

## THE DEPTH FIRST PROCESSES OF GALTON-WATSON TREES CONVERGE TO THE SAME BROWNIAN EXCURSION

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In this paper, we show a strong relation between the depth first processes associated to Galton Watson trees with finite variance, conditioned by the total progeny: the depth first walk, the depth first queue process, the height process; a consequence is that these processes (suitably normalized) converge to the same Brownian excursion. This provides an alternative proof of Aldous' one of the convergence of the depth first walk to the Brownian excursion which does not use the existence of a limit tree. The methods that we introduce allow one to compute some functionals of trees or discrete excursions; for example, we compute the limit law of the process of the height of nodes with a given out-degree, and the process of the height of nodes, root of a given subtree.

**1. Introduction** In this paper, we study some depth first processes associated to Galton-Watson trees conditioned by their total progeny. Consider  $\xi$ , a non-negative, integer-valued random variable that satisfies:

$$(1) \quad \begin{cases} \mathbb{E}(\xi) = 1, & 0 < \text{Var}(\xi) = \sigma^2 < +\infty \\ \text{there exists a constant } \alpha > 0 \text{ s.t. } \mathbb{E}(e^{\alpha\xi}) < +\infty. \end{cases}$$

We consider a Galton-Watson branching process with offspring  $\xi$ , starting with 1 individual in generation 0. We write  $\tau$  for the family tree of this branching process and  $\Omega$  the probability space of all trees with the law induced by  $\xi$ . We note  $\Omega_n$  the space of size  $n$  trees endowed by the conditional law given  $|\tau| = n$ .

*The depth first search* Let  $\tau$  be an ordered tree with  $n$  nodes. We define a function (see Aldous (1991) p.260):

$$\tilde{f} : \{0, \dots, 2n - 2\} \longrightarrow \{\text{nodes of } \tau\},$$

which we regard as a walk around  $\tau$ , as follows:

$$\tilde{f}(0) = \text{root}.$$

Given  $\tilde{f}(i) = v$ , choose, if possible, the left-most child  $w$  of  $v$  which has not already been visited, and set  $\tilde{f}(i + 1) = w$ . If not possible, let  $\tilde{f}(i + 1)$  be the parent of  $v$ .

**1.1. The Processes** **The depth first walk (DFW):** For a tree  $\tau \in \Omega_n$ , we call the DFW of  $\tau$ , the process  $V_n$  defined by:

$$V_n(i) = d(\text{root}, \tilde{f}(i)), \quad 0 \leq i \leq 2n - 2.$$

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For  $i$  from 0 to  $n-1$ , let  $v_i$  be the  $i$ th new node visited by the depth first procedure on  $\tau \in \Omega_n$  ( $v_0 = \text{root}$ ) and

$$\xi_i = \text{the outdegree of } v_i = \text{the number of children of } v_i.$$

**The depth first queue process (DFQP):** For a tree  $\tau \in \Omega_n$ , the DFQP,  $S_n$  is defined by  $S_n(0) = 0$  and:

$$(2) \quad S_n(j) = \sum_{i=0}^{j-1} (\xi_i - 1) \text{ for any } 1 \leq j \leq n.$$

**The height process:** The height process of  $\tau \in \Omega_n$ ,  $h_n$ , is defined by

$$h_n(i) = d(\text{root}, v_i) \text{ for } 0 \leq i \leq n-1.$$

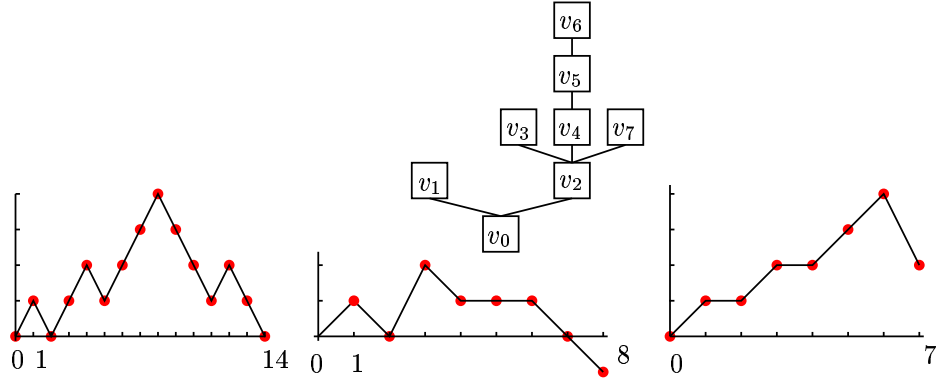


FIGURE 1. A tree and its associated DFW, DFQP, and height process.

Note that the DFW, the DFQP as well as the height process characterize the associated tree.

**The height of nodes with a given out-degree:** Consider  $j$  such that  $p_j \neq 0$ ; for a tree  $\tau \in \Omega_n$ , the process  $h_n^{(j)}$  is defined by  $h_n^{(j)}(0) = 0$  and:

$$h_n^{(j)}(m) = d(\text{root}, \tilde{v}_m) \text{ for } m \geq 1$$

where  $\tilde{v}_m$  is the  $m$ th node with out-degree  $j$  in the list  $(v_0, v_1, \dots, v_{n-1})$ . If such a node does not exist, set  $h_n^{(j)}(m) = 0$ .

**The height of nodes, root of a given type subtree:** Let  $u$  denote a fixed tree in  $\Omega$  with  $|u| = \lambda$  and such that

$$\mathbb{P}(u) > 0.$$

For a tree  $\tau$  from  $\Omega_n$ , the process  $h_n^u$  is defined by  $h_n^u(0) = 0$  and:

$$h_n^u(m) = d(\text{root}, \bar{v}_m) \text{ for } m \geq 1,$$

where  $\bar{v}_m$  is the  $m$ th node in the list  $(v_0, v_1, \dots, v_{n-1})$ , root in  $\tau$  of a subtree equal to  $u$ ; if no such node exists, set  $h_n^u(m) = 0$ .

1.2. *Comments and previous works* Our paper is cast in term of a critical Galton-Watson process. The simply generated trees in the sense of Meir & Moon (1978) (binary trees, general order trees, Cayley trees, Schröder trees, etc...), can be seen as conditioned critical Galton-Watson trees with an *ad hoc* offspring distribution (see Aldous (1991) or Drmota & Gittenberger (1997) for a translation index).

As noted by Kennedy (1976), the distribution on  $\Omega_n$  is unchanged if we replace  $\xi$  with another progeny  $\xi'$ , in the same exponential family. Then, there is no loss of generality in considering critical branching process.

Our aim is to show strong relations between the five depth processes presented above; a consequence of these relations is that these five processes (suitably normalized) converge to the same Brownian excursion.

Lately, a lot of works have been published on the properties of depth search processes as well as on breadth first search processes. One can cite for example, the works of Aldous (1991, 1993, 1998), Drmota and Gittenberger (1997), Takàcs (1993), Chassaing & al. (2000).

In recent works, Le Gall (2000), Le Gall & Le Jan (1998) and Le Gall & Duquesne (2001) investigated the asymptotic distribution of the height process of a sequence of independent Galton-Watson trees; in these works, the offspring variance can be infinite. The limit height process obtained is a Lévy process, which is the reflected Brownian motion, when the variance is finite. One can also mention the works on LIFO queue of Limic (2000,2001) who uses the connection between queueing models and branching processes, particularly between the queue length and the height process.

The limit of  $V_n$  is known; in his famous paper *Continuum random tree III*, Aldous has shown that the DFW, suitably normalized converges to the normalized standard excursion. Here are the main lines of his proof: he first shows that the Galton-Watson tree conditioned by its total progeny converges (suitably normalized) to a limit object: the continuum random tree. Next, he proves that the distribution of the continuum random tree is unique; it does not depend on the law of the progeny (up to a change of scale). He uses then the fact that for the geometrical offspring distribution, the limit of the DFW is the Brownian excursion; thus, it is also the limit of the DFW for the other families of trees.

In the present paper, we assume that  $\xi$  has exponential moments, though Aldous assumes only the existence of the variance. The exponential moment assumption is necessary to obtain exponential concentration type inequalities in the theorems. We come back on this difference in the conclusion.

We think, that the new approach presented in the present paper has several interests.

- The proofs given here are simple and just rely on well known parts of probability theory: the ladder height and ladder epochs of simple random walk, the law of large number and concentration inequalities.
- This new approach provides (by Theorem 2) a strong relation between the DFQP and the DFW (illustrated by the figures at the end of the paper). This relation between these two processes seems to be unknown. Using large or moderate deviations

on DFQP which are rather easy to obtain (DFQP is an usual discrete excursion), one can obtain (upper bounds on) large or moderate deviations on some functionals on the tree: the height, the mean of the depth of the nodes,...

- The methods used are sufficiently general to provide concentration inequalities for the uniform distance between the five processes described in the beginning of the paper.
- The unicity of the continuum random tree: Theorem 1 below shows that the DFW converges to the Brownian excursion, independently from the offspring distribution (it depends only on the variance of  $\xi$ ). The strong “geometrical” relation between a tree and its DFW allows one to have another point of view on the fact that the limit tree - the continuum random tree defined in Aldous (1993) - is the same whatever the distribution of  $\xi$  is. Applying Theorem 1 to lattice distributions, we obtain that in this case also, the limit tree of Galton-Watson trees with lattice distributions is the standard continuum random tree. Indeed, the limit tree is entirely characterized by the limit DFW (see Theorem 20 of Aldous (1993)).

The paper is organized as follows: in Section 2, we present the main results. Section 3 is devoted to the proofs. In 3.1, we exhibit some relations between  $V_n$  and, in one hand, the depth first queue, and in the other hand, the height process.  $V_n$  appears then to be closely related to the number of right minima in a discrete excursion; in Section 3.2, we explain how to reduce the study on discrete excursions to the same study on (non-conditioned) random walks. The ends of the proofs are given in Section 3.3 - 3.6. Section 4 is devoted to simulations that illustrate Theorem 2.

**2. Results** *Lattice considerations:*  $\xi$  may be lattice distributed. In this case, there exists an infinite number of  $n$  for which  $\Omega_n$  is empty. In the sequel of the paper, we note  $PP(\xi)$  the set of “possible population sizes” for the trees in  $\Omega$ , that is:

$$n \in PP(\xi) \Leftrightarrow \Omega_n \neq \emptyset.$$

*Convergence spaces:* In the sequel of the paper, the five processes described in the Introduction will be considered as continuous processes, with linear interpolation between the integral points. The weak convergence considered in our results is the convergence in  $C([0, 1], \|\cdot\|_\infty)$ .

A first interest of this paper is to give a simple proof of the following Aldous’ Theorem:

**THEOREM 1.** *Let  $V_n$  be the DFW for  $\tau \in \Omega_n$ . For  $n \in PP(\xi)$ ,*

$$\left( \frac{V_n(2nt)}{\sqrt{n}} \right)_{t \in [0,1]} \xrightarrow{\text{weakly}} \left( \frac{2}{\sigma} e(t) \right)_{t \in [0,1]}$$

*where  $(e(t))_{t \in [0,1]}$  is a standard normalized Brownian excursion.*

The proof of this theorem relies on the comparison with the DFQP:

**THEOREM 2.** *For any  $\nu > 0$ , there exist two constants  $\gamma > 0$  and  $N > 0$  such that for all  $n \geq N$  and  $n \in PP(\xi)$ ,*

$$\mathbb{P} \left( \sup \left\{ \left| S_n(nt) - \frac{\sigma^2}{2} V_n(2nt) \right|, t \in [0, 1] \right\} \geq n^{1/4+\nu} \right) \leq e^{-\gamma n^\nu}.$$

The limit of the DFQP is well known. We state it in the following Proposition and we give a short proof:

**PROPOSITION 1.** *For  $n \in PP(\xi)$ , we have*

$$\left( \frac{S_n(nt)}{\sqrt{n} \sigma} \right)_{t \in [0,1]} \xrightarrow{\text{weakly}} (e(t))_{t \in [0,1]}.$$

*Proof:* Note by  $(p_i)_{i \geq 0}$  the distribution of  $\xi$ , that is  $p_i = \mathbb{P}(\xi = i)$ . For a tree  $\tau$  from  $\Omega$ , the r.v.  $\xi - 1$  is  $\tilde{p}_i = p_{i+1}$ ,  $\forall i \geq -1$  distributed. By definition (2),  $S_n$  is the process of sums of the r.v.  $(\xi_i - 1)$ . This is a classical fact for the search queue (see Cormen & al. (1990) p. 473-474 or Chassaing & Marckert (2001)), that the queue gets empty only when all the nodes have been visited. The condition  $|\tau| = n$  on the tree is then equivalent to the condition:

$$\{S_n(0) = 0, S_n(n) = -1, S_n(i) \geq 0 \text{ for each } i, 1 \leq i \leq n-1\}.$$

Thus,  $(S_n(j))_{0 \leq j \leq n}$  is a discrete excursion. Since the weak convergence of a discrete excursion to a Brownian excursion is well known Csaki & Mohanty (1981) and Durrett & al. (1977), the proposition is proved.  $\square$

Theorem 1 is then a corollary of Theorem 2; note that Theorem 2 shows that, even if  $V_n$  is not a conditioned random walk (except if  $\xi$  is geometrically distributed) it is “close” to the discrete excursion  $S_n$  for the uniform distance. Some illustrations by simulations are given at the end of the paper.

*The height process* At the end of Subsection 3.1, we will show that the DFW “interpolates” the height process. Theorem 2 and the following Theorem 3 express that  $h_n$  and  $V_n$  have very similar properties up to a change of scale with factor 2.

**THEOREM 3.** *For any  $\nu > 0$ , there exist two constants  $\gamma > 0$  and  $N > 0$  such that for all  $n \geq N$  and  $n \in PP(\xi)$ ,*

$$\mathbb{P} \left( \sup \left\{ \left| S_n(nt) - \frac{\sigma^2}{2} h_n(nt) \right|, t \in [0, 1] \right\} \geq n^{1/4+\nu} \right) \leq e^{-\gamma n^\nu}.$$

**COROLLARY 1.** *For  $n \in PP(\xi)$ , we have*

$$\left( \frac{h_n(nt)}{\sqrt{n}} \right)_{t \in [0,1]} \xrightarrow{\text{weakly}} \left( \frac{2}{\sigma} e(t) \right)_{t \in [0,1]}.$$

The methods and arguments used in the proofs of Theorems 2 and 3 allow one to derive similar results for other processes associated to the depth first search:

*The height of nodes with a given out-degree* Let  $h_n^{(j)}$  be the process of the height of nodes with out-degree  $j$  for  $\tau \in \Omega_n$  (by convention, the height of a non-existing node is 0).

**THEOREM 4.** *For any  $\nu > 0$ , there exist two constants  $\gamma > 0$  and  $N > 0$  such that for all  $n \geq N$  and  $n \in PP(\xi)$ ,*

$$\mathbb{P} \left( \sup \left\{ \left| h_n(nt) - h_n^{(j)}(p_j nt) \right|, t \in [0, 1] \right\} \geq n^{1/4+\nu} \right) \leq e^{-\gamma n^\nu}$$

**COROLLARY 2.** *For  $n \in PP(\xi)$ , we have*

$$\left( \frac{h_n^{(j)}(p_j nt)}{\sqrt{n}} \right)_{t \in [0,1]} \xrightarrow{\text{weakly}} \left( \frac{2}{\sigma} e(t) \right)_{t \in [0,1]}.$$

This theorem reveals the “good repartition” of the  $j$ -nodes among all the nodes. It also extends the results obtained by Gutjahr & Pflug (1992) and Marckert (2000) about the height process of the leaves of binary trees (on geometric trees). In his paper, Drmota (1994) gives the law of the height of the  $i$ th leaf; this can also be obtained by Corollary 2 (with  $j = 0$ ) using the density of the Brownian excursion (Durrett & al. (1977)):

$$\mathbb{P}(e(t) \in dy) = \frac{2y^2 \exp(-y^2/2t(1-t))}{(2\pi t^3(1-t)^3)^{1/2}} dy.$$

*Occurrences of a given subtree* Note  $u$  a fixed tree with size  $\lambda$  in  $\Omega$  such that

$$q_u = \mathbb{P}(u) > 0.$$

For  $\tau \in \Omega_n$ , let  $h_n^u$  be the process of the height of nodes that are root of a subtree equal to  $u$ . We have:

**THEOREM 5.** *For any  $\nu > 0$ , there exist two constants  $\gamma > 0$  and  $N > 0$  such that for all  $n \geq N$  and  $n \in PP(\xi)$ ,*

$$\mathbb{P} \left( \sup \{ |h_n(nt) - h_n^u(q_u nt)|, t \in [0, 1] \} \geq n^{1/4+\nu} \right) \leq e^{-\gamma n^\nu}$$

**COROLLARY 3.** *For  $n \in PP(\xi)$ , we have*

$$\left( \frac{h_n^u(q_u nt)}{\sqrt{n}} \right)_{t \in [0,1]} \xrightarrow{\text{weakly}} \left( \frac{2}{\sigma} e(t) \right)_{t \in [0,1]}.$$

Once again, Theorem 5 reveals the good repartition of the subtrees with a given type in the trees  $\tau \in \Omega_n$ .

We refer to Flajolet & al. (1997) and Steyaert & Flajolet (1983) for the same problematic on random binary search trees (recall that BST are not conditioned Galton-Watson trees).

The proof of Theorem 5 allows one to derive a concentration inequality for the number of subtrees with type  $u$  in the trees of  $\Omega_n$ . For a tree  $\tau$  from  $\Omega_n$ , we note  $N_u(n)$  the number of occurrences of  $u$  in  $\tau$ , that is, the number of nodes in  $\tau$  root of a subtree equal to  $u$ .

COROLLARY 4. For any  $\nu > 0$ , there exist two constants  $\gamma > 0$  and  $N > 0$  such that, for all  $n \geq N$  and  $n \in PP(\xi)$ ,

$$\mathbb{P}(|N_u(n) - q_u n| \geq n^{1/2+\nu}) \leq e^{-\gamma n^\nu}.$$

*Conclusion:* All the depth first processes have the same limit. Note  $DF_n$  the process indexed by  $[0, 1]^5$ :

$$DF_n(t_1, t_2, t_3, t_4, t_5) = \left( \frac{S_n(nt_1)}{\sigma\sqrt{n}}, \frac{V_n(2nt_2)}{\sqrt{n}}, \frac{h_n(nt_3)}{\sqrt{n}}, \frac{h_n^{(j)}(p_j nt_4)}{\sqrt{n}}, \frac{h_n^u(q_u nt_5)}{\sqrt{n}} \right).$$

As a consequence of the four last theorems, we have the convergence of the five depth first processes to the same Brownian excursion.

$$DF_n(t_1, t_2, t_3, t_4, t_5)_{[0,1]^5} \xrightarrow{\text{weakly}} \left( e(t_1), \frac{2}{\sigma}e(t_2), \frac{2}{\sigma}e(t_3), \frac{2}{\sigma}e(t_4), \frac{2}{\sigma}e(t_5) \right)_{[0,1]^5}.$$

### 3. Proof

3.1. *Relation between the three depth processes DFW, DFQP,  $h_n$*  The relations proved in the present subsection are also noted in Le Gall & Le Jan (1998), Le Gall (2000) and Le Gall & Duquesne (2001).

#### Relation between $S_n$ and the height process

LEMMA 1. For any  $l \in \{0, \dots, n-1\}$ ,

$$\begin{aligned} h_n(l) &= \sum_{j=1}^l I_{\left\{ \min_{0 \leq k \leq l-j} \{S_n(j+k) - S_n(j-1)\} \geq 0 \right\}} \\ (3) \quad &= \#\{j \mid 0 \leq j \leq l-1, \min_{0 \leq k \leq l-j} \{S_n(j+k)\} = S_n(j)\}. \end{aligned}$$

$h_n(l)$  is then the number of “right” minima (in the weak sense) of  $(S_n(i))_{i=0, \dots, l}$  on the interval  $\llbracket 0, l-1 \rrbracket$ .

*Proof:*  $h_n(l)$ , by definition, is the depth of the node  $v_l$  and then it is the number of ancestors of  $v_l$ . Our proof is based on a queue interpretation. We consider a queue with one server where the service rule is: last in, first out; the server serves one client per unit time. Now, consider a tree  $t \in \Omega_n$  with nodes  $(v_0, \dots, v_{n-1})$  (sorted according to their first visit time during the depth first procedure). The nodes are the clients. The root,  $v_0$ , arrives at time 0. The sons of  $v_i$  arrive at time  $i+1$  from the right-most one, to the left-most one (the left-most one arrives the last, so it will be served the first). At time  $i$ , the client which is at the head of the queue gets its service and disappears from the queue at time  $i+1$ .

The queue size in function of time is  $(S_n(i) + 1)_{i=0, \dots, n}$ , where  $S_n$  is the DFQP; indeed, when the client  $i$  is served ( $-1$ ), it is replaced by its offspring ( $+\xi_i$ ), and the queue size at time 0 is 1.

As an illustration, consider on Figure 2 the node  $v_5$ . It is served at time 5 (which is natural, by construction). In the tree, its height is three since he has three

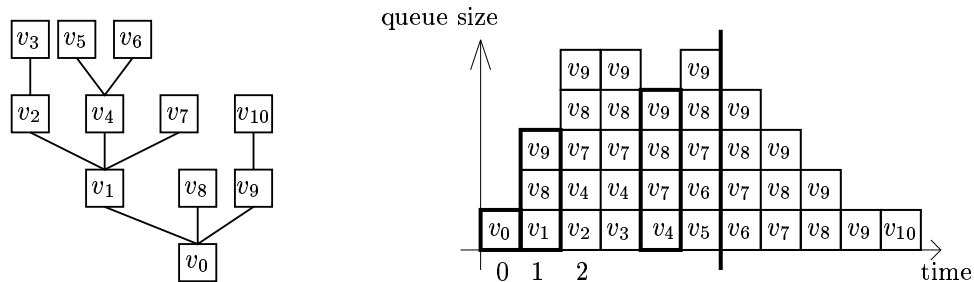


FIGURE 2. The LIFO queue associated to a tree

ancestors  $v_4$ ,  $v_1$  and  $v_0$ . The times when its ancestors are served are the times of right minima before time  $v_5$ .

Let us come back to the proof. First, note that by the queue procedure, the parents are served before the children. Consider  $v_k$ . By construction,  $v_k$  is served at time  $k$ . Note  $t(v_k)$  the subtree of  $t$  with root  $v_k$ . After  $v_k$ , the next  $|t(v_k)| - 1$  clients that are served are exactly the nodes of  $t(v_k)$ ; but these nodes were not in the queue at time  $k$ . So,  $S_n(k) \leq S_n(k+i)$ , for all  $i \in \{1, \dots, |t(v_k)| - 1\}$ . Moreover,  $S_n(k) > S_n(k + |t(v_k)|)$ . Indeed, at time  $k + |t(v_k)|$ , all the descendants of  $v_k$  have left the queue. The queue contains the same elements that were in the queue at time  $k$  except  $v_k$  that has disappeared. It follows, that  $S_n(k)$  is a right minimum in the set  $\llbracket 0, k + |t(v_k)| - 1 \rrbracket$  but not in  $\llbracket 0, k + |t(v_k)| \rrbracket$ .

Consider now  $v_j$  a node of  $t$ . All the ancestors of  $v_j$  are served before  $v_j$  and are right minima in  $\llbracket 0, l - 1 \rrbracket$ . Moreover, a node  $v_i$  that is served before time  $l$  and that is not an ancestor of  $v_l$  is not a right minima in  $\llbracket 0, l - 1 \rrbracket$  since  $S_n(i + |t(v_i)|) < S_n(i)$  and  $i + |t(v_i)| < l$  (because in this case  $v_l$  is not in  $t(v_i)$  and all the nodes of  $t(v_i)$  are served before  $v_l$ ; thus  $l > i + |t(v_i)|$ ). Thus, the number of right minima of  $S_n$  in  $\llbracket 0, l - 1 \rrbracket$  is exactly the number of ancestors of  $v_l$ .  $\square$

*Relation between  $V_n$  and the height process  $h_n$*  Consider again  $(v_0, \dots, v_{n-1})$  the set of nodes sorted as said above. For any  $l \in \{0, \dots, n - 1\}$  we note

$$m(l) = \inf\{k | \tilde{f}(k) = v_l\},$$

where  $\tilde{f}$  is defined at the beginning of the paper. Since  $h_n(l)$  is the height of  $v_l$ , for any  $l \in \{0, \dots, n - 1\}$ , the point  $(m(l), h_n(l))$  belongs to the graph of the DFW  $\{(k, V_n(k)), k \in \{0, \dots, 2n - 2\}\}$ .

LEMMA 2. For any  $l \in \{0, \dots, n - 1\}$ ,

$$m(l) + h_n(l) = 2l.$$

*Proof:* Consider  $h_n(l)$  and  $h_n(l+1)$  the successive heights of the nodes  $v_l$  and  $v_{l+1}$ . Two cases arise:

- $h_n(l+1) = h_n(l) + 1$ . In this case,  $v_{l+1}$  is necessarily the left most son of  $v_l$ . And then  $m(l+1) = m(l) + 1$ . Thus  $h_n(l+1) + m(l+1) = 2 + h_n(l) + m(l)$ .
- $h_n(l+1) \leq h_n(l)$ . In this case, the function  $\tilde{f}$  visits the ancestors of the node  $v_l$



till it finds one node  $v$  with at least one non visited son;  $v_{l+1}$  is the left most one of these non visited sons. We have

$$\begin{aligned} h_n(l+1) &= d(\text{root}, v) + 1 \\ m(l+1) &= m(l) + h_n(l) - d(\text{root}, v) + 1. \end{aligned}$$

Thus  $h_n(l+1) + m(l+1) = m(l) + h_n(l) + 2$ .  $\square$

LEMMA 3. For  $l \in \{0, \dots, n-1\}$  and for any  $k \in \llbracket m(l), m(l+1) \rrbracket$ :

$$\begin{aligned} V_n(m(l+1)) - 1 &\leq V_n(k) \leq V_n(m(l)) \\ h_n(l+1) - 1 &\leq V_n(k) \leq h_n(l). \end{aligned}$$

*Proof:* The DFW interpolates in the following sense the points  $(m(l), h_n(l))_l$ :

$$V_n(k) = \begin{cases} h_n(l) & \text{if } k = m(l) \text{ for a given } l \\ h_n(l) - (k - m(l)) & \text{if } k \in \llbracket m(l) + 1, m(l+1) - 1 \rrbracket \text{ for a given } l. \end{cases}$$

The end of the proof relies on the properties of the depth first search and follows the same arguments as those used in Lemma 2.  $\square$

3.2. *Study via non-conditioned walks, lattice considerations* Consider a random walk  $\omega = (Z_i(\omega))_{i \geq 0}$  with increments  $(z_i)_{i \geq 0}$ , i.i.d. such that  $Z_0 = 0$ ,  $\mathbb{E}(z_1) = 0$  and  $0 < \text{Var}(z_1) < +\infty$ . Assume, moreover that  $z_i$  takes its values on the set  $\{-1, 0, 1, 2, \dots\}$ .

We denote by  $RW_Z(n)$  the set of  $n$  length random walks  $(Z_i)_{0 \leq i \leq n}$  with the law induced by the law of the  $z_i$  and  $EX_Z(n)$  the set of corresponding ‘‘excursions’’:

$$\{\omega \in EX_Z(n)\} \Leftrightarrow \{\omega \in RW_Z(n), Z_n(\omega) = -1, Z_i(\omega) \geq 0, \forall i \in \{0, \dots, n-1\}\}.$$

The terminology ‘‘excursion’’ for this kind of path is justified by the fact that adding 1 to each  $Z_i$  (for  $i = 1, \dots, n$ ) provides an usual discrete excursion.

**Existence and probability of size  $n$  excursions (lattice considerations)**

Let  $T_{-1}$  denote the hitting time of  $-1$  by  $(Z_i)_{i \geq 0}$ :

$$T_{-1} = \inf\{i \geq 0, Z_i = -1\}.$$

We have (Otter’s (1949) formula):

$$\mathbb{P}(\omega \in EX_Z(n)) = \mathbb{P}(T_{-1}(\omega) = n) = \frac{1}{n} \mathbb{P}(Z_n = -1).$$

Since  $z_i$  may have a lattice distribution, the above probability may be zero for some  $n$ . Assume that  $z_i$  takes its values on the lattice  $-1 + kh$ , where  $h$  is the maximal span. Let  $(a(n))_n$  be a sequence of indices such that  $a(n) \rightarrow +\infty$  and  $\mathbb{P}(Z_{a(n)} = -1) \neq 0$ . The local central limit theorem (see Port (1994) p.706) says

$$\frac{\sqrt{a(n)}}{h} \mathbb{P}(Z_{a(n)} = -1) - \frac{1}{\sqrt{2\pi\sigma}} \exp\left(-\frac{(-1)^2}{2\text{Var}(z_1)a(n)}\right) \rightarrow 0, \text{ as } n \rightarrow +\infty,$$

and then

$$\mathbb{P}(Z_{a(n)} = -1) \sim \frac{h}{\sqrt{2\pi a(n)}\sigma}.$$

Hence, for such indices  $a(n)$ , we have:

$$\mathbb{P}(EX_Z(a(n))) \sim \frac{h}{\sqrt{2\pi\sigma a(n)}^{3/2}}.$$

When  $\xi$  is lattice distributed, the total progeny in the GW process with offspring  $\xi$  is also lattice distributed and then the length of the DFW and the length of the DFQP are also lattice distributed. The appearance of the set of indices  $a(n)$  is necessary as well as the one of  $PP(\xi)$ .

**From non-conditioned walk to excursion (the conditioning argument)**

Let  $A_n$  be a subset of  $RW_Z(n)$  and  $n$  such that  $\mathbb{P}(Z_n = -1) \neq 0$ . Then

$$\mathbb{P}(A_n | EX_Z(n)) \leq \frac{\mathbb{P}(A_n)}{\mathbb{P}(EX_Z(n))}.$$

Thus,

$$(4) \quad \mathbb{P}(A_n) = o(n^{-\beta}) \implies \mathbb{P}(A_n | EX_Z(n)) = o(n^{-\beta+3/2}).$$

This implies that: if there exists a moderate deviations principle (or large deviations principle) for a functional of a centered random walk, there exists an upper bound for the analogous principle on the associated excursion. This remark is one of the key points of the present paper and also of Marckert (2000). Its interest is that almost all random variables are much more difficult to handle on  $EX_Z(n)$  than on  $RW_Z(n)$ . But, exponential concentration type inequalities on  $RW_Z(n)$  remain true on  $EX_Z(n)$ .

**3.3. The number of right minima** The aim of this section is to show a “strong relation” between the number of right minima of a random walk  $(W(i))_{0 \leq i \leq l}$  in the set  $\llbracket 0, l-1 \rrbracket$  and the value  $W(l) - \min_{0 \leq i \leq l} \{W(i)\}$ . Thanks to the conditioning argument (Subsection 3.2), this strong relation will be transported to the excursion in Section 3.4.

Set  $l$  a fixed integer and  $(W(j))_{j \in \{0, \dots, l\}}$  a random walk with i.i.d. increments with law  $(\tilde{p}_i)_{i \geq -1}$ . The number of right minima of  $(W(i))_i$  (say  $RM(l)$ ) on  $\llbracket 0, l \rrbracket$  is:

$$RM(l) = \#\{j | 0 \leq j \leq l-1, \min_{0 \leq k \leq l-j} \{W(j+k)\} = W(j)\}.$$

**3.3.1. The right minima viewed as ladder epochs on an associated random walk** In order to study the random variable  $RM(l)$ , we introduce the associated random walk  $(W^\bullet(j))_{j \in \{0, \dots, l\}}$  defined by:

$$(5) \quad W^\bullet(j) = W(l) - W(l-j) \text{ for any } j \in \{0, \dots, l\}.$$

“Geometrically”, the graph of the random walk  $W^\bullet$  is the reflexion of the graph of  $W$  according to the center  $(l/2, W(l)/2)$ . The increments of  $(W^\bullet(i))_{i \geq 0}$  are also

$(\tilde{p}_i)_{i \geq -1}$  distributed. Denote by  $R^\bullet(l)$  the number of records (in the large sense) of the sample path  $(W^\bullet(j))_{j=0, \dots, l}$ :

$$R^\bullet(l) = \#\{j, 1 \leq j \leq l, \max_{0 \leq k \leq j} \{W^\bullet(k)\} = W^\bullet(j)\}.$$

By (5), we obtain:

$$(6) \quad RM(l) = R^\bullet(l),$$

$$(7) \quad \max_{0 \leq i \leq l} W^\bullet(i) = W(l) - \min_{0 \leq i \leq l} W(i).$$

$$(8) \quad \left| \max_{0 \leq j \leq l} \{W^\bullet(j)\} - R^\bullet(l) \frac{\sigma^2}{2} \right| = \left| W(l) - \min_{0 \leq i \leq l} \{W(i)\} - RM(l) \frac{\sigma^2}{2} \right|.$$

Even if  $R^\bullet(l)$  and  $RM(l)$  are very similar functionals of the paths,  $R^\bullet(l)$  appears to be easier to handle. Indeed, the record times are stopping times and then, Markov properties of the paths can be used to study  $R^\bullet(l)$ . This is the reason for the construction of the path  $W^\bullet$ .

3.3.2. *Concentration inequality on  $\max_{0 \leq j \leq l} \{W^\bullet(j)\}$ .* Our aim here is to show a concentration inequality for the difference between  $\frac{\sigma^2}{2} R^\bullet(l)$  and  $\max_{0 \leq j \leq l} \{W^\bullet(j)\}$ :

PROPOSITION 2. *For any  $\nu > 0$  there exist two constants  $\gamma > 0$  and  $N > 0$  such that for any  $n \geq N$  and any  $l, 0 \leq l \leq n$ ,*

$$(9) \quad \mathbb{P} \left( \left| \max_{0 \leq j \leq l} \{W^\bullet(j)\} - \frac{\sigma^2}{2} R^\bullet(l) \right| \geq n^{1/4+\nu} \right) \leq e^{-\gamma n^\nu}.$$

We can assume that  $W^\bullet$  is an infinite length r.w. with increment law  $(\tilde{p}_i)_{i \geq -1}$ . Consider  $0 = \tau(0) < \tau(1) < \dots < \tau(R^\bullet(l)) < \dots$  the weak ladder epochs for  $W^\bullet$ .

$$\tau_k = \inf \{j > \tau_{k-1}, W^\bullet(j) \geq W^\bullet(\tau_{k-1})\}, \text{ for } k \geq 1.$$

$R^\bullet(l)$  is the number of positive ladder epochs before time  $l$ . The weak ladder heights are the variables:

$$(10) \quad LH_k = W^\bullet(\tau_k) - W^\bullet(\tau_{k-1}), \text{ for } k \geq 1.$$

We have

$$(11) \quad \max_{0 \leq j \leq l} \{W^\bullet(j)\} = \sum_{k=1}^{R^\bullet(l)} LH(k).$$

To prove Proposition 2, we need the lemma:

LEMMA 4.

$$\mathbb{E}(LH_1) = \frac{\sigma^2}{2}.$$

*Proof* : For this proof or more general questions about ladder heights, we refer to Feller (1971), chap. XII and p. 425-426. The fundamental relation is:

(12)

$$\text{for any } r \geq 0, \quad \mathbb{P}(LH_1 = r) = \mathbb{P}(W^\bullet(1) = r) + \frac{\mathbb{P}(W^\bullet(1) = -1)}{1 - \mathbb{P}(LH_1 = 0)} \mathbb{P}(LH_1 = r + 1).$$

Now, write

$$\sum_{r \geq 0} r \mathbb{P}(LH_1 = r) = \sum_{r \geq 0} r \left( \mathbb{P}(W^\bullet(1) = r) + \frac{\mathbb{P}(W^\bullet(1) = -1)}{1 - \mathbb{P}(LH_1 = 0)} \mathbb{P}(LH_1 = r + 1) \right).$$

We obtain,

$$\mathbb{E}(LH_1) \left( 1 - \frac{\mathbb{P}(W^\bullet(1) = -1)}{1 - \mathbb{P}(LH_1 = 0)} \right) = 0.$$

Using that  $\mathbb{E}(LH_1) > 0$ , we obtain  $\mathbb{P}(LH_1 = 0) = 1 - \mathbb{P}(W^\bullet(1) = -1)$  and by (12):

$$(13) \quad \mathbb{P}(LH_1 = r) = \mathbb{P}(W^\bullet(1) = r) + \mathbb{P}(LH_1 = r + 1) = \mathbb{P}(W^\bullet(1) \geq r).$$

The end of the proof is immediate.  $\square$

**Remark 1:** According to (13),  $LH_1$  has a moment of order  $p$  (resp. an exponential moment) if and only if  $W^\bullet(1)$  has a moment of order  $p + 1$  (resp. an exponential moment).

*Proof of Proposition 2:* The random walk  $W^\bullet$  enjoys the strong Markov property, and then, the r.v.  $\tau_k - \tau_{k-1}$  (for  $k \geq 1$ ) are i.i.d. as well as the r.v.  $LH_k$ . Let  $K(l, n)$  denote the left hand side of (9). It follows from (11) that

$$K(l, n) = \mathbb{P} \left( \left| \sum_{k=1}^{R^\bullet(l)} \left( LH_k - \frac{\sigma^2}{2} \right) \right| \geq n^{1/4+\nu} \right).$$

Then

$$K(l, n) \leq A(l, n) + B(l, n)$$

where

$$A(l, n) = \mathbb{P}(R^\bullet(l) \geq n^{1/2+\nu})$$

$$B(l, n) = \mathbb{P} \left( \left| \sum_{k=1}^{R^\bullet(l)} \left( LH_k - \frac{\sigma^2}{2} \right) \right| \geq n^{1/4+\nu}, R^\bullet(l) \leq n^{1/2+\nu} \right).$$

$A(l, n)$  and  $B(l, n)$  appears to be probability of moderate deviations of some r.v.; indeed,  $R^\bullet(l)$ , the number of records of a centered random walk is known to be of the order  $\sqrt{l}$  and since  $(\sum_{k=1}^n (LH_k - \frac{\sigma^2}{2})) / \sqrt{n}$  converges to a Gaussian (non degenerate) distribution,  $B(l, n)$  appears also to be the probability of a rare event. To bound the probability of these rare events, we will use Petrov's results (see the Appendices). We define  $(U_k)_{k \geq 0}$  by:

$$U_k = \sum_{j=1}^k \left( LH_j - \frac{\sigma^2}{2} \right).$$

The r.v.  $(LH_j - \frac{\sigma^2}{2})_j$  are i.i.d. centered r.v.. It follows from Remark 1 and assumptions (1) that there exists a constant  $\beta > 0$  such that:

$$\mathbb{E}(e^{\beta|LH_i - \frac{\sigma^2}{2}|}) < +\infty.$$

The two following Lemmas 5 and 6 allow one to conclude the proof of Proposition 2.

LEMMA 5. *There exist two constants  $c_1 > 0$  and  $N_1 > 0$  such that for any  $n \geq N_1$  and any  $l, 0 \leq l \leq n$ ,*

$$A(l, n) \leq \exp(-c_1 n^\nu)$$

*Proof:* In order to simplify the notations, in this proof, the indices  $n^{1/2+\nu}$  have to be read,  $\lfloor n^{1/2+\nu} \rfloor$ .

$$\begin{aligned} A(l, n) &\leq \mathbb{P}(\tau_{n^{1/2+\nu}} \leq l) \\ &\leq \mathbb{P}(\tau_{n^{1/2+\nu}} \leq n) \\ &\leq \mathbb{P}(\max_{0 \leq j \leq \tau_{n^{1/2+\nu}}} W^\bullet(j) \leq \max_{0 \leq j \leq n} W^\bullet(j)) \\ &\leq \mathbb{P}(W^\bullet(\tau_{n^{1/2+\nu}}) \leq n^{1/2+\nu/2}) + \mathbb{P}(\max_{0 \leq j \leq n} W^\bullet(j) > n^{1/2+\nu/2}) \\ &\leq \mathbb{P}(U_{n^{1/2+\nu}} + \lfloor n^{1/2+\nu} \rfloor \frac{\sigma^2}{2} \leq n^{1/2+\frac{\nu}{2}}) + 2\mathbb{P}(W^\bullet(n) \geq n^{1/2+\frac{\nu}{2}} - \sqrt{2\sigma^2(n-1)}) \\ &\leq \mathbb{P}(-U_{n^{1/2+\nu}} \geq \lfloor n^{1/2+\nu} \rfloor \frac{\sigma^2}{2} - n^{1/2+\frac{\nu}{2}}) + 2 \exp\left(-\frac{(n^{1/2+\frac{\nu}{2}} - \sqrt{2\sigma^2(n-1)})^2}{2ng_w}\right) \\ &\leq \exp(-c_2 n^{1/2+\nu}) + 2 \exp(-c_3 n^\nu) \end{aligned}$$

where  $g_w$  is the constant  $g$  from Petrov Lemmas relative to the r.w.  $(W^\bullet(k))_{k \geq 1}$ ,  $c_2$  and  $c_3$  are two constants given in using Petrov's lemmas 8; Lemma 7 has been used to bound the most right term in the fourth inequality.  $\square$

LEMMA 6. *There exist two constants  $\tilde{c}_1 > 0$  and  $N_2 > 0$  such that for any  $n \geq N_2$  and any  $l, 0 \leq l \leq n$ ,*

$$B(l, n) \leq \exp(-\tilde{c}_1 n^\nu).$$

*Proof:*

$$\begin{aligned} B(l, n) &\leq \sum_{j=1}^{n^{1/2+\nu}} \mathbb{P}(|U_j| \geq n^{1/4+\nu}) \\ &\leq \sum_{j=1}^{\lfloor c_4 n^{1/4+\nu} \rfloor} \exp(-T' c_4 n^{1/4+\nu}/2) + \sum_{j=\lfloor c_4 n^{1/4+\nu} \rfloor}^{n^{1/2+\nu}} \exp\left(-\frac{n^{1/2+2\nu}}{2jg'}\right) \\ &\leq \exp(-\tilde{c}_1 n^\nu) \end{aligned}$$

For well chosen constants  $\tilde{c}_1$ ,  $g'$ ,  $T'$ ,  $c_4$  by Lemma 8, since  $LH_k$  has exponential moments.  $\square$

**3.3.3. Concentration inequality on  $W$**  A simple application of Proposition 2 and (8) gives: for any  $\nu > 0$  there exist two constants  $\gamma > 0$  and  $N > 0$  such that, for any  $n$ ,  $n > N$  and any  $l$ ,  $0 \leq l \leq n$

$$(14) \quad \mathbb{P}\left(\left|W(l) - \min\{W(i), 0 \leq i \leq l\} - RM(l)\frac{\sigma^2}{2}\right| \geq n^{1/4+\nu}\right) \leq e^{-\gamma n^\nu}$$

One deduces immediately:

**COROLLARY 5.** *For any  $\nu > 0$  there exist two constants  $\gamma > 0$  and  $N > 0$  such that, for any  $n$ ,  $n > N$*

$$\mathbb{P}\left(\sup_{0 \leq l \leq n} \left\{\left|W(l) - \min\{W(i), 0 \leq i \leq l\} - RM(l)\frac{\sigma^2}{2}\right| \geq n^{1/4+\nu}\right\}\right) \leq e^{-\gamma n^\nu}$$

### 3.4. The depth processes

**3.4.1. Proof of Theorem 3** Consider the excursion  $(S_n(i))_{0 \leq i \leq n}$ . We have, for any  $l$  smaller than  $n$ :

$$\min_{0 \leq i \leq l} \{S_n(i)\} = 0.$$

Recall that  $S_n$  is  $(W_i)_{i=0, \dots, n}$  conditioned to be an excursion. We have

$$\mathbb{P}\left(\sup_{0 \leq l \leq n} \left\{\left|S_n(l) - h_n(l)\frac{\sigma^2}{2}\right| \geq n^{1/4+\nu}\right\}\right) = \mathbb{P}\left(\sup_{0 \leq l \leq n} \left\{\left|W(l) - \min_{0 \leq i \leq l} \{W(i)\} - RM(l)\frac{\sigma^2}{2}\right| \geq n^{1/4+\nu} \mid W \in EX_W\right\}\right).$$

It follows from Corollary 5 and (4) that there exists a constant  $C > 0$  such that

$$\mathbb{P}\left(\sup_{0 \leq l \leq n} \left\{\left|S_n(l) - h_n(l)\frac{\sigma^2}{2}\right| \geq n^{1/4+\nu}\right\}\right) = O(n^{5/2}) \exp(-Cn^\nu). \quad \square$$

**3.4.2. Proof of Theorem 2:** For any  $t \in [0, 1[$ , we denote by  $\zeta(nt)$  the integer such that

$$(15) \quad [2nt] \in [m(\zeta(nt)), m(\zeta(nt) + 1)[.$$

$$\mathbb{P}\left(\sup_{t \in [0, 1]} \left|S_n(nt) - \frac{\sigma^2}{2} V_n(2nt)\right| \geq n^{1/4+\nu}\right) \leq A + B + C + D$$

where

$$(16) \quad A = \mathbb{P}\left(\sup_{t \in [0,1]} |S_n(nt) - S_n(\zeta(nt))| \geq \frac{n^{1/4+\nu}}{3}, \sup_t |\zeta(nt) - nt| < n^{1/2+\nu}\right)$$

$$B = \mathbb{P}\left(\sup_{t \in [0,1]} \left|S_n(\zeta(nt)) - \frac{\sigma^2}{2} h_n(\zeta(nt))\right| \geq \frac{n^{1/4+\nu}}{3}\right)$$

$$C = \mathbb{P}\left(\frac{\sigma^2}{2} \sup_{t \in [0,1]} |h_n(\zeta(nt)) - V_n(2nt)| \geq \frac{n^{1/4+\nu}}{3}\right)$$

$$(17) \quad D = \mathbb{P}\left(\sup_{t \in [0,1]} |\zeta(nt) - nt| > n^{1/2+\nu}\right)$$

$$\begin{aligned} A &= O(n^{3/2}) \mathbb{P}\left(\sup_{k \leq n^{1/2+\nu}} \sup_{0 \leq j \leq n-k} |W(j) - W(j+k)| \geq \frac{n^{1/4+\nu}}{3}\right) \\ &\leq O(n^{3/2}) \sum_{j=1}^n \mathbb{P}\left(\sup_{k \leq n^{1/2+\nu}} |W(k)| \geq \frac{n^{1/4+\nu}}{3}\right) \\ &\leq e^{-c_5 n^\nu}, \end{aligned}$$

for a certain constant  $c_5 > 0$ , if  $n$  is large enough using Petrov's Lemmas. According to Theorem 3, there exists a constant  $c_6 > 0$ , such that :

$$(18) \quad B = O(n^{5/2} \exp(-c_6 n^\nu)).$$

Since  $[2nt] \in [m(\zeta(nt)), m(\zeta(nt) + 1)[$ , we obtain by Lemma 3:

$$C \leq \mathbb{P}\left(\frac{\sigma^2}{2} \sup_{l \in \{1, \dots, n\}} \{|h_n(l) - h_n(l+1) + 1|\} \geq \frac{n^{1/4+\nu}}{3}\right).$$

$$C \leq \mathbb{P}\left(\frac{\sigma^2}{2} \sup_{l \in \{1, \dots, n\}} \{|h_n(l+1) - 1 - \frac{2}{\sigma^2} S_n(l)|\} \geq \frac{n^{1/4+\nu}}{6}\right)$$

$$+ \mathbb{P}\left(\frac{\sigma^2}{2} \sup_{l \in \{1, \dots, n\}} \{|h_n(l) - \frac{2}{\sigma^2} S_n(l)|\} \geq \frac{n^{1/4+\nu}}{6}\right)$$

we conclude using Theorem 3.

To bound  $D$ , we use (15) and Lemma 2; we have:

$$\begin{aligned} \sup_{t \in [0,1]} |2\zeta(nt) - 2nt| &\leq \sup_{t \in [0,1]} |2\zeta(nt) - m(\zeta(nt))| + \sup_{t \in [0,1]} |m(\zeta(nt)) - 2nt| \\ &\leq 2 \sup_{0 \leq l \leq n} \{h_n(l)\}. \end{aligned}$$

since  $h_n(l)$  is the number of right minima of  $S_n$  (see Lemma 3), we use (6) and the conditioning argument to show that

$$\mathbb{P}\left(\sup_{0 \leq l \leq n} \{h_n(l)\} \geq n^{1/2+\nu}\right) = O(n^{3/2}) \mathbb{P}\left(\sup_{0 \leq l \leq n} \{R^\bullet(l)\} \geq n^{1/2+\nu}\right).$$

We conclude using Lemma 5.  $\square$

3.5. *The heights of nodes with out-degree  $j$*  *Proof of Theorem 4:* Consider again the sequence  $(v_0, \dots, v_{n-1})$  of ordered nodes. Let  $\mu(i)$  be the index of the  $i$ th node with out-degree  $j$ , then:

$$h_n^{(j)}(i) = h_n(\mu(i)).$$

$\mu(i)$  is the position of the  $i$ th increment  $j - 1$  in the DFQP,  $S_n$ . Note  $\mu^*(i)$  the position of the  $i$ th increment  $j - 1$  in  $W$  (which is  $S_n$  with the excursion condition removed);  $\mu^*(i)$  can be written:

$$(19) \quad \mu^*(i) = \sum_{k=1}^i \mathcal{G}(k)$$

where the  $(\mathcal{G}(k))_{k \geq 0}$  are i.i.d. r.v. geometrically distributed with parameter  $p_j$ . Hence, concentration inequality can again be obtained using Petrov's Lemmas: for any  $\nu > 0$ , there exists a constant  $C > 0$ , such that,

$$(20) \quad \mathbb{P}\left(\sup_{0 \leq i \leq n} \left| \mu^*(i) - \frac{i}{p_j} \right| \geq n^{1/2+\nu}\right) = O(\exp(-Cn^\nu)).$$

Using the conditioning argument (see Subsection 3.2), we obtain that:

$$(21) \quad \mathbb{P}\left(\sup_{0 \leq i \leq n} \left| \mu(i) - \frac{i}{p_j} \right| \geq n^{1/2+\nu}\right) = O(n^{3/2} \exp(-Cn^\nu)).$$

We conclude using that

$$(22) \quad \sup_k \left| h_n^{(j)}(k) - h_n\left(\frac{k}{p_j}\right) \right| \leq \sup_k \left| h_n(\mu(k)) - \frac{2}{\sigma^2} S_n(\mu(k)) \right|$$

$$(23) \quad + \frac{2}{\sigma^2} \sup_k \left| S_n(\mu(k)) - S_n\left(\frac{k}{p_j}\right) \right| + \sup_k \left| \frac{2}{\sigma^2} S_n\left(\frac{k}{p_j}\right) - h_n\left(\frac{k}{p_j}\right) \right|.$$

The two extreme terms in the right hand side satisfy concentration exponential inequality by Theorem 3; we carry out exponential bound for the last term using the same computation as that used for  $A$  and  $D$  in (16) and (17).  $\square$

**Remark 2:** An other object of interest is the number of nodes with out-degree  $j$ :

$$N_j^{(n)} = \# \{ i \mid \xi_i = j \}.$$

The exact law of  $N_j^{(n)}$  seems to be difficult to compute because the tree is conditioned to be of size  $n$ . But, with our method we can obtain moderate deviations. Indeed, if we remove the tree size condition, the sequence  $(\xi_i)_{0 \leq i \leq n-1}$  is a sequence of i.i.d. r.v.,  $N_j^{(n)}$  is binomial  $B(n, p_j)$  distributed and then  $\bar{N}_j^{(n)}$  is concentrated around  $np_j$ . Now, using the conditioning argument given in subsection 3.2, one finds that for any  $\nu > 0$ , there exists a constant  $C$  such that,

$$\mathbb{P}(|N_j^{(n)} - p_j n| \geq n^{1/2+\nu}) = O(\exp(-Cn^\nu)).$$

Other elements about the law of  $N_j^{(n)}$  and about the joint law of the  $(N_j^{(n)})_j$  are given in Drmota (1996).



**3.6. Occurrence of a given subtree** We denote by  $N_u(n)$  the number of nodes in  $\tau \in \Omega_n$  roots of a subtree equal to  $u$ . For  $1 \leq i \leq N_u(n)$ , let  $\phi(i)$  denote the index of the  $i$ th node among  $(v_0, v_1, \dots, v_{n-1})$ , that is the root of a  $u$ -type subtree and

$$\rho(i) = \phi(i) + \lambda - 1.$$

Each (finite) tree  $t$  from  $\Omega$  can be described by its sequence of successive progenies (SSP),  $Seq = (\xi_0, \dots, \xi_{n-1})$ , where  $n$  is the number of nodes of  $t$ . We denote by  $Seq(i, j)$  the sequence  $(\xi_i, \dots, \xi_{i+j-1})$  and by  $Seq(i)$  the  $i+1$ th coordinate of  $Seq$ , in order that  $Seq(i) = \xi_i$ . Let  $Word = (a_0, \dots, a_{\lambda-1})$  denote the SSP of the subtree  $u$ . In a first step, we establish a concentration theorem for occurrences of  $Word$  in a sequence  $Seq^*$ , when the  $(Seq^*(i))_i$  are i.i.d. r.v. with the same law as  $\xi$ . Then, by the connexion given in subsection 3.2, we transport the non conditioned concentration type inequality obtained on  $Seq^*$  to a concentration type inequality on  $Seq$ , that is on  $\Omega_n$ .

So, consider  $(Seq^*(i))_i$  and set  $N_u^*(n)$  the number of occurrences of  $Word$  in  $Seq^*$ . For  $1 \leq i \leq N_u^*(n)$ , we define the sequence  $(\phi^*(i))_i$ , by

$$\phi^*(1) = \inf\{k, Seq^*(k, k + \lambda - 1) = Word\}$$

$$\phi^*(i) = \inf\{k, k > \phi^*(i - 1), Seq^*(k, k + \lambda - 1) = Word\}.$$

An important point is that, by the properties of the subtree (and its induced DFW), two sequences  $Word$  in  $Seq^*$  can not overlap. To express more simply the Markov properties of  $(\phi^*(i))_{0 \leq i}$ , for  $1 \leq i \leq N_u^*(n)$ , we note

$$\rho^*(i) = \phi^*(i) + \lambda - 1.$$

$\rho^*(i)$  is the index of the last letter of the  $i$ th occurrence of  $Word$  in  $Seq^*$ . We have:

$$\rho^*(1) + 1 \stackrel{d}{=} \rho^*(i + 1) - \rho^*(i) \text{ for all } i \geq 1,$$

(we set  $\rho^*(0) = -1$ ).

**Mean and tail of the distribution of  $\rho^*(1)$**

$$\begin{aligned} \mathbb{P}(\rho^*(1) = \lambda + k - 1) &= \mathbb{P}(\rho^*(1) > k - 1 \cap (Seq^*(k, \lambda) = Word)) \\ (24) \qquad \qquad \qquad &= \mathbb{P}(\rho^*(1) > k - 1) \mathbb{P}(Seq^*(k, \lambda) = Word) \\ &= \mathbb{P}(\rho^*(1) > k - 1) \mathbb{P}(Seq^*(0, \lambda - 1) = Word) \end{aligned}$$

The first equation comes from the fact that two sequences  $Word$  can not overlap. The second is due to the fact that the  $(Seq^*(i))_i$  are independent.

We have:

$$(25) \quad \mathbb{P}(Seq^*(0, \lambda - 1) = Word) \stackrel{def}{=} q_u = \prod_i p_i^{n_i} \text{ where } n_i = \#\{j | a_j = i\}.$$

Since  $\mathbb{P}(\rho^*(1) < \lambda - 1) = 0$ , the distribution of  $\rho^*(1)$  is well defined by the above equations.

We sum equality (24) from  $k$  equal 0 to  $+\infty$ ; we obtain:

$$1 = q_u(\mathbb{E}(\rho^*(1)) + 1)$$

and then:

$$(26) \quad \mathbb{E}(\rho^*(1)) + 1 = \mathbb{E}(\rho^*(i+1) - \rho^*(i)) = q_u^{-1}.$$

Let  $k = m\lambda + r$ , with  $0 \leq r < \lambda$

$$\{\rho^*(1) > k\} \subset \bigcap_{0 \leq s \leq m-1} \{Seq(s\lambda, \lambda) \neq Word\}.$$

Since, the events in the right hand side are independent, and have the same probability:

$$\begin{aligned} \mathbb{P}(\rho^*(1) > k) &\leq (1 - q_u)^m \\ &\leq \beta \delta^k \end{aligned}$$

where  $\delta = (1 - q_u)^{1/\lambda}$ , and  $\beta = (1 - q_u)^{-1}$ . Thus  $\rho^*(1)$  has exponential moment.

Now, one can copy the arguments given in Subsection 3.5. Indeed, we have

$$\rho^*(j) = 1 + \sum_{i=1}^j (\rho^*(i) - \rho^*(i-1))$$

to be compared with (19). We have

$$\mathbb{P}\left(\sup_{0 \leq i \leq N_u^*(n)} \left| \rho^*(i) - \frac{i}{q_u} \right| \geq n^{1/2+\nu}\right) \leq \mathbb{P}\left(\sup_{0 \leq i \leq n} \left| r^*(i) - \frac{i}{q_u} \right| \geq n^{1/2+\nu}\right)$$

where  $(r^*(i))_{i \geq 0}$  is a random walk with increment distributed as  $\rho^*(1)$ . Using the existence of an exponential moment for  $\rho^*(1)$  and Petrov's Lemmas, we obtain that for any  $\nu > 0$ , there exists a constant  $C > 0$  such that:

$$\mathbb{P}\left(\sup_{0 \leq i \leq N_u^*(n)} \left| \rho^*(i) - \frac{i}{q_u} \right| \geq n^{1/2+\nu}\right) = O(\exp(-Cn^\nu))$$

and then by the conditioning arguments,

$$(27) \quad \mathbb{P}\left(\sup_{0 \leq i \leq N_u(n)} \left| \rho(i) - \frac{i}{q_u} \right| \geq n^{1/2+\nu}\right) = O(n^{3/2} \exp(-Cn^\nu))$$

to be compared with (20). The end of the proof is similar to the end of the proof of Proposition 2.  $\square$

*Proof of Corollary 4:* We have:

$$N_u(n) = \max\{i | \rho(i) \leq n\} \text{ and } N_u^*(n) = \max\{i | \rho^*(i) \leq n\}.$$

Then

$$\begin{aligned} \mathbb{P}(|N_u(n) - q_u n| \geq n^{1/2+\nu}) &= \mathbb{P}(|N_u(n)/q_u - n| \geq n^{1/2+\nu}/q_u) \\ &\leq \mathbb{P}(|N_u(n)/q_u - \rho(N_u(n))| \geq n^{1/2+\nu}/(2q_u)) \\ &\quad + \mathbb{P}(|\rho(N_u(n)) - n| \geq n^{1/2+\nu}/(2q_u)) \end{aligned}$$

The first term in the right side is exponentially small thanks to (27). For the second term, we have

$$\mathbb{P}(|\rho(N_u(n)) - n| \geq n^{1/2+\nu}/(2q_u)) = O(n^{3/2}) \mathbb{P}(|\rho^*(N_u^*(n)) - n| \geq n^{1/2+\nu}/(2q_u))$$

Since

$$|\rho^*(N_u^*(n)) - n| \leq \rho^*(N_u^*(n) + 1) - \rho^*(N_u^*(n))$$

and by the Markov property  $\rho^*(N_u^*(n) + 1) - \rho^*(N_u^*(n)) \stackrel{d}{=} \rho^*(1) + 1$ , we have:

$$\begin{aligned} \mathbb{P}(|\rho(N_u(n)) - n| \geq n^{1/2+\nu}/(2q_u)) &\leq O(n^{3/2})\mathbb{P}(\rho_1^* \geq n^{1/2+\nu}/(2q_u) - 1) \\ &\leq \beta \delta^{n^{1/2+\nu}/(2q_u)}. \end{aligned}$$

This concludes the proof of the corollary.  $\square$

*Appendices: Petrov's results about moderate deviations of random walk* The two following lemmas are direct corollaries of Theorems 2.2, 2.4, and 2.6 from Petrov (1975).

LEMMA 7. *Let  $(Z_i)_{i \geq 0}$  a random walk with increment  $z_i$ , centered, with  $\mathbb{E}(z_i^2) < +\infty$ . Then, for every  $x$ :*

$$\mathbb{P}\left(\max_{0 \leq k \leq n} \{Z_k\} \geq x\right) \leq 2 \mathbb{P}(Z_n \geq x - \sqrt{2(n-1)\mathbb{E}(z_i^2)})$$

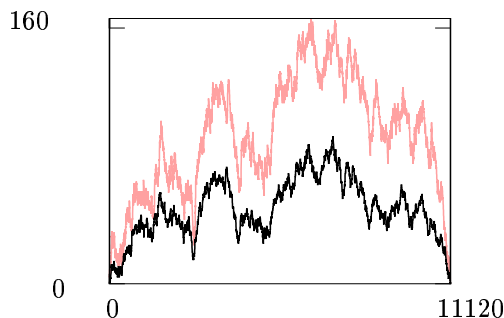
LEMMA 8. *If moreover, there exists a constant  $a > 0$  such that  $\mathbb{E}(e^{a|z_1|}) < +\infty$  then there exist a constant  $g > 0$  and a constant  $T > 0$  such that*

$$\mathbb{P}(Z_n \geq x) \leq \exp\left(-\frac{x^2}{2ng}\right) \quad \text{if } 0 \leq x \leq ngT$$

$$\mathbb{P}(Z_n \geq x) \leq \exp\left(-\frac{Tx}{2}\right) \quad \text{if } x \geq ngT.$$

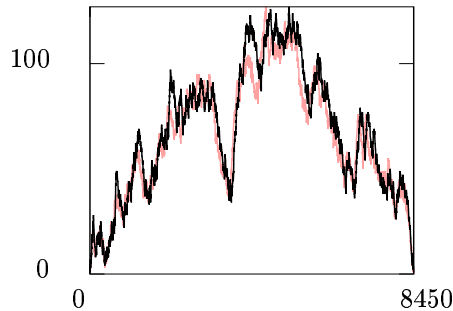
**4. Simulations** Theorem 2 possesses very visual interpretations, as can be seen on the following simulations.

1) On the following figure, we have simulated a tree with size 5560 and with lattice offspring distribution  $p_0 = 13/18, p_2 = 1/6, p_6 = 1/9$ ; the corresponding value of  $\sigma^2/2$  is  $11/6$ . In black, the DFW, in grey the DFQP. One can observe the quasi



proportionality of the two processes. Note also that since  $2l \geq m(l)$  (for any  $l$ ), the DFW “lags” from the DFQP.

2) On the following figure, we have simulated a size 4208 tree with a non lattice offspring distribution  $p_0 = 8/15, p_1 = 4/15, p_3 = 2/15, p_5 = 1/15$ , (in this case  $\sigma^2/2 = 16/15$ )



**5. Conclusion** We have obtained exponential bounds for the probability of small deviations for the uniform distance between the normalized depth processes, that is, for example:

$$(28) \quad \mathbb{P}\left(\sup_{0 \leq t \leq 1} \left| \frac{S_n(2nt)}{\sqrt{n}} - \frac{\sigma^2 h_n(nt)}{2\sqrt{n}} \right| \geq n^{-1/4+\nu}\right) \leq e^{-\gamma n^\nu}.$$

A corollary of these results is that the five depth processes have the same Brownian excursion as limit.

If one only wants to obtain the weak convergence of the five processes to the same limit, one may weaken the assumption on  $\xi$  and only assumes the existence of moment of order  $p$  (instead of the existence of an exponential moment). In this case, Petrov's inequality, may be replaced by Fuk & Nagaev (1971) inequality (see also Petrov (1975) p.78):

$$\mathbb{P}(W_n \geq x) \leq \frac{c_1 n}{x^p} + \exp(-c_2 \frac{x^2}{n}),$$

for two constants  $c_1 > 0$  and  $c_2 > 0$ .

Taking  $p \geq 9$  and using the same proofs as the ones given in the present paper, one shows that the uniform distances between the normalized depth processes converge to 0 in probability.

One can probably obtain this convergence for smaller values of  $p$  with much more technical computations, but this is not in the spirit of the present paper.

Whatever, one can note, that the ladder height has a finite variance if and only if  $p \geq 3$ , and so, we can conjecture that in this case the uniform distance between  $S_n$  and  $V_n$  converges to zero; some simulations "show" that this convergence to zero is not "obvious" if  $p < 3$ .

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