

# Some liminf results for the Increments of stable subordinators

Abdelkader Bahram<sup>1</sup> and Bader Almohaimed<sup>2</sup>

## Abstract

Let  $\{X(t), 0 \leq t < \infty\}$  be a stable subordinator defined on a probability space  $(\Omega, \mathcal{F}, \mathcal{A})$ . In this paper we establish some liminf for increments of stable subordinators and we obtain similar results for delayed sums.

**Mathematics Subject Classification:** 60F15; 60G17

**Keywords:** Increments; Stable Subordinators; Law of Iterated Logarithm

## 1 Introduction

Let  $\{X(t), 0 \leq t < \infty\}$  be a stable subordinator with exponent  $\alpha$ ,  $0 < \alpha < 1$ , defined on a probability space  $(\Omega, \mathcal{F}, \mathcal{A})$ . Let  $a_t$ ,  $t > 0$ , be a non-negative

---

<sup>1</sup> Department of Mathematics, Djillali Liabes University, Sidi Bel-Abbes, Algeria.

Department of Mathematics, Faculty of Science, Qassim University, Saudi Arabia.

E-mail: menaouar.1926@yahoo.fr

<sup>2</sup> Department of Mathematics, Faculty of Science, Qassim University, Saudi Arabia.

E-mail: bsmhiemied@qu.edu.sa

valued function of  $t$  such that (i)  $0 < a_t \leq 1$ , (ii)  $a_t \rightarrow \infty$  as  $t \rightarrow \infty$  (iii)  $a_t/t \rightarrow 0$  as  $t \rightarrow \infty$ . let  $Y(t) = X(t + a_t) - X(t)$ ,  $t > 0$  and  $Y(0) = 0$ . Define  $\lambda_\beta(t) = \theta_\alpha a_t^{\frac{1}{\alpha}} (\log \frac{t}{a_t} (\log t)^\beta (\log a_t)^{1-\beta})^{\frac{\alpha-1}{\alpha}}$ , where  $\theta_\alpha = (B(\alpha))^{\frac{1-\alpha}{\alpha}}$ ,  $B(\alpha) = (1 - \alpha) \alpha^{\frac{\alpha-1}{\alpha}} (\cos(\frac{\pi\alpha}{2}))^{\frac{1}{\alpha-1}}$ ,  $0 < \alpha < 1$  and  $0 \leq \beta \leq 1$ . Observe that the process has the property that  $t^{-\frac{1}{\alpha}} X(t)$  and  $X(1)$  are identically distributed. A real valued increasing process  $\{X(t), t > 0\}$  with stationary independent is called a subordinator. For any given  $t$ , the characteristic function of  $X(t)$  is the form

$$E(e^{\{iuX(t)\}}) = \exp \left\{ -t|u|^\alpha \left( 1 - \frac{iu}{|u|} \tan \left( \frac{\pi\alpha}{2} \right) \right) \right\}, \quad 0 < \beta < 1.$$

Throughout the paper  $\varepsilon$ ,  $c$ ,  $\delta$  and  $K$  (integer), with or without suffix, stand for positive constants; i.o. means infinitely often; we shall define for each  $u \geq 0$  the functions  $\log u = \log(\max(u, 1))$ ,  $\log \log u = \log \log(\max(u, 3))$ ,  $g(t) = (t \log t)/a_t$  and  $g_\beta(t) = \frac{t}{a_t} (\log t)^\beta (\log a_t)^{1-\beta}$  with  $0 \leq \beta \leq 1$ , so that  $\lambda_{(t,\beta)} = (2a_t \log g_\beta(t))^{-\frac{1}{2}}$ .

Vasudeva and Divanji [6] have obtained the following limit inferior for the increments of stable subordinators. Under certain condition on  $a_t$ , it was shown that  $\liminf_{t \rightarrow \infty} \frac{Y(t)}{\lambda_1(t)} = 1 a.s$

Hwang et al.[2] and Bahram and Shehawy [1] studied this subsequence principle for increments of Gaussian processes in obtaining limsup. In this paper we study an almost sure limit inferior behaviour for increments of stable subordinators for proper selection of subsequences and extended to delayed sums.

## 2 Main results

In order to prove Theorem 2.1, we need to give the following Lemma.

**Lemma 2.1.** (see [5] or [6]) Let  $X_1$  be a positive stable random variable with characteristic function

$$E(\exp\{iuX_1\}) = \exp \left\{ -|u|^\alpha \left( 1 - \frac{iu}{|u|} \tan \left( \frac{\pi\alpha}{2} \right) \right) \right\}, \quad 0 < \alpha < 1. \quad \text{Then, as } x \rightarrow 0$$

$$P(X_1 \leq x) \simeq \frac{x^{\frac{\alpha}{2(1-\alpha)}}}{\sqrt{2\pi\alpha B(\alpha)}} \exp \left\{ -B(\alpha)x^{\frac{\alpha}{\alpha-1}} \right\}$$

where

$$B(\alpha) = (1 - \alpha)\alpha^{\frac{\alpha-1}{\alpha}} \left(\cos\left(\frac{\pi\alpha}{2}\right)\right)^{\frac{1}{\alpha-1}}.$$

**Theorem 2.1.** Let  $a_t, t > 0$  be a nondecreasing function of  $t$  such *i)*  $0 < a_t \leq t$ , *ii)*  $a_t \rightarrow \infty$ , as  $t \rightarrow \infty$  and *iii)*  $a_t/t \rightarrow 0$  as  $t \rightarrow \infty$ . Let  $(t_k)$  be an increasing sequence of positive integers such that

$$\limsup_{k \rightarrow \infty} \frac{t_{k+1} - t_k}{a_{t_k}} < 1. \tag{1}$$

Then

$$\liminf_{k \rightarrow \infty} \frac{Y(t_k)}{\lambda_\beta(t_k)} = \varepsilon^* \quad a.s.,$$

where

$$\varepsilon^* = \inf\{\varepsilon > 0 : \sum_k (g_\beta(t_k))^{-\varepsilon^{-\gamma}} < \infty, \quad 0 \leq \beta \leq 1\} \quad \text{and} \quad \gamma = \frac{\alpha}{\alpha - 1}, \quad 0 < \alpha < 1.$$

**Proof** Equivalently, we show that for any given  $\varepsilon_1 > 0$ , as  $k \rightarrow \infty$ ,

$$P(Y(t_k) \leq (\varepsilon^* + \varepsilon_1)\lambda_\beta(t_k) \quad i.o.) = 1 \tag{2}$$

and

$$P(Y(t_k) \leq (\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k) \quad i.o.) = 0. \tag{3}$$

The condition (1) implies that  $t_{k+1} < t_k + a_{t_k}$ , for large  $k$  and by Mjnheer [5], we have

$$P(Y(t_k) \leq (\varepsilon^* + \varepsilon_1)\lambda_\beta(t_k)) = P\left(X_1 \leq \frac{(\varepsilon^* + \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}\right) \tag{4}$$

Observe that

$$\frac{(\varepsilon^* + \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}} = (\varepsilon^* + \varepsilon_1)\theta_\alpha (\log g_\beta(t_k))^{\frac{\alpha-1}{\alpha}}$$

taken as  $x$ , in the above lemma, one can find a  $k_1$  and some constant  $C_1$ , such that for all  $k \geq k_1$ ,

$$P(X_1 \leq \frac{(\varepsilon^* + \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}) \geq c_1 (\log g_\beta(t_k))^{-\frac{1}{2}} \exp\{-(\varepsilon^* + \varepsilon_1)^{\frac{\alpha}{\alpha-1}} \log g_\beta(t_k)\}$$

where  $g_\beta(t) = \frac{t}{a_t} (\log t)^\beta (\log a_t)^{1-\beta}$  and  $0 \leq \beta \leq 1$ .

Notice that from the definition of  $\varepsilon_*$ , we have  $\varepsilon_* \geq 1$  implies that there exists

$\varepsilon_2 > 0$  such that  $(\varepsilon^* + \varepsilon_1)^{\frac{\alpha}{\alpha-1}} < (1 - \varepsilon_2) < 1$ . Hence  $P(X_1 \leq \frac{(\varepsilon^* + \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}) \geq \frac{c_1}{(\log g_\beta(t_k))^{\frac{1}{2}} (g_\beta(t_k))^{1-\varepsilon_2}}$ . Let  $l_k = \frac{t_k}{a_{t_k}}$  and  $m_k = (\log t_k)^\beta (\log a_{t_k})^{1-\beta}$ . Since  $\frac{a_{t_k}}{t_k} \rightarrow 0$ , as  $k \rightarrow \infty$ ,  $l_k$  is non decreasing and  $m_k \rightarrow \infty$ , as  $k \rightarrow \infty$ , one can find a constant  $k_2 \geq k_1$  such that  $\frac{l_k^{\varepsilon_2} m_k^{\varepsilon_2}}{(\log l_k m_k)^{\frac{1}{2}}} \geq 1$ , whenever  $k \geq k_2$ . By condition (1), for all  $k \geq k_2$ , we therefore have,

$$\begin{aligned}
P(X_1 \leq \frac{(\varepsilon^* + \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}) &\geq c_1 (g_\beta(t_k))^{-1} \\
&= c_1 \left( \frac{t_k (\log t_k)^\beta (\log a_{t_k})^{1-\beta}}{a_{t_k}} \right)^{-1} \\
&= c_1 \left( \frac{a_{t_k}}{t_k} \left( \frac{\log a_{t_k}}{\log t_k} \right)^\beta \frac{1}{\log a_{t_k}} \right) \\
&\geq c_1 \left( \frac{a_{t_k}}{t_k} \left( \frac{\log a_{t_k}}{\log t_k} \right) \frac{1}{\log a_{t_k}} \right) \\
&= c_1 (g(t_k))^{-1} \\
&= c_1 \frac{t_{k+1} - t_k}{t_k \log t_k}. \tag{5}
\end{aligned}$$

Observing that

$$\sum_{k=k_2}^{\infty} \frac{t_{k+1} - t_k}{t_k \log t_k} \geq \int_c^{\infty} \frac{dt}{t \log t}$$

for some  $c > 0$  and that  $\int_c^{\infty} \frac{dt}{t \log t} = \infty$ . Hence from (4) and (5), we get

$$\sum_{k=k_2}^{\infty} P(Y(t_k) \leq (\varepsilon^* + \varepsilon_1)\lambda_{(t_k, \beta)}) = \infty.$$

The Condition (1) implies that  $t_{k+1} \leq t_k + a_{t_k}$ , for large  $k$  one can observe that  $Y(t_k)$ 's are mutually independent and hence by Borel-Cantelli Lemma, we have,

$$P(Y(t_k) \leq (\varepsilon^* + \varepsilon_1)\lambda_{(t_k, \beta)} \quad i.o) = 1,$$

which establishes (2).

Now we complete the proof by showing that, for any  $\varepsilon_1 \in (0, 1)$ ,

$$P(Y(t_k) \leq (\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k) \quad i.o) = 1.$$

From condition (1), we have  $t_{k+1} \leq t_k + a_{t_k}$ , for large  $k$  and from Mijneer [5], one can find a  $k_2$  such that for all  $k \geq k_2$ ,

$$P(Y(t_k) \leq (\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k) \quad i.o) = P(X(t_k + a_{t_k}) - X(t_k)) \leq (\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k) \quad i.o)$$

Hence in order to prove (3), it is enough to show that

$$P(X(t_k + a_{t_k}) - X(t_k) \leq (\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k) \quad i.o) = 0. \tag{6}$$

We know that  $t^{1/\alpha}X(t) = X(1)$ , which implies

$$P(X(t_k + a_{t_k}) - X(t_k) \leq (\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k)) = P\left(X(1) \leq \frac{(\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}\right)$$

and

$$\frac{(\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}} = (\varepsilon^* - \varepsilon_1)\theta_\alpha (\log(g_\beta(t_k)))^{(\alpha-1)/\alpha}.$$

By taking  $x = (\varepsilon^* - \varepsilon_1)\theta_\alpha (\log(g_\beta(t_k)))^{(\alpha-1)/\alpha}$ , where  $g_\beta(t) = \frac{t}{a_t}(\log t)^\beta(\log a_t)^{1-\beta}$ , in the above lemma, one can find a  $k_4$  and  $c_1$  such that for all  $k \geq k_4$ ,

$$P\left(X(1) \leq \frac{(\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}\right) \leq \frac{c_2}{(\log(g_\beta(t_k)))^{1/2}} \exp\{-(\varepsilon^* - \varepsilon_1)^{\frac{\alpha}{\alpha-1}} \log g_\beta(t_k)\}.$$

Observe that using properties of  $\{a_t\}$ , one can find some constant  $C_3$  and  $k_4$  such that for all  $k \geq k_4$ ,

$$\begin{aligned} P(X(1) \leq (\varepsilon^* - \varepsilon_1)\theta_\alpha (\log(g_\beta(t_k)))^{(\alpha-1)/\alpha}) \\ \leq \frac{c_3}{(g_\beta(t_k))^{(\varepsilon^* - \varepsilon_1)^{\frac{\alpha}{\alpha-1}}}}. \end{aligned}$$

Notice that  $\varepsilon^* = \inf\{\varepsilon > 0 : \sum_k (g_\beta(t_k))^{-\varepsilon^{-\gamma}} < \infty, \quad 0 \leq \beta \leq 1\}$  and  $\gamma = \frac{\alpha}{\alpha-1} < 0, 0 < \alpha < 1$  which yields  $\varepsilon^* \geq 1$ .

Since  $\varepsilon_1 \in (0, 1)$ , choose  $\varepsilon_1$  sufficiently small one can find  $k_5$  such that for all  $k \geq k_5$ ,

$$\sum_{k=k_5}^{\infty} P\left(X(1) \leq \frac{(\varepsilon^* - \varepsilon_1)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}\right) \leq \sum_{k=k_5}^{\infty} \frac{C_3}{(g_\beta(t_k))^{(\varepsilon^* - \varepsilon_1)^\gamma}} < \infty,$$

where  $\gamma = \frac{\alpha}{\alpha-1}, \quad 0 < \beta < 1$ .

By Borel-Cantelli Lemma, (3) holds which implies (6) holds and proof of the theorem is completed. □

**Theorem 2.2.** *Let  $a_t, t > 0$  be a nondecreasing function of  $t$  such i)  $0 < a_t < t$ , ii)  $a_t \rightarrow \infty$ , as  $t \rightarrow \infty$  and iii)  $a_t/t \rightarrow 0$  as  $t \rightarrow \infty$ . Let  $(t_k)$  be an increasing sequence of positive integers such that*

$$\liminf_{k \rightarrow \infty} \frac{t_{k+1} - t_k}{a_{t_k}} > 1. \tag{7}$$

Then

$$\liminf_{k \rightarrow \infty} \frac{Y(t_k)}{\lambda_\beta(t_k)} = 1 \quad a.s.,$$

where  $0 \leq \beta \leq 1$ .

**Proof** To prove the Theorem, it is enough to show that for any  $\varepsilon \in (0, 1)$ ,

$$P(Y(t_k) \leq (1 + \varepsilon)\lambda_\beta(t_k) \quad i.o.) = 1 \quad (8)$$

and

$$P(Y(t_k) \leq (1 - \varepsilon)\lambda_\beta(t_k) \quad i.o.) = 0 \quad (9)$$

By the Theorem of Vasudeva and Divanji [6], we claim that

$$\liminf_{k \rightarrow \infty} \frac{Y(t_k)}{\lambda_\beta(t_k)} \geq \liminf_{k \rightarrow \infty} \frac{Y(t_k)}{\lambda_1(t_k)} \geq \liminf_{t \rightarrow \infty} \frac{Y(t)}{\lambda_1(t)} = 1 \quad a.s.,$$

which establishes (9).

The condition (7) implies that there exists a  $k_1$  such that  $t_{k+1} > t_k + a_{t_k}$ , for all  $k \geq k_1$ . This in turn implies that  $\{Y(t_k), k \geq 1\}$  is a sequence of mutually independent r.v.s. We can observe that with a minor modification, the proof of (8) follows on similar lines of (2). That is using Lemma 2.1, one can find  $C_1$  and  $k_2$  such that for all  $k \geq k_2$ .

$$P(X_1 \leq \frac{(1 + \varepsilon)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}) \geq c_1(g(t_k))^{-(1+\varepsilon)\frac{\alpha}{\alpha-1}}.$$

Choose  $\varepsilon' > 0$  such that  $(1 + \varepsilon)^{a_{t_k}^{1/\alpha}} < (1 - \varepsilon') < 1$  and hence we have,

$$P(X_1 \leq \frac{(1 + \varepsilon)\lambda_\beta(t_k)}{a_{t_k}^{1/\alpha}}) \geq c_1(g(t_k))^{-(1-\varepsilon')}.$$

Following similar arguments of proof of (4) and (5), we get

$$P(Y(t_k) \leq (1 + \varepsilon)\lambda_\beta(t_k) = \infty,$$

which in turn implies the proof of (8). Hence the proof of the Theorem is completed.  $\square$

### 3 Similar result for delayed sums

Let  $\{X_n, n \geq 1\}$  be a sequence of i.i.d strictly positive stable r.v.s with index  $\alpha, 0 < \alpha < 1$ . Let  $\{a_n, n \geq 0\}$  be a sequence of non-decreasing functions of positive integers of  $n$  such that  $0 < a_n < n$ , for all  $n$  and we assume that  $a_n/n \downarrow 0$  as  $n \rightarrow \infty$ . Define  $\lambda_\beta(n) = \theta_\alpha a_n^{\frac{1}{\alpha}} (\log \frac{n}{a_n} + \beta \log \log n + (1 - \beta) \log \log a_n)^{\frac{\alpha-1}{\alpha}}$ , where  $\theta_\alpha = (B(\alpha))^{\frac{1-\alpha}{\alpha}}$ ,  $B(\beta) = (1 - \alpha)\alpha^{\frac{1}{1-\alpha}} (\cos(\frac{\pi\alpha}{2}))^{\frac{1}{\alpha-1}}$ ,  $0 \leq \beta \leq 1$  and  $0 < \alpha < 1$ . Observe that the process has the property that  $n^{-1/\alpha}X(n)$  and  $X(1)$  are identically distributed. Let  $S_n = \sum_{k=1}^n X_k$  and set  $M_n = S_{n+a_n} - S_n$ , where  $\{M_n, n \geq 1\}$  is called a (forward) delayed sum (See Lai [3]). Define the  $X_n = X(n) - X(n - 1), n = 1, 2, \dots ; X(0) = 0$ , then  $S_n = \sum_{k=1}^n X_k$  with  $S_0 = 0$ , which yields  $M_n = S_{n+a_n} - S_n = X(n + a_n) + X(n) = Y(n)$ .

**Theorem 3.1.** *Let  $\{a_n, n > 0\}$  be a sequence of non-decreasing functions of positive integers of  $n$  such that  $0 < a_n < n$  non-decreasing and  $a_n/n$  non-increasing. Let  $(n_k, k \geq 1)$  be any increasing sequence of positive integers such that*

$$\limsup \frac{n_{k+1} - n_k}{a_{n_k}} < 1. \tag{10}$$

Then

$$\liminf_{k \rightarrow \infty} \frac{M_{n_k}}{\lambda_\beta(n_k)} = \varepsilon^* \quad a.s.,$$

where

$$\varepsilon^* = \inf \{ \varepsilon > 0 : \sum_k (g_\beta(n_k))^{-\varepsilon^\gamma} < \infty, \quad 0 \leq \alpha \leq 1 \},$$

and

$$\gamma = \frac{\alpha}{\alpha - 1}, \quad 0 < \alpha < 1.$$

**Proof** To prove the theorem it is sufficient to show that for any given  $\varepsilon_1 \in (0, 1)$

$$P(M_{n_k} \leq (\varepsilon_* + \varepsilon)\lambda_\beta(n_k) \quad i.o.) = 1, \tag{11}$$

and

$$P(M_{n_k} \leq (\varepsilon_* - \varepsilon_2)\lambda_\beta(n_k) \quad i.o.) = 0. \tag{12}$$

The proof of (11) is an immediate consequence of (2) and the proof of (12) follows on the similar lines of Vasudeva and Divanji [6]. Hence the details are omitted. □

**Theorem 3.2.** Let  $\{a_n, n > 0\}$  be a sequence of non-decreasing functions of positive integers of  $n$  such that 1)  $0 < a_n \leq n$ ,  $n > 0$ , 2)  $a_n \rightarrow \infty$ , as  $n \rightarrow \infty$ , and  $a_n/n \rightarrow 0$ , as  $n \rightarrow \infty$ . Let  $(n_k, k \geq 1)$  be any increasing sequence of positive integers such that  $\liminf_{k \rightarrow \infty} \frac{n_{k+1} - n_k}{a_{n_k}} > 1$ . Then

$$\liminf_{k \rightarrow \infty} \frac{M_{n_k}}{\lambda_\beta(n_k)} = 1 \quad a.s.$$

**Proof** The proof of Theorem 3.2 is a direct consequence of above Theorem 2.2 and hence the details are omitted.  $\square$

## References

- [1] A. Bahram and S. Shehawy, Study of the Convergence of the Increments of Gaussian Process, *Applied Mathematics*, **6**, (2015), 933-939.
- [2] K.S. Hwang, Y.K. Choi and J.S. Jung, On superior limits for the increments of Gaussian Processes, *Statistics and Probability Letter*, **35**, (1997), 289-296.
- [3] T.L. Lai, Limit Theorems for delayed Sums, *Annals of Probability*, **2**, (1973), 432-440.
- [4] J.L. Mijhneer, *Sample Path Properties of Stable Process*, Mathematisch Centrum, Amsterdam, 1975.
- [5] J.L. Mijhneer, On the law of iterated logarithm for subsequences for a stable subordinator, *Journal of Mathematical Sciences*, **76**, (1995), 2283-2286.
- [6] R.Vasudeva and G.Divanji, Law of Iterated Logarithm for Increments of Stable Subordinators, *Stochastic Processes and Their Applications*, **28**, (1988), 293-300.