $\alpha > 1$. Such a distribution function F is said to have a regularly varying tail.

- The class ERV: A distribution function F with support on $[0, \infty)$ belongs to ERV $(-\alpha, -\beta)$ if

$$y^{-\beta} \leq \liminf_{x \rightarrow \infty} \frac{\bar{F}(xy)}{\bar{F}(x)} \leq y^{-\alpha}, \quad y > 1,$$

for some $0 < \alpha \leq \beta < \infty$. In this case, the distribution function F is said to have an extended regularly varying tail.

- *The class \mathcal{C}* : A distribution function F with support on $[0, \infty)$ belongs to \mathcal{C} if

$$\lim_{y \searrow 1} \liminf_{x \rightarrow \infty} \frac{\overline{F}(xy)}{\overline{F}(x)} = 1$$

or, equivalently,

$$\lim_{y \nearrow 1} \limsup_{x \rightarrow \infty} \frac{\overline{F}(xy)}{\overline{F}(x)} = 1.$$

Now the distribution function F is said to have a consistently varying tail. Some results on the class \mathcal{C} can be found in [2, 4].

- *The class \mathcal{D}* : A distribution function F with support on $[0, \infty)$ belongs to \mathcal{D} if

$$\limsup_{x \rightarrow \infty} \frac{\overline{F}(xy)}{\overline{F}(x)} < \infty$$

for any $0 \leq y \leq 1$ (or, equivalently, for $y = 1/2$). Such a distribution function F is said to have a dominatedly varying tail.

The following inclusions are well known: for any $0 < \alpha \leq \gamma \leq \beta < \infty$,

$$\mathcal{R}_{-\gamma} \subset \text{ERV}(-\alpha, -\beta) \subset \mathcal{C} \subset \mathcal{D}.$$

For heavy-tailed distributions, we refer to [5].

For two positive functions $a(\cdot)$ and $b(\cdot)$, we write

$$a(x) \asymp b(x) \quad \text{if } 0 < \liminf_{x \rightarrow \infty} \frac{a(x)}{b(x)} \leq \limsup_{x \rightarrow \infty} \frac{a(x)}{b(x)} < \infty,$$

$$a(x) \lesssim b(x) \quad \text{if } \limsup_{x \rightarrow \infty} \frac{a(x)}{b(x)} \leq 1, \quad \text{and}$$

$$a(x) \gtrsim b(x) \quad \text{if } \liminf_{x \rightarrow \infty} \frac{a(x)}{b(x)} \geq 1.$$

For a distribution function F , the upper Matuszewska index γ_F is defined as follows:

$$\gamma(y) := \liminf_{x \rightarrow \infty} \frac{\overline{F}(xy)}{\overline{F}(x)}, \quad \gamma_F := \inf \left\{ -\frac{\log \gamma(y)}{\log y} : y > 1 \right\}.$$

For more information about the Matuszewska index, see in [1, 4].

1.2 The two risk models

1.2.1 The renewal risk model

This model has the following structure:

1. The individual claim sizes $\{X_n, n \geq 1\}$ are i.i.d. nonnegative r.v.'s with common distribution function F and finite mean $\mu = \mathbf{E}X_1$.

2. The inter-arrival times $\{Y_n, n \geq 1\}$ are i.i.d. nonnegative r.v.'s with finite mean $\mathbf{E}Y_1 = 1/\lambda$, independent of $\{X_n, n \geq 1\}$.
3. The number of claims in the interval $[0, t]$ is denoted by

$$N(t) = \sup\{n \geq 1: T_n \leq t\}, \quad t \geq 0,$$

where $T_n = \sum_{i=1}^n Y_i$.

1.2.2 The compound renewal risk model

In [20], in relation to an application in insurance and finance, there was introduced a more realistic risk model, called *compound renewal model*. It was noticed that, for every accident moment, it is natural to expect more than one individual claim.

That risk model has the following structure:

1. The individual claim sizes $\{X_n, n \geq 1\}$ are i.i.d. nonnegative r.v.'s with common distribution function F and finite mean $\mu = \mathbf{E}X_1$.
2. The inter-arrival times for the accident $\{Y_n, n \geq 1\}$ are i.i.d. nonnegative r.v.'s with finite mean $\mathbf{E}Y_1 = 1/\lambda$, independent of $\{X_n, n \geq 1\}$.
3. The number of accidents in the interval $[0, t]$ is denoted by

$$\tau(t) = \sup\{n \geq 1: T_n \leq t\}$$

for every $t \geq 0$ with $\mathbf{E}(\tau(t)) = \nu(t)$.

The number of individual claims caused by the n th accident is a nonnegative, integer-valued r.v. Z_n , and $\{Z_n, n \geq 1\}$ constitutes a sequence of i.i.d. r.v.'s with common distribution function G and finite mean θ , independent of $\{X_n, n \geq 1\}$ and $\{Y_n, n \geq 1\}$. The total of claims up to time t is given by

$$N(t) = \sum_{i=1}^{\tau(t)} Z_i, \quad t \geq 0, \tag{1.1}$$

with

$$\lambda(t) = \mathbf{E}(N(t)) = \mathbf{E}\left(\sum_{i=1}^{\tau(t)} Z_i\right) = \theta\nu(t).$$

In catastrophe insurance, for instance, $\tau(t)$ is the number of events that occur in the interval $[0, t]$. Such events are natural catastrophe events like windstorms, earthquakes, etc. The sequence $\{Z_n, n \geq 1\}$ represents the number of individual claims caused by the n th catastrophic event.

1.3 Precise large deviations

We consider the asymptotic relation

$$\mathbf{P}(S_n - n\mu > x) \sim n\bar{F}(x) \tag{1.2}$$

as $n \rightarrow \infty$, where

$$S_n = \sum_{i=1}^n X_i \tag{1.3}$$

denotes the n th partial sum of the sequence $\{X_i, i \geq 1\}$.

We recall the following propositions established in the literature on precise large deviations for nonrandom sums.

Proposition 1. (See [14, 15].) *If $F \in \mathcal{R}_{-\alpha}$ with $a > 1$, then for any fixed $\gamma > 0$, relation (1.2) holds uniformly for $x \geq \gamma n$.*

Proposition 2. (See [3].) *If $F \in \text{ERV}(-\alpha, -\beta)$ with $1 < \alpha \leq \beta < \infty$, then for any fixed $\gamma > 0$, relation (1.2) holds uniformly for $x \geq \gamma n$.*

Proposition 3. (See [16].) *If $F \in \mathcal{C}$, then for any fixed $\gamma > 0$, relation (1.2) holds uniformly for $x \geq \gamma n$.*

More results in the field of precise large deviations for nonrandom sums can be found in [8, 9, 7, 12, 13]. A very general treatment of large deviations for subexponentials can be found in [18].

Further, precise large deviations for random sums of the type

$$S_{N(t)} = \sum_{k=1}^{N(t)} X_k, \quad t > 0, \quad (1.4)$$

were investigated by many researchers, where the asymptotic relation (1.2) was transformed into

$$\mathbf{P}(S_{N(t)} - \mu\lambda(t) > x) \sim \lambda(t)\bar{F}(x) \quad (1.5)$$

as $t \rightarrow \infty$, uniformly for $x \geq \gamma\lambda(t)$.

In the sequel, we need the following well-known propositions.

Proposition 4. (See [10, Thm. 3.1].) *Let $F \in \text{ERV}(-\alpha, -\beta)$ for some $1 < \alpha \leq \beta < \infty$. Furthermore, suppose that $\{N(t), t \geq 0\}$ is an integer counting process satisfying*

$$\text{Assumption } N_1: \quad \frac{N(t)}{\lambda(t)} \xrightarrow{\mathbf{P}} 1$$

as $t \rightarrow \infty$ and

$$\text{Assumption } N_2: \quad \sum_{k > (1+\delta)\lambda(t)} (1+\epsilon)^k \mathbf{P}(N(t) = k) = o(1)$$

as $t \rightarrow \infty$ for any $\delta > 0$ and some sufficiently small $\epsilon > 0$. Then, for any fixed $\gamma > 0$, relation (1.5) holds uniformly for $x \geq \gamma\lambda(t)$.

Proposition 5. (See [20, Thm. 2.1].) *Let $F \in \text{ERV}(-\alpha, -\beta)$ for some $1 < \alpha \leq \beta < \infty$. Furthermore, suppose that $\{N(t), t \geq 0\}$ is an integer counting process satisfying*

$$\text{Assumption } A: \quad \mathbf{E}((N(t))^p I_{(N(t) > (1+\delta)\lambda(t))}) = O(\lambda(t))$$

as $t \rightarrow \infty$ for some $p > \beta$ and all $\delta > 0$. Then, for any fixed $\gamma > 0$, relation (1.5) holds uniformly for $x \geq \gamma\lambda(t)$.

Proposition 6. (See [16, Thm. 4.1].) *Let $F \in \mathcal{C}$ and suppose that $\{N(t), t \geq 0\}$ is an integer counting process satisfying Assumption A for some $p > \gamma_F$ and all $\delta > 0$. Then, for any fixed $\gamma > 0$, relation (1.5) holds uniformly for $x \geq \gamma\lambda(t)$.*

Proposition 7. (See [20, Thm. 2.3].) *Suppose that $Z_1 \geq 1$, $F \in \text{ERV}(-\alpha, -\beta)$ for some $1 < \alpha \leq \beta < \infty$, and $\mathbf{E}Z_1^p < \infty$ for some $p > \beta + 2$. Then the compound renewal process $\{N(t), t > 0\}$ is an integer counting process satisfying Assumption A, and, consequently, for any fixed $\gamma > 0$, (1.5) holds uniformly for $x \geq \gamma\lambda(t)$.*

A nice review on recent developments on large deviations for random sums is given in [11]. In the case of independent r.v.'s, the precise large deviation has been established in [17] and [19].

Remark 1. Assumption N_1 has the following equivalent formulation given in [10]: There exists a positive function $\epsilon(t)$ such that $\epsilon(t) \rightarrow 0$ and

$$\mathbf{P}[|N(t) - \lambda(t)| > \epsilon(t)\lambda(t)] = o(1).$$

2 MAIN RESULTS

Theorem 1. Let $\{X_k, k \geq 1\}$ be a sequence of i.i.d. nonnegative r.v.'s with common distribution function $F \in \mathcal{C}$ and finite mean μ , and let $\{Z_i, i \geq 1\}$ be a sequence of i.i.d. integer-valued r.v.'s with $Z_1 \geq 1$, $\mathbf{E}Z_1^p < \infty$ for some $p > \gamma_F + 2$ and common distribution function G with finite mean θ . Also, let $\{N(t), t \geq 0\}$ be a compound renewal process driven by a sequence of i.i.d. nonnegative random variables $\{Y_k, k \geq 1\}$ with finite mean. Suppose that the sequences $\{X_k, k \geq 1\}$, $\{Z_i, i \geq 1\}$, and $\{Y_k, k \geq 1\}$ are mutually independent. Then for any fixed $\gamma > 0$, the asymptotic relation (1.5) holds uniformly for $x \geq \gamma\lambda(t)$.

For the proof, we need the following lemmas.

Lemma 1. (See [16, Lemma 2.1].) For a distribution function $F \in \mathcal{D}$ with finite expectation, we have $1 \leq \gamma_F < \infty$, and, for any $\gamma > \gamma_F$,

$$x^{-\gamma} = o(\overline{F}(x)) \quad \text{as } x \rightarrow \infty.$$

The next lemma presents a reformulation of the standard Fuk and Nagaev inequality.

Lemma 2. (See [16, Lemma 2.3].) Let $\{X_k, k \geq 1\}$ be a sequence of i.i.d. nonnegative random variables with common distribution function F and finite expectation μ . Then, for all $u > 0$, $x > 0$ and $n \geq 1$, we have

$$\mathbf{P}(S_n > x) \leq n\overline{F}\left(\frac{x}{u}\right) + \left(\frac{e\mu n}{x}\right)^u.$$

Lemma 3. (See [20, Lemma 3.5].) Suppose that $\{N(t), t \geq 0\}$ is a renewal counting process. Then, for any positive δ and m , we have

$$\sum_{k > (1+\delta)\lambda(t)} k^m \mathbf{P}(N(t) \geq k) = o(1), \quad t \rightarrow \infty.$$

Remark 2. From [20, Thm. 2.3] we obtain that the compound renewal counting process satisfies Assumption A, and from Theorem 2.1 in the same paper we get that Assumption A implies Assumption N_1 . Hence, the compound renewal counting process satisfies Assumption N_1 .

3 THE PROOF

Let us observe that

$$\mathbf{P}(S_{N(t)} - \mu\lambda(t) > x) = \sum_{k=1}^{\infty} \mathbf{P}(S_k - \mu\lambda(t) > x) \mathbf{P}(N(t) = k),$$

which we can split into three parts

$$\sum_{k < (1-\delta)\lambda(t)} + \sum_{(1-\delta)\lambda(t) \leq k \leq (1+\delta)\lambda(t)} + \sum_{k > (1+\delta)\lambda(t)} := I_1 + I_2 + I_3, \tag{3.1}$$

where δ is arbitrarily chosen from the interval $(0, \gamma/\mu)$.

We have

$$\begin{aligned} I_1 &= \sum_{k < (1-\delta)\lambda(t)} \mathbf{P}(S_k - \mu\lambda(t) > x) \mathbf{P}(N(t) = k) \\ &\leq \sum_{k < (1-\delta)\lambda(t)} \mathbf{P}[S_{(1-\delta)\lambda(t)} > x + \mu\lambda(t)] \mathbf{P}(N(t) = k) \\ &= \sum_{k < (1-\delta)\lambda(t)} \mathbf{P}[S_{(1-\delta)\lambda(t)} - (1-\delta)\lambda(t)\mu > x + \mu\lambda(t) - (1-\delta)\lambda(t)\mu] \mathbf{P}(N(t) = k). \end{aligned}$$

Combining this expression with the result in Proposition 3, we get

$$\begin{aligned} I_1 &\lesssim \sum_{k < (1-\delta)\lambda(t)} (1-\delta)\lambda(t) \bar{F}(x + \mu\lambda(t) - (1-\delta)\lambda(t)\mu) \mathbf{P}(N(t) = k) \\ &= \sum_{k < (1-\delta)\lambda(t)} (1-\delta)\lambda(t) \bar{F}(x + \delta\mu\lambda(t)) \mathbf{P}(N(t) = k) \\ &\leq (1-\delta)\lambda(t) \bar{F}(x) \sum_{k < (1-\delta)\lambda(t)} \mathbf{P}(N(t) = k) \\ &= (1-\delta)\lambda(t) \bar{F}(x) \mathbf{P}(N(t) < (1-\delta)\lambda(t)) \\ &= (1-\delta)\lambda(t) \bar{F}(x) \mathbf{P}(N(t) - \lambda(t) < -\delta\lambda(t)). \end{aligned}$$

Now we use Assumption N_1 to get

$$I_1 = o(\lambda(t) \bar{F}(x)) \tag{3.2}$$

for every fixed $0 < \delta < \gamma/\mu$.

Next, we deal with

$$\begin{aligned} I_2 &= \sum_{(1-\delta)\lambda(t) \leq k \leq (1+\delta)\lambda(t)} \mathbf{P}(S_k - \mu\lambda(t) > x) \mathbf{P}(N(t) = k) \\ &\leq \mathbf{P}(S_{(1+\delta)\lambda(t)} - \mu\lambda(t) > x) \mathbf{P}\left[(1-\delta) \leq \frac{N(t)}{\lambda(t)} \leq (1+\delta)\right]. \end{aligned}$$

Under Assumption N_1 , we have

$$I_2 \lesssim \mathbf{P}(S_{(1+\delta)\lambda(t)} - (1+\delta)\lambda(t)\mu > x + \mu\lambda(t) - (1+\delta)\lambda(t)\mu)$$

as $t \rightarrow \infty$. Now we use the result of Proposition 3 to get

$$I_2 \lesssim (1+\delta)\lambda(t) \bar{F}[x + \mu\lambda(t) - (1+\delta)\lambda(t)\mu].$$

Since

$$x + \mu\lambda(t) - (1+\delta)\lambda(t)\mu \geq \left(1 - \frac{\delta\mu}{\gamma}\right)x \quad \text{or, equivalently,} \quad x \geq \gamma\lambda(t),$$

we get

$$I_2 \lesssim (1+\delta)\lambda(t) \bar{F}\left[\left(1 - \frac{\delta\mu}{\gamma}\right)x\right]$$

as $t \rightarrow \infty$. Also,

$$I_2 = \sum_{(1-\delta)\lambda(t) \leq k \leq (1+\delta)\lambda(t)} \mathbf{P}(S_k - \mu\lambda(t) > x) \mathbf{P}(N(t) = k) \\ \geq \mathbf{P}(S_{(1-\delta)\lambda(t)} - \mu\lambda(t) > x) \mathbf{P}\left((1-\delta) \leq \frac{N(t)}{\lambda(t)} \leq (1+\delta)\right).$$

By Assumption N_1 we obtain

$$I_2 \gtrsim \mathbf{P}(S_{(1-\delta)\lambda(t)} - (1-\delta)\lambda(t)\mu > x + \mu\lambda(t) - (1-\delta)\lambda(t)\mu)$$

as $t \rightarrow \infty$. Further, from the result of Proposition 3 we get

$$I_2 \gtrsim (1-\delta)\lambda(t)\bar{F}(x + \mu\lambda(t) - (1-\delta)\lambda(t)\mu).$$

By the inequality

$$x + \mu\lambda(t) - (1-\delta)\lambda(t)\mu \leq \left(1 + \frac{\delta\mu}{\gamma}\right)x \quad \text{or, equivalently,} \quad x \geq \gamma\lambda(t),$$

we obtain

$$I_2 \gtrsim (1-\delta)\lambda(t)\bar{F}\left[\left(1 + \frac{\delta\mu}{\gamma}\right)x\right]$$

as $t \rightarrow \infty$. From $F \in \mathcal{C}$ we get

$$\lim_{\delta \searrow 0} \limsup_{x \rightarrow \infty} \frac{1}{\bar{F}(x)} \bar{F}\left[\left(1 + \frac{\delta\mu}{\gamma}\right)x\right] = \lim_{\delta \searrow 0} \limsup_{x \rightarrow \infty} \frac{1}{\bar{F}(x)} \bar{F}\left[\left(1 - \frac{\delta\mu}{\gamma}\right)x\right] = 1,$$

and so

$$\lim_{\delta \searrow 0} \limsup_{t \rightarrow \infty} \sup_{x \geq \gamma\lambda(t)} \left| \frac{I_2}{\lambda(t)\bar{F}(x)} - 1 \right| = 0. \tag{3.3}$$

Now we calculate the

$$I_3 = \sum_{k > (1+\delta)\lambda(t)} \mathbf{P}(S_k - \mu\lambda(t) > x) \mathbf{P}(N(t) = k) \leq \sum_{k > (1+\delta)\lambda(t)} \mathbf{P}(S_k > x) \mathbf{P}(N(t) = k).$$

From Lemma 2, choosing $\gamma_F \leq u < p - 1$, we get the inequality

$$\leq \sum_{k > (1+\delta)\lambda(t)} \left[k\bar{F}\left(\frac{x}{u}\right) + \left(\frac{e\mu k}{x}\right)^u \right] \mathbf{P}(N(t) = k) \\ \leq \sum_{k > (1+\delta)\lambda(t)} k\bar{F}\left(\frac{x}{u}\right) \mathbf{P}(N(t) = k) + \left(\frac{e\mu}{x}\right)^u \sum_{k > (1+\delta)\lambda(t)} k^u \mathbf{P}(N(t) = k) \\ := K_1 + K_2.$$

For the first term, we obtain

$$K_1 \asymp \sum_{k > (1+\delta)\lambda(t)} k\bar{F}(x) \mathbf{P}(N(t) = k)$$

$$\begin{aligned} &\leq \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \sum_{n \geq k/(\tau+\theta)} \mathbf{P}\left(\sum_{i=1}^n Z_i = k\right) \mathbf{P}(\tau(t) = n) \\ &\quad + \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \sum_{1 \leq n \leq k/(\tau+\theta)} \mathbf{P}\left(\sum_{i=1}^n Z_i = k\right) \mathbf{P}(\tau(t) = n) \\ &= J_1 + J_2, \end{aligned}$$

where $\theta = \mathbf{E}Z_1 \geq 1$, and $\tau > 0$ is such that

$$\frac{(1 + \delta)\theta}{\tau + \theta} > 1.$$

Further, we deal with

$$\begin{aligned} J_1 &= \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \sum_{n \geq k/(\tau+\theta)} \mathbf{P}\left(\sum_{i=1}^n Z_i = k\right) \mathbf{P}(\tau(t) = n) \\ &\leq \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \sum_{n \geq k/(\tau+\theta)} \mathbf{P}(\tau(t) = n) = \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \mathbf{P}\left(\tau(t) \geq \frac{k}{\tau + \theta}\right). \end{aligned}$$

From Lemma 3 we get the expression

$$J_1 = o(1)\bar{F}(x) = o(\bar{F}(x)).$$

Further, we write

$$\begin{aligned} J_2 &= \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \sum_{1 \leq n < k/(\tau+\theta)} \mathbf{P}\left(\sum_{i=1}^n Z_i = k\right) \mathbf{P}(\tau(t) = n) \\ &\leq \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \sum_{1 \leq n < k/(\tau+\theta)} \mathbf{P}\left(\sum_{i=1}^n Z_i = k\right) \\ &\leq \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \mathbf{P}\left(\sum_{1 \leq i < k/(\tau+\theta)} Z_i \geq k\right) \\ &\leq \bar{F}(x) \sum_{k>(1+\delta)\lambda(t)} k \mathbf{P}\left(\sum_{1 \leq i < k/(\tau+\theta)} (Z_i - \theta) \geq \frac{\tau k}{\tau + \theta}\right). \end{aligned}$$

By [6, Corol. 3], which says that, for i.i.d. r.v.'s $\{U_n, n \geq 1\}$,

$$\mathbf{P}\left(\sum_{k=1}^n U_k \geq x\right) \leq \left(1 + \frac{2}{p}\right)^p n \mathbf{E}|U_1|^p x^{-p} + \exp\{-2(p+2)^{-2} e^{-p} x^2 (n \mathbf{E}U_1^2)^{-1}\}$$

if $\mathbf{E}U_1 = 0$ and $\mathbf{E}|U_1|^p < \infty$ for some $p \geq 3$, we get

$$\mathbf{P}\left(\sum_{1 \leq i \leq k/(\tau+\theta)} (Z_i - \theta) \geq \frac{\tau k}{\tau + \theta}\right)$$

$$\begin{aligned} &\leq \left(1 + \frac{2}{p}\right)^p \frac{k}{\tau + \theta} \mathbf{E}|Z_i - \theta|^p \left(\frac{\tau k}{\tau + \theta}\right)^{-p} \\ &\quad + \exp\left\{-2(-p + 2)^{-2} e^{-p} \left(\frac{k}{\tau + \theta} \mathbf{E}(Z_i - \theta)^2\right)^{-1} \left(\frac{\tau k}{\tau + \theta}\right)^2\right\} \\ &\leq C_p k^{-p+1} + \exp\{-D_p k\}, \end{aligned}$$

where C_p and D_p are positive constants. Therefore, we get the asymptotic expressions

$$J_2 = \bar{F}(x) o(1) = o(\bar{F}(x))$$

and

$$K_1 = J_1 + J_2 = o(\bar{F}(x)).$$

Now we examine

$$K_2 = \left(\frac{e\mu}{x}\right)^u \sum_{k > (1+\delta)\lambda(t)} k^u \mathbf{P}(N(t) = k) = (e\mu)^u x^{-u} \sum_{k > (1+\delta)\lambda(t)} k^u \mathbf{P}(N(t) = k).$$

In combination with Lemma 1, we get

$$K_2 = (e\mu)^u o(\bar{F}(x)) \sum_{k > (1+\delta)\lambda(t)} k^u \mathbf{P}(N(t) = k).$$

In the same way, we obtain

$$\sum_{k > (1+\delta)\lambda(t)} k^u \mathbf{P}(N(t) = k) = o(1),$$

and so we get the relation

$$K_2 = o(\bar{F}(x));$$

hence,

$$I_3 = K_1 + K_2 = o(\bar{F}(x)). \tag{3.4}$$

Finally, substituting (3.2)–(3.4) into (3.1), we obtain the result.

Acknowledgments. We have the pleasant duty to thank the referee for her/his substantial comments on the initial version of the paper.

REFERENCES

1. N.H. Bingham, Goldie, and J.L. Teugels, *Regular Variation*, Cambridge Univ. Press, 1987.
2. D.B.H. Cline, Intermediate regular and π variation, *Proc. London Math. Soc.*, **68**:594–616, 1994.
3. D.B.H. Cline and T. Hsing, Large deviation probabilities for sums and maxima or rv’s with heavy or subexponential tails, preprint, Texas A&M University, 1991.
4. D.B.H. Cline and G. Samorodnitsky, Subexponentiality of the product of independent random variables, *Stoch. Process. Appl.*, **49**:75–98, 1994.

5. P. Embrechts, C. Klüppelberg, and T. Mikosch, *Modelling Extremal Events for Insurance and Finance*, Springer, Berlin, 1997.
6. D. Fuk and S.V. Nagaev, Probability for sums of independent random variables, *Theory Probab. Appl.*, **16**:600–675, 1971.
7. C.C. Heyde, On large deviation probabilities in the case of attraction to a nonnormal stable law, *Sankhyā, Ser. A*, **30**:253–258, 1968.
8. C.C. Heyde, A contribution to the theory of the large deviations for sums of independent random variables, *Z. Wahrscheinlichkeitstheor. Verw. Geb.*, **7**:303–308, 1967.
9. C.C. Heyde, On large deviation problems for sums of random variables which are not attracted to the normal law, *Ann. Math. Stat.*, **38**:1575–1578, 1967.
10. C. Klüppelberg and T. Mikosch, Large deviation probabilities of heavy-tailed random sums with applications in insurance and finance, *J. Appl. Probab.*, **34**:293–308, 1997.
11. T. Mikosch and S.V. Nagaev, Large deviation of heavy-tailed sums with applications in insurance, *Extremes*, **1**:81–110, 1998.
12. A.V. Nagaev, Integral limit theorems for large deviations when Cramer's condition is not fulfilled I, *Theory Probab. Appl.*, **14**:51–64, 1969.
13. A.V. Nagaev, Integral limit theorems for large deviations when Cramer's condition is not fulfilled II, *Theory Probab. Appl.*, **14**:193–208, 1969.
14. S.V. Nagaev, Large deviation for sum of independent random variables, in *Transactions of the Sixth Prague Conference on Information Theory, Statistical Decision Functions, and Random Processes*, Academic Prague, 1973, pp. 657–674.
15. S.V. Nagaev, Large deviation of sums in independent random variables, *Ann. Probab.*, **7**:745–789, 1979.
16. K.W. Ng, Q. Tang, J.A. Yan, and H. Yang, Precise large deviation for sums of random variables with consistently varying tails, *J. Appl. Probab.*, **41**:93–107, 2004.
17. V. Paulauskas and A. Skučaitė, Some asymptotic results for one-sided large deviation probabilities, *Lith. Math. J.*, **43**(3):318–326, 2003.
18. I.F. Pinelis, On the asymptotic equivalence of probabilities of large deviation for sums and maxima of independent random variables, in *Tr. Inst. Mat.*, Vol. 5, Nauka, Novosibirsk, 1985.
19. A. Skučaitė, Large deviations for sums of independent heavy-tailed random variables, *Lith. Math. J.*, **44**(2):198–208, 2004.
20. Q. Tang, C. Su, T. Jiang, and J.S. Zang, Large deviation for heavy-tailed random sums in compound renewal model, *Stat. Probab. Lett.*, **52**:91–100, 2001.