On the number of zero increments of random walks with a barrier

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This is the second submission of this document.
Continuing the line of research initiated in Iksanov and Möhle (2008) and Negadajlov (2008) we investigate the asymptotic (as $n \to \infty$) behaviour of $V_n$, the number of zero increments before the absorption in a random walk with the barrier $n$. In particular, when the step of the unrestricted random walk has a finite mean, we prove that the number of zero increments converges in distribution. We also establish a weak law of large numbers for $V_n$ under a regular variation assumption.

Keywords: absorption time; recursion with random indices; random walk; undershoot.

1 Introduction and results

Let $\{\xi_k : k \in \mathbb{N}\}$ be independent copies of a random variable $\xi$ with proper and non-degenerate distribution

$$p_k := \mathbb{P}\{\xi = k\}, \quad k \in \mathbb{N}, \quad p_1 > 0.$$  

Fix $n \in \mathbb{N}$. The sequence $\{R_k^{(n)} : k \in \mathbb{N}_0 := \{0, 1, \ldots\}\}$ defined as follows: $R_0^{(n)} := 0$ and

$$R_k^{(n)} := R_{k-1}^{(n)} + \xi_k 1_{\{R_{k-1}^{(n)} + \xi_k < n\}}, \quad k \in \mathbb{N},$$

is called a random walk with the barrier $n$. The Markov chain $\{R_k^{(n)} : k \in \mathbb{N}_0\}$ is non-decreasing and has the unique absorbing state $n - 1$. Denote by

$$T_n := \inf\{k \in \mathbb{N} : R_k^{(n)} = n - 1\} = \sum_{l=1}^{\infty} 1_{\{R_l^{(n)} < n-1\}} + 1$$

the absorption time. Other interesting functionals acting on random walks with a barrier are the number of non-zero increments (jumps)

$$M_n := \#\{k \in \mathbb{N} : R_{k-1}^{(n)} \neq R_k^{(n)}\} = \sum_{l=0}^{\infty} 1_{\{R_l^{(n)} + \xi_{l+1} < n\}}$$

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and the number of zero increments before the absorption time

\[ V_n := \# \{ k \in [0, T_n - 1] : R^{(n)}_{k-1} = R^{(n)}_k \} = \sum_{l=0}^{T_n-1} 1_{\{R^{(n)}_{l} + \xi_{l+1} \geq n\}}. \]

Notice that the last summand of the latter sum always equals zero. We however do not replace the upper limit of the summation by \( T_n - 2 \) to avoid summation over the empty set.

Two real-world examples where the random walk with a barrier naturally arises are given next.

(a) An enterprise can pay 500000 Euro per month as a salary to all called-in employees. Potential employees apply for salaries which are independent copies of a random variable \( \xi \) taking values 100, 200, 300, \ldots Euro. Then \( T_{500001}, M_{500001} \) and \( V_{500001} \) define the number of people which apply for a job, to be hired and to be turned down, respectively.

(b) A private company (an express, say) specializes in transportation of goods and uses three lorries which altogether are able to transport 10 tones. Assume that requests for transportation are independent copies of a random variable \( \xi \) taking values in the set of multiples of 100 kg. Then \( T_{10001}, M_{10001} \) and \( V_{10001} \) define the total number of requests, the number of requests to be satisfied and to be rejected, respectively.

In \([5]\) and \([7]\) the weak convergence of properly normalized and centered \( M_n \) and \( T_n \), respectively, was investigated. Assuming that the distribution of \( \xi \) belongs to the domain of attraction of a stable law, a complete characterization of normalizing constants and possible limiting laws was obtained. The class of limiting laws is comprised of stable laws and laws of the integrals \( \int_0^{\infty} e^{-\lambda t} dt \) where \( \{X_t : t \geq 0\} \) is a zero-drift subordinator with known Lévy measure.

Furthermore, in \([7]\) it was proved that all the results obtained in \([5]\) for \( M_n \) remain valid with \( M_n \) replaced by \( T_n \). From this last observation it follows that the number of zero increments \( V_n = T_n - M_n \) is "essentially smaller" than the number of non-zero increments \( M_n \).

The aim of the paper is to investigate asymptotic behaviour of \( V_n \) as \( n \to \infty \). In Theorem 1.1 it is proved that under the assumption \( m := \mathbb{E} \xi < \infty \) the sequence \( \{V_n : n \in \mathbb{N}\} \) converges in distribution. In Theorem 1.2 assuming that \( m = \infty \) and that the tail of distribution of \( \xi \) is regularly varying at \( \infty \), a weak law of large numbers is established. All the proofs rely upon the following observation: marginal distributions of the sequence \( \{V_n : n \in \mathbb{N}\} \) satisfy the recursions \([9]\) which we call recursions with random indices. It is well-known that recursive structure is typical for many objects arising in the Analysis of Algorithms, in the theory of random graphs, in the coalescent theory, etc. In this paper, we exhibit a relatively new model where recursions appear in a natural way and exploiting them turns out to be a right way of thinking.

Before formulating the main results, notice that Hinderer and Walk \([4]\) investigated processes more general than random walks with a barrier but the circle of problems they considered was different from ours. As far as we know, the term "random walk with a barrier" was introduced in \([5]\).

**Theorem 1.1** If \( \mathbb{E} \xi < \infty \), then, as \( n \to \infty \),

\[ V_n \overset{d}{\to} V := V_Y + 1 - 2 \cdot 1_{\{Y=1\}}, \]

where \( Y \) is a random variable with distribution

\[ \mathbb{P}\{Y = k\} = (\mathbb{E} \xi)^{-1} \mathbb{P}\{\xi \geq k\}, \quad k \in \mathbb{N}, \]

which is independent of \( \{V_n : n \in \mathbb{N}\} \). In particular, \( \mathbb{P}\{V = 0\} = (\mathbb{E} \xi)^{-1} \).
Throughout the paper by $\psi(x) := \frac{\Gamma'(x)}{\Gamma(x)}$ we denote the logarithmic derivative of the gamma function. Also, set
\[ m(x) := \int_0^x P\{\xi > y\}dy, \quad x > 0. \]

**Theorem 1.2** Suppose that for some $\alpha \in (0, 1]$ and some $L$ slowly varying at $+\infty$
\[ \sum_{k=n}^{\infty} p_k \sim n^{-\alpha} L(n), \quad n \to \infty. \tag{1} \]
Then, as $n \to \infty$,
\[ V_n \xrightarrow{\mathbb{E}V_n} 1, \tag{2} \]
and $\mathbb{E}V_n \sim (\psi(1) - \psi(1 - \alpha))^{-1} \log n$, if $\alpha \in (0, 1)$, and $\mathbb{E}V_n \sim \log m(n)$, if $\alpha = 1$.

The rest of the paper is organized as follows: in Section 2 we discuss certain recursions with random indices which naturally arise in the context of this work and in Section 3 the proofs of our main results are given.

**Notation:** the record $r(\cdot) \sim s(\cdot)$ means that $r(\cdot)/s(\cdot) \to 1$, as an argument goes to $+\infty$; the record $X_n \xrightarrow{d} (\Rightarrow, \to \mathbb{P}) X$ means that the limit relation holds as $n \to \infty$.

**2 Recursions with random indices**

For fixed $m, i \in \mathbb{N}$ define $\hat{R}^{(m)}_0(i) := 0$ and
\[ \hat{R}^{(m)}_k(i) := \hat{R}^{(m)}_{k-1}(i) + \xi_{i+k}1_{\{\hat{R}^{(m)}_{k-1}(i) + \xi_{i+k} < m\}}, \quad k \in \mathbb{N}, \]
and
\[ \hat{T}^{(m)}_n(i) := \sum_{l=1}^{\infty} 1_{\{\hat{R}^{(m)}_l(i) + \xi_{i+l} < m\}}, \quad \hat{V}^{(m)}_n(i) := \sum_{l=0}^{\hat{R}^{(m)}_n(i)+\xi_{i+n}+1} 1_{\{\hat{R}^{(m)}_l(i) + \xi_{i+l}+1 = m\}}. \]

For fixed $n \in \mathbb{N}$ and any $i \in \mathbb{N}$ marginal distributions of the sequences $\{R^{(n)}_k : k \in \mathbb{N}_0\}$ and $\{R^{(n)}_k(i) : k \in \mathbb{N}_0\}$ are the same. Therefore,
\[ \hat{T}^{(m)}_n(i) \xrightarrow{d} T_n, \quad \text{AND} \quad \hat{V}^{(m)}_n(i) \xrightarrow{d} V_n. \tag{3} \]

Introduce the notation $S_0 := 0, S_n := \xi_1 + \cdots + \xi_n$ and
\[ N_n := \inf\{k \geq 1 : S_k \geq n\} \quad \text{AND} \quad Y_n := n - S_{N_n-1}, \quad n \in \mathbb{N}. \]

The sequence $\{S_n : n \in \mathbb{N}_0\}$ is an unrestricted random walk on the basis of which the random walk with barrier is constructed. The random variable $Y_n$ is called *undershoot at* $n$.

With probability one, we have
\[ T_n = (1 + \hat{T}^{(m)}_n(N_n))1_{\{Y_n \neq 1\}} + N_n - 1 = \hat{T}^{(m)}_n(N_n) + N_n - 2 \cdot 1_{\{Y_n = 1\}}, \tag{4} \]
which can be checked as follows. If \( n - R_{N_n-1}^{(n)} = n - S_{N_n-1} = Y_n = 1 \), then \( T_n = N_n - 1 \), and if \( Y_n \neq 1 \), then \( R_{N_n-1}^{(n)} = S_{N_n-1} < n - 1 \). Therefore,

\[
T_n = N_n - 1 + \left( 1 + \sum_{l=1}^{\infty} 1\{R_{N_n+i}^{(n)} < n-1\} \right) 1\{Y_n \neq 1\} = N_n - 1 + (1 + \hat{T}_n(N_n)) 1\{Y_n \neq 1\}.
\]

Similarly, with probability one we have

\[
V_n = (\hat{V}_n(N_n) + 1) 1\{Y_n \neq 1\} = \hat{V}_n(N_n) + 1 - 2 \cdot 1\{Y_n = 1\}, \tag{5}
\]

which can be verified as follows. If \( n - R_{N_n-1}^{(n)} = n - S_{N_n-1} = Y_n = 1 \), then \( V_n = 0 \), and if \( Y_n \neq 1 \), then the first non-zero increment of the random walk with barrier is \( R_{N_n}^{(n)} - R_{N_n-1}^{(n)} \), and \( T_n \geq N_n + 1 \). Therefore,

\[
V_n = (1 + \sum_{l=N_n}^{T_{n-1}-1} 1\{R_{N_n+i}^{(n)} \geq \xi_{N_n+i+1} \geq n\}) 1\{Y_n \neq 1\} = (1 + \sum_{l=0}^{T_{n-1}-N_n-1} 1\{R_{N_n+i}^{(n)} \geq \xi_{N_n+i+1} \geq n\}) 1\{Y_n \neq 1\} = (1 + \hat{V}_n(N_n)) 1\{Y_n \neq 1\}.
\]

\( \hat{V}_n \) is independent of the sequence \( \{V'_k : k \in \mathbb{N}\} \) which is a copy of \( \{V_k : k \in \mathbb{N}\} \).

3 Proofs

Letting \( n \to \infty \) in the equality of distributions \( \hat{V}_n \) and taking into account the convergence \( Y_n \to_d Y \) (see, for example, p. 371 in \[3\]) the result of Theorem 1.1 follows.

To prove Theorem 1.2 we need an auxiliary result.

**Lemma 3.1** If \( \alpha \) holds with \( \alpha \in [0,1) \), then, as \( n \to \infty \),

\[
\mathbb{E} \log Y_n = \log n - (\psi(1) - \psi(1 - \alpha)) + o(1) \text{ AND} \\
\mathbb{E} \log^2 Y_n = \log^2 n - 2(\psi(1) - \psi(1 - \alpha)) \log n + o(\log n).
\]

If \( \alpha \) holds with \( \alpha = 1 \), then, as \( n \to \infty \),

\[
\mathbb{E} \log m(Y_n) = \log m(n) - 1 + o(1) \text{ AND} \\
\mathbb{E} \log^2 m(Y_n) = \log^2 m(n) - 2 \log m(n) + o(\log m(n)).
\]
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**Proof:** If \( (1) \) holds with \( \alpha \in [0, 1) \), then according, for example, to Theorem 8.6.5 in [1]

\[
\log n - \log Y_n \rightarrow_d (- \log \eta_\alpha),
\]

where a random variable \( \eta_\alpha \), \( \alpha \in (0, 1) \), has the beta distribution with parameters \( 1 - \alpha \) and \( \alpha \), i.e.

\[
P\{\eta_\alpha \in dx\} = \frac{\sin \pi \alpha}{\pi} x^{-\alpha}(1-x)^{1-\alpha} dx, \ x \in (0, 1), \text{ and } P\{\eta_0 = 1\} = 1.
\]

If we can show that for each \( \epsilon \in (0, 1 - \alpha) \)

\[
\sup_{n \geq 1} \mathbb{E}(n/Y_n)^\delta = \sup_{n \geq 1} \mathbb{E}f_k(\log^k(n/Y_n)) < \infty, \ k \in \mathbb{N}, \quad (9)
\]

where \( f_k(x) := \exp(\delta x^{1/k}) \), then by a Vallée-Poussin theorem (see [6]) for each \( k \in \mathbb{N} \) the sequence \( \{(\log n - \log Y_n)^k : n \in \mathbb{N}\} \) is uniformly integrable. If it is true then

\[
\lim_{n \to \infty} \log n - \log Y_n = \mathbb{E}(- \log \eta_\alpha) = \psi(1) - \psi(1 - \alpha) < \infty \text{ which proves (7), and } \lim_{n \to \infty} \mathbb{E}(\log n - \log Y_n)^2 = \mathbb{E}\log^2 \eta_\alpha = (\psi(1 - \alpha) - \psi(1))^2 + \psi'(1 - \alpha) - \psi'(1) < \infty, \text{ which together with (7) proves (8).}
\]

Let us establish (9). For \( j \in \mathbb{N}_0 \) set \( u_j := \sum_{k=0}^{j-1} P\{S_k = j\}. \) Then

\[
\mathbb{E}Y_n^{-\delta} = \sum_{k=0}^{n-1} (n-k)^{-\delta} P\{\xi \geq n-k\} u_k.
\]

The condition (1) implies that \( \sum_{k=0}^{n} u_k \sim \frac{1}{\Gamma(1 - \alpha)} \frac{1}{1+\alpha} L^{n_0} \) (see, for example, Theorem 8.7.3 in [1]). The following four relations hold by Corollary 1.7.3 in [1]:

\[
U(s) := \sum_{n=0}^{\infty} s^n u_n \sim \frac{1}{\Gamma(1 - \alpha)} \frac{1}{\Gamma((1-s)^{-1}1+\alpha)} \quad \text{AS } s \uparrow 1;
\]

\[
V(s) := \sum_{n=1}^{\infty} s^n n^{-\delta} P\{\xi \geq n\} \sim \frac{\Gamma(1 - \alpha - \delta)}{\Gamma(1 - \alpha)} \frac{1}{(1-s)^{1-\alpha-\delta}} \quad \text{AS } s \uparrow 1;
\]

\[
\sum_{n=1}^{\infty} s^n \mathbb{E}Y_n^{-\delta} = U(s)V(s) \sim \frac{\Gamma(1 - \alpha - \delta)}{\Gamma(1 - \alpha)} \frac{1}{(1-s)^{1-\delta}} \quad \text{AS } s \uparrow 1;
\]

\[
\sum_{k=1}^{n} \mathbb{E}Y_k^{-\delta} \sim \frac{\Gamma(1 - \alpha - \delta)}{\Gamma(1 - \alpha)} \frac{1}{\Gamma(2 - \alpha)} n^{1-\delta}.
\]

The last equivalence entails (9).

Assume now that (1) holds with \( \alpha = 1 \). By Theorem 6 in [2] and its proof,

\[
\log m(n) - \log m(Y_n) \rightarrow_d (- \log R),
\]

where \( R \) is a random variable with the uniform distribution on \([0, 1]\). Therefore, the stated result will be proved if, for example, we can show that the sequences \( \{(\log m(n) - \log m(Y_n))^k : n \in \mathbb{N}\}, k = 1, 2, \) are uniformly integrable. This latter fact will follows from the following relation: for each \( \epsilon \in (0, 1) \)

\[
\sup_{n \geq 1} \mathbb{E} \left( \frac{m(n)}{m(Y_n)} \right)^\epsilon < \infty. \quad (10)
\]
The proof that follows is similar to the previous one. Therefore, we only give a sketch. As before, all the unexplained equivalences connected with asymptotic behaviour of the sequences and corresponding generating functions can be justified by an appeal to Corollary 1.7.3 in [1]. Since \( \sum_{k=0}^{n} u_k \sim n/m(n) \) (see p. 266 in [2]), then

\[
K(s) := \sum_{n=0}^{\infty} s^n u_n \sim \frac{1}{(m((1-s)^{-1}))(1-s)} \quad \text{as } s \uparrow 1.
\]

Fix any \( \varepsilon \in (0,1) \). The function \( r_{\varepsilon}(x) := \int_{0}^{x} m^{-\varepsilon}(y) P\{\xi \geq y\} dy \) is slowly varying at \( \infty \), and \( \sum_{k=1}^{\infty} m^{-\varepsilon}(k) P\{\xi \geq k\} \sim r_{\varepsilon}(n) \). Consequently,

\[
Z(s) := \sum_{n=1}^{\infty} s^n m^{-\varepsilon}(n) P\{\xi \geq n\} \sim r_{\varepsilon}((1-s)^{-1}) \quad \text{as } s \uparrow 1.
\]

Hence, as \( s \uparrow 1 \),

\[
\sum_{n=1}^{\infty} s^n E m^{-\varepsilon}(Y_n) = K(s) Z(s) \sim \frac{r_{\varepsilon}((1-s)^{-1})}{(m((1-s)^{-1}))(1-s)}.
\]

and

\[
\sum_{k=1}^{n} E m^{-\varepsilon}(Y_k) \sim \frac{nr_{\varepsilon}(n)}{m(n)}.
\]  

(11)

By L’Hospital’s rule, \( r_{\varepsilon}(n)/m(n) \sim (1-e)^{-1}m^{-\varepsilon}(n) \). Therefore, (11) implies (10). □

Proof of Theorem 1.2 Suppose [1] holds with \( \alpha \in (0,1) \). Set \( a_n := EV_n \) and \( b_n := EV_n^2 \). From (6) we obtain \( a_1 = 1 \),

\[
a_n = \frac{1}{1 - P\{Y_n = n\}} \left( \sum_{k=1}^{n-1} a_k P\{Y_n = k\} + 1 - 2P\{Y_n = 1\} \right), \quad n = 2, 3, \ldots
\]  

(12)

and \( b_1 = 1 \),

\[
b_n = \frac{1}{1 - P\{Y_n = n\}} \left( \sum_{k=1}^{n-1} b_k P\{Y_n = k\} + 2a_n - 1 \right).
\]  

(13)

Let us check that

\[
b_n \sim a_n^2 \sim k_n^2 \log^2 n,
\]  

(14)

where \( k_n := (\psi(1) - \psi(1-\alpha))^{-1} \). Suppose the condition \( \lim_{n \to \infty} \frac{a_n}{\log n} \leq k_n \) does not hold. Then there exists an \( \varepsilon > 0 \) such that the inequalities \( a_n > (k_n + \varepsilon) \log n \) hold for infinitely many values of \( n \). It is possible to decrease \( \varepsilon \) so that the inequality \( a_n > (k_n + \varepsilon) \log n + c \) holds infinitely often for any fixed positive \( c \). Thus we can define \( n_c := \inf\{n \geq 1 : a_n > (k_n + \varepsilon) \log n + c \} \).

Then

\[
a_n \leq (k_n + \varepsilon) \log n + c \quad \text{for all } n \in \{1, 2, \ldots, n_c - 1\}.
\]

From (12) it follows that

\[
(k_n + \varepsilon) \log n_c + c <
\]
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\[
\begin{align*}
&\frac{1}{1 - \mathbb{P}\{Y_{n_c} = n_c\}} \left( \sum_{i=1}^{n_c-1} \left( (k_\alpha + \epsilon) \log i + \epsilon \right) \mathbb{P}\{Y_{n_c} = i\} + 1 - 2\mathbb{P}\{Y_{n_c} = 1\} \right) = \\
&= c + (k_\alpha + \epsilon) \left( \mathbb{E} \log Y_{n_c} - \mathbb{E} \log Y_{n_e} \right) + \\
&\quad + \frac{1}{1 - \mathbb{P}\{Y_{n_c} = n_c\}} \left( 1 - 2\mathbb{P}\{Y_{n_c} = 1\} \right) = \\
(n_c \to \infty \text{ as } c \to \infty; \text{ therefore, using (7), the equality } \mathbb{P}\{Y_n = 1\} = u_n - 1 \text{ and the fact that by the elementary renewal theorem } \lim_{n \to \infty} u_n = 0, \text{ we obtain}) \\
&= c + (k_\alpha + \epsilon) \log n_c - (k_\alpha + \epsilon)k_\alpha^{-1} + 1 + o(1).
\end{align*}
\]

Letting \( c \) go to \( \infty \) leads to \( \epsilon \leq 0 \), which is a contradiction. Thus, we have already checked that

\[
\lim_{n \to \infty} \frac{a_n}{\log n} \leq k_\alpha.
\]

A symmetric argument proves the converse inequality for the lower bound. Therefore,

\[
a_n \sim k_\alpha \log n.
\]

The asymptotic relation (14) for \( b_n \) can be checked exactly in the same way by using the recursion (13). Thus, we already know that \( \mathbb{E}V_n^2 \sim (\mathbb{E}V_n)^2 \). Therefore, by Chebyshev’s inequality, for any \( \delta > 0 \)

\[
\mathbb{P}\{|V_n/\mathbb{E}V_n - 1| > \delta\} \leq \frac{\mathbb{D}V_n}{(\delta \mathbb{E}V_n)^2} \to 0,
\]

which proves (2).

In the case when (1) holds with \( \alpha = 1 \), the proof of the fact that \( \mathbb{E}V_n^2 \sim (\mathbb{E}V_n)^2 \sim (\log m(n))^2 \) is absolutely analogous to the previous one. Thus it is omitted. Certainly, instead of the first part of Lemma 3.1 one has to use the second one. \( \square \)

References


