

# A generalisation of almost sure limit sets of vector sequences of moving maxima

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## Abstract

We obtain almost sure limit sets of vector sequences of moving maxima and moving second maxima of a sequence of independent negative binomial random variable when  $p = \infty$ .

**Keywords:** Almost sure, moving maxima, moving second maxima.

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## INTRODUCTION

Let  $\{X_n, n \geq 1\}$  be sequence of independent negative binomial random variables with probability mass function (p.m.f.)

$$P(X=k) = \binom{k-1}{r-1} a^r (1-a)^{k-r},$$

$$k = r, r+1, \dots, \quad 0 < a < 1$$

Define moving maxima and moving second maxima by

$$Y_{k(n)}^* = \max \{X_{n-k(n)+1}, X_{n-k(n)+2}, \dots, X_n\}$$

$$S_{k(n)}^* = \text{second largest} \{X_{n-k(n)+1}, X_{n-k(n)+2}, \dots, X_n\}$$

where  $k(n)$  is a non decreasing sequence of positive integers. If  $k(n) = n$  then  $M_n^{(1)} = Y_{k(n)}^*$  and  $M_n^{(2)} = S_{(k)n}^*$

Thus  $Y_{k(n)}^*$  and  $S_{(k)n}^*$  may be regarded as general forms of  $M_n^{(1)}$  and  $M_n^{(2)}$  respectively.

Next we make the following assumptions on  $k(n)$ .

$k(n)$  is non-decreasing

$$\sup [k(n+1) - k(n)] \leq \mu \text{ (finite)}$$

and

$$k(n) = \left[ \frac{n}{(\log n)^{t(n)}} \right], \quad t(n) \rightarrow p, 0 \leq p < \infty, \text{ as } n \rightarrow \infty$$

in particular, if

$$k(n) = \begin{cases} [\alpha n], 0 < \alpha \leq 1 & \text{then } p = 0 \\ \left[ \frac{n}{(\log n)^q} \right], 0 < q < \infty, & \text{then } p = q \\ [n^\alpha], 0 < \alpha < 1, & \text{then } p = \infty \end{cases}$$

now define,

$$a(n) = \frac{\log \log n}{-\log(1-a)}, \quad b(n) = \frac{\log n}{-\log(1-a)}$$

Then we establish the following.

When  $0 \leq p < \infty$ , Nagesh S.(2014) establishes the result,

**Theorem 1.1:**

The almost sure limit set of the vector sequence  $\{(Y_{k(n)}^* - b(n))/a(n), (S_{k(n)}^* - b(n))/a(n)\}$ ,

$n \geq 1$  coincides with the region,

$$S = \{(x,y) : r-p-1 \leq x \leq r, r-p-1 \leq y \leq r, x \geq y, x+y \leq 2r-p-1\}.$$

A generalization of this result when  $p = \infty$ , is presented in the next section.

**ALMOST SURE LIMIT SET RESULT AND PROOF**

**Theorem 2.1:**

The almost sure limit set of the vector sequence

$$\left\{ \frac{Y_{k(n)}^*}{b(n)}, \frac{S_{k(n)}^*}{b(n)} \right\} n \geq 1 \text{ coincides with the region,}$$

$$S = \{(x, y) : \Delta \leq x, \Delta \leq y \text{ and } x + y \leq 1 + \Delta\},$$

when  $0 \leq \Delta < 1$ .

For proving the theorem we make use of the following Lemmas.

**Lemma 2.1.1**

For every  $\varepsilon > 0, x > y > \Delta$  and  $x+y \leq 1 + \Delta$

$$P \left\{ Y_{k(l_i)}^* > c_{l_i}(x + \varepsilon), S_{k(l_i)}^* > c_{l_i}(y) \text{ i.o.} \right\} = 0$$

and

$$P \left\{ Y_{k(l_i)}^* > c_{l_i}(x), S_{k(l_i)}^* > c_{l_i}(y + \varepsilon) \text{ i.o.} \right\} = 0$$

where  $l_i = [i^\theta]$  with  $\theta = 1/(x+y-2\Delta+\varepsilon/2)$

and  $c_{l_i}(x) = x b(l_i)$

**Proof:**

We have

$$P \left\{ Y_{k(l_i)}^* > c_{l_i}(x + \varepsilon), S_{k(l_i)}^* > c_{l_i}(y) \right\} \\ = P \left\{ Y_{k(l_i)}^* > c_{l_i}(x + \varepsilon) \right\} - P \left\{ Y_{k(l_i)}^* > c_{l_i}(x + \varepsilon), S_{k(l_i)}^* \leq c_{l_i}(y) \right\}$$

$$= \{ 1 - F^{k(l_i)}(c_{l_i}(x+\varepsilon)) \} - k(l_i) \{ F^{k(l_i)-1}(c_{l_i}(y)) \} \{ 1 - F(c_{l_i}(x+\varepsilon)) \} \\ = 1 - \exp \{ -k(l_i) [ 1 - F(c_{l_i}(x+\varepsilon)) ] (1 + o(1)) \} - k(l_i) [ \exp \{ -(k(l_i)-1) [ 1 - F(c_{l_i}(y)) ] (1 + o(1)) \} ]$$

$$[ 1 - F(c_{l_i}(y)) ] (1 + o(1)) ] [ 1 - F(c_{l_i}(x+\varepsilon)) ] \\ = 1 + o(1) \{ k(l_i) (k(l_i)-1) \} [ 1 - F(c_{l_i}(x+\varepsilon)) ] [ 1 - F(c_{l_i}(y)) ],$$

Since  $k(l_i) [ 1 - F(c_{l_i}(x+\varepsilon)) ] \rightarrow 0$  for  $x > \Delta$ ,

$$k(l_i) [ 1 - F(c_{l_i}(y)) ] \rightarrow 0 \text{ for } y > \Delta,$$

and

$$\frac{[ 1 - F(c_{l_i}(x+\varepsilon)) ]}{[ 1 - F(c_{l_i}(y)) ]} \rightarrow 0 \text{ for } x > y \text{ as } i \rightarrow \infty$$

$$\leq \text{Const. } k^2(l_i) [ 1 - F(c_{l_i}(x+\varepsilon)) ] [ 1 - F(c_{l_i}(y)) ]$$

for all  $i$  large

$$\leq \text{Const} \left( \frac{1}{i^{(\theta(x+\varepsilon-\Delta+\eta))}} \right) \left( \frac{1}{i^{(\theta(y-\Delta+\eta))}} \right), \eta > 0$$

$$= \text{Const} \left( \frac{1}{i^{\theta(x+\varepsilon-\Delta+\eta+y-\Delta+\eta)}} \right)$$

$$= \text{Const} \left( \frac{1}{i^{\theta(x+y+\varepsilon-2\Delta-2\eta)}} \right)$$

$$= \text{Const} \left( \frac{1}{i^{1+\eta_7}} \right), \eta_7 > 0.$$

Thus the proof of the Lemma is complete. .

### Lemma 2.1.2

For every  $\varepsilon > 0$  and  $x_0 = \Delta - \varepsilon$ ,

$$P \{ S_{k(n)}^* \leq c_n(x_0) \text{ i. o.} \} = 0.$$

**Proof:**

In view of Lemma Barndorff-Neilson(1961) and Devroye (1981)

it is sufficient to prove,

$$P \{ S_{k(n)}^* \leq c_n(x_0) \} \rightarrow 0$$

$$\text{and } \sum_{n=1}^{\infty} P \{ S_{k(n)}^* \leq c_n(x_0) \text{ and } S_{k(n+1)}^* > c_n(x_0) \} < \infty$$

Note that

$$P \{ S_{k(n)}^* \leq c_n(x_0) \} = P \{ S_{k(n)}^* \leq c_n(x_0), Y_{k(n)}^* \leq c_n(x_0) \} \\ + P \{ S_{k(n)}^* \leq c_n(x_0), Y_{k(n)}^* > c_n(x_0) \}$$

$$= P \{ Y_{k(n)}^* \leq c_n(x_0) \} + P \{ S_{k(n)}^* \leq c_n(x_0), Y_{k(n)}^* > c_n(x_0) \}$$

$$= F^{k(n)}(c_n(x_0)) + k(n) [ F^{k(n)-1}(c_n(x_0)) ] [ 1 - F(c_n(x_0)) ]$$

$$= E_n [ F^{k(n)-1}(c_n(x_0)) ] \left[ 1 + \frac{F(c_n(x_0))}{E_n} \right],$$

where  $E_n = k(n) \{ 1 - F(c_n(x_0)) \}$

$$\leq \left( \frac{1}{n^{(x_0 - \Delta - \eta)}} \right) \left( \exp \{ -k(n)[1 - F(c_n(x_0))] (1 + o(1)) \} \right)$$

$$\leq \left( \frac{1}{n^{(x_0 - \Delta - \eta)}} \right) \left( \exp \left\{ \frac{1}{n^{(x_0 - \Delta - \eta)}} \right\} \right)$$

$$= \left( n^{3\epsilon/2} (\exp(-n^{\epsilon/2})) \right)$$

$\rightarrow 0$ , as  $n \rightarrow \infty$ .

Now since  $\{c_n(x_0)\}$  is non decreasing, notice that

$$P \{ S_{k(n)}^* \leq c_n(x_0) \text{ and } S_{k(n+1)}^* > c_n(x_0) \}$$

$$\leq P \{ S_{k(n)}^* \leq c_n(x_0), Y_{k(n)}^* \leq c_n(x_0), S_{k(n+1)}^* > c_n(x_0) \}$$

$$+ P \{ S_{k(n)}^* \leq c_n(x_0), Y_{k(n)}^* > c_n(x_0), S_{k(n+1)}^* > c_n(x_0) \} \tag{2.1.1}$$

**Case(i):**

$$n - k(n+1) + 2 \geq n - k(n) + 1$$

RHS of (2.1.1)

$$\leq k(n) [F^{k(n)-1}(c_n(x_0))] [1 - F(c_n(x_0))]^2$$

$$= E_n [F^{k(n)-1}(c_n(x_0))] [1 - F(c_n(x_0))]$$

$$\leq n^{3\epsilon/2} \left( \exp \{ -n^{\epsilon/2} \} \right) \left( \frac{1}{n^{x_0} (\log n)^{-(r-1)}} \right)$$

$$= \left( \frac{1}{n^{(\Delta - \epsilon - 3\epsilon/2)}} \right) \left( \frac{1}{(\log n)^{-(r-1)}} \right) \left( \exp \{ -n^{\epsilon/2} \} \right)$$

$$\leq \frac{\text{Const.}}{n^{1 + \eta_8}} \eta_8 > 0.$$

**Case(ii):**

$$n - k(n+1) - 2 < n - k(n) + 1.$$

Then proceeding as in the proof of Lemma 2.5, in Nagesh S.(2014),

R.H.S. of (2.1.1) is majorised by

$$\left( \frac{1}{n^{1 + \eta_9}} \right), \eta_9 > 0.$$

Thus the proof of the Lemma is complete.

**Lemma 2.1.3 :**

For  $\Delta \leq x < 1 + \Delta$ ,  $\Delta \leq y < 1 + \Delta$ ,  $x > y$  and  $x + y \leq 1 + \Delta$ ,

$$P \{ Y_{k(l_i)}^* > c_{l_i}(x), S_{k(l_i)}^* > c_{l_i}(y) \text{ i. o. } \} = 1$$

**Proof:**

We have

$$P \{ Y_{k(l_i)}^* > c_{l_i}(x), S_{k(l_i)}^* > c_{l_i}(y) \}$$

$$= P \{ Y_{k(l_i)}^* > c_{l_i}(x) \} - P \{ Y_{k(l_i)}^* > c_{l_i}(x), S_{k(l_i)}^* \leq c_{l_i}(y) \}$$

$$= [1 - F^{k(l_i)}(c_{l_i}(x))] - [k(l_i) [F^{k(l_i)-1}(c_{l_i}(y))] [1 - F(c_{l_i}(x))]]$$

$$= 1 - \exp \{ -k(l_i) [1 - F(c_{l_i}(x))] (1 + o(1)) \} - k(l_i) \{ \exp$$

$$\begin{aligned}
 & (-k(l_i) - 1) [1 - F(c_{l_i}(y))] (1 + o(1)) \} [1 - F(c_{l_i}(x))] \\
 & = (1 + o(1)) [k(l_i) (k(l_i) - 1)] [1 - F(c_{l_i}(x))] [1 - F(c_{l_i}(y))] \\
 & \geq \text{Const. } k^2(l_i) [1 - F(c_{l_i}(x))] [1 - F(c_{l_i}(y))] \text{ for all } i \text{ large} \\
 & \geq \text{Const. } \left( \frac{1}{l_i^{(x - \Delta = \eta)}} \right) \left( \frac{1}{l_i^{(y - \Delta = \eta)}} \right), \eta > 0 \\
 & = \text{Const. } \left( \frac{1}{l_i^{(x + y - 2\Delta + 2\eta)}} \right) \\
 & = \text{Const } \left( \frac{1}{i^{\theta(x + y - 2\Delta + 2\eta)}} \right) \\
 & = \text{Const } \left( \frac{1}{i^{1 - \eta_{10}}} \right) \quad \eta_{10} > 0,
 \end{aligned}$$

Now we have to show that

$$l_{i+1} - k(l_{i+1}) + 1 - l_i > 0$$

$$\begin{aligned}
 l_{i+1} - k(l_{i+1}) + 1 - l_i & = l_{i+1} \left[ 1 - \frac{k(l_{i+1})}{l_{i+1}} + \frac{1}{l_{i+1}} + \frac{l_i}{l_{i+1}} \right] \tag{2.1.2} \\
 \frac{l_i}{l_{i+1}} & = \frac{i^\theta}{(i+1)^\theta} = \frac{i^\theta}{i^\theta \left(1 + \frac{1}{i}\right)^\theta} = \left(1 + \frac{1}{i}\right)^{-\theta} = 1 - \frac{\theta}{i}(1 + o(1))
 \end{aligned}$$

Therefore,

$$\begin{aligned}
 1 - \frac{l_i}{l_{i+1}} & = \frac{\theta}{i}(1 + o(1)) \\
 1 - \frac{l_i}{l_{i+1}} - \frac{k(l_{i+1})}{l_{i+1}} & = \frac{\theta}{i}(1 + o(1)) - \frac{k(l_{i+1})}{l_{i+1}} \\
 & = \frac{\theta}{i} \left[ 1 + o(1) - \frac{i}{\theta} \frac{k(l_{i+1})}{l_{i+1}} \right] \\
 & = i \frac{k(l_{i+1})}{l_{i+1}} \rightarrow 0, \text{ iff } \theta > \frac{1}{1 - \Delta}
 \end{aligned}$$

Therefore R.H.S. of (2.1.2) is  $= \frac{(i+1)^\theta}{i}(1 + o(1)) = i^{\theta-1} \rightarrow \infty$  as  $\theta > 1$

Thus the events  $\{ Y_{k(l_i)}^* > c_{l_i}(x), S_{k(l_i)}^* > c_{l_i}(y) \}$   $i=1,2, \dots$ , are independent for all  $i$  large. Hence by Borel-Cantelli; Lemma, the required result is obtained.

**Lemma 2.1.4:**

For all  $x \geq \Delta, y \geq \Delta$  with  $x+y > \Delta+1$  and for every  $\varepsilon > 0$ ,

$$P \left\{ Y_{k(n)}^* > c_n(x + \varepsilon), S_{k(n)}^* > c_n(y + \varepsilon) \text{ i.o.} \right\} = 0$$

**Proof:**

Define the events

$$A_n = \{ Y_{k(n)}^* > c_n(x + \varepsilon) , S_{k(n)}^* > c_n(y + \varepsilon) \}$$

and

$$B_i = \{ Y_{k(n)}^* > c_{n_i}(x + \varepsilon) , S_{k(n)}^* > c_{n_i}(y + \varepsilon) \text{ for at least one } n \in [n_i, n_{i+1}] \}$$

Where  $n_i = [i^\theta]$ ,  $i > 1$  and  $\theta = \left( \frac{1}{x + y - 2\Delta + \varepsilon/2} \right)$

Notice that

$$P \{A_n \text{ i.o. in } n\} \leq P \{B_i \text{ i. o. in } i\} \tag{2.1.3}$$

But  $P(B_i) \leq P \{ \max (X_{n_i - k(n_i) + 1}, \dots, X_{n_{i+1}}) > c_{n_i}(x + \varepsilon) ,$   
 $\text{second } \max (X_{n_i - k(n_i) + 1}, \dots, X_{n_{i+1}}) > c_{n_i}(y + \varepsilon) \}$

$$\leq [n_{i+1} - n_i + k(n_i) - 1]^2 [1 - F(c_{n_i}(x + \varepsilon))] [1 - F(c_{n_i}(y + \varepsilon))]$$

$$= \left[ \frac{n_{i+1} - n_i + k(n_i) - 1}{n_i} \right]^2 n_i [1 - F(c_{n_i}(x + \varepsilon))] n_i [1 - F(c_{n_i}(y + \varepsilon))]$$

$$\leq \text{Const} \left( \frac{1}{i^{1 + \eta_{11}}} \right) \eta_{11} > 0,$$

By Borel-Cantelli Lemma, R.H.S. of (2.1.3) is zero and thus the Lemma is proved.

**PROOF OF THE THEOREM**

By Lemmas 2.1.1, 2.1.2, 2.1.3, 2.1.4, it follows that the required limit set is contained in S. This completes the proof of the theorem.

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