# Coexistence of lazy frogs on $\mathbb{Z}$

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#### Abstract

We study the so-called frog model on  $\mathbb{Z}$  with two types of lazy frogs, with parameters  $p_1, p_2 \in (0, 1]$  respectively, and finite expected number of dormant frogs per site. We show that for any such  $p_1$  and  $p_2$  there is positive probability that the two types coexist (i.e. that both types activate infinitely many frogs). This answers a question of Deijfen, Hirscher, and Lopes in dimension one.

**Keywords:** Frog model, coexistence, random walk, competing growth.

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#### 1 Introduction and main result

Frog models are growth models that have been studied in various guises for about 2 decades. They are also related to the activated random walks (see e.g. [7]) with sleep rate 0. The name has been attributed to R. Durrett, but the first occurrence of this class of models in print seems to be [9]. In such models, on a graph G = (V, E), at each site  $v \in V$  there are  $\eta(v)$  dormant frogs, where  $(\eta(v))_{v \in V}$  are i.i.d.  $\mathbb{Z}_+ = \{0, 1, \ldots\}$ -valued random variables with law  $\nu$ . Let  $\mu = \mathbb{E}[\eta(x)]$  denote the mean of  $\nu$ . Some chosen finite collection of sites (often just the origin), and all dormant frogs located at them, are activated at time 0. Each active frog moves independently of all others according to a simple (symmetric) random walk, and remains active thereafter. On the first visit of any (active) frog to a site  $v \in V$ , that site, and all dormant frogs there are instantly activated.

In this paper, time is discrete, and an active frog can only move at times  $n \in \mathbb{N} = \{1, 2, \ldots\}$  after being activated. In the *lazy* frog model, each active frog moves according to a lazy (symmetric) simple random walk, i.e. it moves with probability  $p \in (0, 1]$  on any given step (independent of previous steps and other frogs). By applying the subadditive ergodic theorem, shape theorems are known for such models see (see e.g. [1] when there is exactly one frog per site, and [2] more generally). In general dimensions the shape theorem describes the set of sites  $\xi_n$  that have been visited by time n, and takes the following form: there exists a non-empty convex set  $A \subset \mathbb{R}^d$  such that for any  $\varepsilon > 0$ , a.s. one has that  $(1 - \varepsilon)A \subset n^{-1}\xi_n \subset (1 + \varepsilon)A$  for all n sufficiently large.

<sup>&</sup>lt;sup>1</sup>The reader will hopefully excuse the awkward terminology here - that active frogs are lazy walkers!

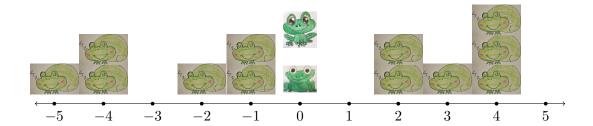


Figure 1: Our default initial configuration: frogs at the origin are activated at time 0 (at least one of each type), and all other frogs are dormant.

We assume throughout this paper that  $G = \mathbb{Z}$  (with nearest-neighbour edges) and that  $\mu \in (0,\infty)$ . In this setting  $\xi_n \subset \mathbb{Z}$  and the shape theorem takes a simpler form. Let  $R_n = \max \xi_n$  denote the right-most point visited by time n, and  $L_n = \min \xi_n$  denote the leftmost point. As stated in [4, Theorem 1.2] for p > 0 there exists a constant  $A = A(\nu, p) > 0$  such that

$$n^{-1}R_n \to A$$
 and  $n^{-1}L_n \to -A$ , as  $n \to \infty$ , almost surely. (1)

As the authors point out in [4] the proof of this result (where for p < 1 the frogs are lazy) requires only minor modifications from the case p = 1 handled in [2, 1].

Deijfen, Hirscher, and Lopes [4] introduced a model with two types of (lazy) frogs. Active frogs of type i have probability  $p_i \in (0,1]$  of moving on any given step, independent of all other frogs, but again an active frog never becomes dormant. When a frog of type  $i \in \{1, 2\}$ visits a site containing dormant frogs (i.e. a site previously unvisited by active frogs) all of the dormant frogs at that site become active frogs of type i. Since the model is defined in discrete time, it is possible that a site y containing dormant frogs is simultaneously visited by active frogs of two different types, in which case a tiebreaker rule is specified to decide which active frog is the one that activates the dormant frogs at y. Although it does not affect the validity of our main results, for definiteness we will assume that "the" activator is chosen uniformly at random (and independent of the past) from among those frogs that are first to arrive at a site. The locations of the active frogs at time 0 will also be largely unimportant for us, but elements of our proof require that there is a finite number of active frogs of each type initially. To fix ideas, we will assume that at time 0 there are  $\eta(x)$ dormant frogs at each  $x \neq 0$  and  $\eta_i(0)$  active frogs of type  $i \in \{1,2\}$  at the origin, and we will condition on  $\eta_1(0) \wedge \eta_2(0) > 0$ . See Figure 1. However as a tool for proving our main result, we will compare the evolution of the model with this initial distribution of active and dormant frogs with 1-type or 2-type frog models with different initial distributions.

Let  $N_n(i)$  denote the number of (active) frogs of type i at time n. Let  $G_i = \{N_n(i) \to \infty\}$ . We say that there is coexistence of frogs of types 1 and 2 if  $G_1 \cap G_2$  occurs. Deijfen et al [4] show that if  $p_1 = p_2 > 0$  then one has coexistence on  $\mathbb{Z}^d$  with positive probability. They ask [4, Section 1.3] what happens when  $p_1 \neq p_2$ , and surmise that if  $\mu < \infty$  then there is coexistence with positive probability if and only if  $A(p_1) = A(p_2)$  (where A(p) denotes the linear growth rate of the rightmost visited point of a one-type lazy frog model with

parameter p). In [5] it is shown that the initial configuration of active frogs (as long as it is finite) does not affect whether there is coexistence with positive probability. Our main result is the following:

**Theorem 1.** Let  $G = \mathbb{Z}$  and  $\mu \in (0, \infty)$ .

For any  $p_1, p_2 \in (0,1]$ ,  $\mathbb{P}(G_1 \cap G_2) > 0$ , i.e. with positive probability there will be infinitely many frogs of each kind.

This resolves the open problem of [4] on  $\mathbb{Z}$  (we do not know what happens in more than 1 dimension), except that it is a priori possible that A(p) does not depend on p. Thus, to ensure that our theorem has new content, we should provide examples where  $A(p) \neq A(p')$  if  $p \neq p'$ . Trivially  $A(p) \leq 1$  for all p. A simple coupling argument (see e.g. Lemma 4 below) shows that  $A(p) \leq A(p')$  when p < p'. We expect that the inequality is strict if A(p') < 1. We do not prove this here, but instead show that arbitrarily small growth rates are possible by taking p close to 0. This then gives us the aforementioned examples (for any p' > 0 one can find p < p' such that A(p) < A(p')).

**Proposition 2.** If  $\mu \in (0, \infty)$  then  $A(p) \downarrow 0$  as  $p \downarrow 0$ .

Proposition 2 follows from an elementary relation between the frog model and a particular branching random walk, together with standard bounds on the expected speed of the maximum of a branching random walk (see Section 5). As indicated above, from Theorem 1 and Proposition 2 we have the following corollary.

**Corollary 3.** For any  $\nu$  such that  $\mu \in (0, \infty)$  and any p > 0 there exists  $p' \in (0, p)$  with 0 < A(p') < A(p) such that one has coexistence for the corresponding 2-type frog model on  $\mathbb{Z}$  with positive probability.

Concepts of coexistence (or lack thereof) of two or more types of interacting particles have been considered in various other settings as well, including the setting of first passage percolation, see e.g. the survey [3] and more recently [6]. Our results give a small and relatively simple contribution to the literature in this general area.

To prove Theorem 1, we will consider the one-sided one-type model with  $\eta(x)$  dormant frogs at x > 0 and 0 frogs at y < 0 (and e.g.  $\eta(0)$  active frogs at 0 at time 0). Let  $A_+(p) = \liminf_{n \to \infty} n^{-1} R_n \ge 0$  denote the liminf linear growth rate of the right-most activated site, and  $A_-(p) = \liminf_{n \to \infty} n^{-1} L_n \le 0$  the liminf linear growth rate of the left-most. A simple coupling argument (see Lemma 4 below) shows that the liming linear growth rate  $\limsup_{n \to \infty} n^{-1} R_n$  is at most A(p). Note that  $A_+(p)$  and  $A_-(p)$  are a priori random, whereas A(p) has been proved to be deterministic. We will prove that almost surely  $A_+(p) = A(p)$  and  $A_-(p) = 0$ . Assuming that this is true it is fairly easy to see how to proceed - one then shows that with positive probability both of the following happen:

- (1) type 1 frogs only ever activate frogs to the right of 0, with the right-most (resp. left-most) site visited by type 1 frogs up to time n being roughly  $nA(p_1)$  (resp. not growing linearly with time), and
- (2) type 2 frogs only ever activate frogs to the left of 0, with the left-most (resp. right-most) site visited by type 2 frogs up to time n being roughly  $-nA(p_2)$  (resp. not growing linearly with time).

This then proves Theorem 1. Of course one needs to start with an active frog at time 0 to ensure that A(p) > 0, so we will always (implicitly or explicitly) condition on having at least one active frog at 0 at time 0.

In Section 2 we state and prove a simple monotonicity result (for one-type frog models) that will be used in other parts of the proof. In Section 3 we prove that  $A_{+}(p) = A(p)$  and  $A_{-}(p) = 0$ . Proposition 2 is proved in Section 5. Two-type frog models appear only in Section 4, where Theorem 1 is proved.

## 2 Frog domination

We use the notation (x,l) to denote the l-th frog at location x at time zero, where  $l \leq \eta(x)$ . For  $p \in (0,1]$ ,  $\eta = (\eta(x))_{x \in \mathbb{Z}} \in \mathbb{Z}^{\mathbb{Z}_+}$  and (finite or infinite)  $S \subset \mathbb{Z}$  we will denote by  $X(p,\eta,S)$  a (one-type) frog process on  $\mathbb{Z}$  with initial configuration  $\eta$  and initial activated set of sites S.

For  $\eta, \eta' \in \mathbb{Z}^{\mathbb{Z}_+}$  we write  $\eta \leq \eta'$  if  $\eta(x) \leq \eta'(x)$  for every  $x \in \mathbb{Z}$ . We have the following straightforward monotonicity result. Although this result is undoubtedly well-known, we will apply this result often, so we include a proof for completeness.

**Lemma 4.** For any  $\eta, \eta'$ , such that  $\eta \leq \eta'$ , and any S, S', such that  $S \subset S'$ . There exists a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  on which for every  $0 we can define frog processes <math>X(p, \eta, S)$  and  $X'(p', \eta', S')$  such that  $\xi_n \subset \xi'_n$  for every  $n \in \mathbb{Z}_+$ ,  $\mathbb{P}$ -almost surely.

Proof. Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space on which  $\Delta = (\Delta_{x,l,j})_{x \in \mathbb{Z}, l, j \in \mathbb{N}}$  is a collection of i.i.d. random variables, satisfying  $\mathbb{P}(\Delta_{x,l,j} = 1) = \mathbb{P}(\Delta_{x,l,j} = -1) = 1/2$  and  $(U_{x,l,k})_{x \in \mathbb{Z}, l, k \in \mathbb{Z}_+}$  are independent uniform random variables on [0,1], independent of  $\Delta$ . Given  $0 , <math>\eta \in \mathbb{Z}^{\mathbb{Z}_+}$  and  $S \subset \mathbb{Z}$ , let  $X(p,\eta,S)$  denote the frog process with  $\eta(x)$  frogs starting at x, with sites in S activated at time 0, and such that, once activated, the frog (x,l) with  $1 \le l \le \eta(x)$  conducts a lazy walk, stepping k time units after it was activated  $(k \ge 1)$  if  $U_{x,l,k} \le p$ . Its jth step is given by  $\Delta_{x,l,j}$ . Also let  $X'(p',\eta',S')$  denote the frog process with  $\eta'(x)$  frogs starting at x, with sites in S' activated at time 0, and such that, once activated, the frog (x,l) with  $l \le \eta'(x)$  conducts a lazy walk, stepping k time units after it was activated  $(k \ge 1)$  if  $U_{x,l,k} \le p'$ . Its jth step is given by  $\Delta_{x,l,j}$ .

Now we use the trivial facts that by construction on this space

- (\*) for fixed p the range (set of vertices visited) of any particle/frog  $\beta = (x, l)$  at time n is increasing in its time since activation,
- (\*\*) each frog (x,l) has the same sequence of moves in the X process and the X' process (given by the  $\Delta_{x,l,j}$  variables),
- $(\star \star \star)$  if  $U_{x,l,k} \leq p$  then  $U_{x,l,k} \leq p'$ , so the range of a single p frog is dominated by that of the corresponding p' frog, relative to their times of activation,

to conclude that all frogs in the X process are activated no earlier than in the X' process and their ranges trail those in the X' from that time forward.

## 3 One-sided frogs

Let  $\mathbb{P}^+$  denote the law of the one-sided frog model with  $(\eta(x))_{x\in\mathbb{Z}_+}$  i.i.d. each with law  $\nu$ , conditional on  $\eta(0)\geq 1$ , and where only the origin is activated at time 0. When proving our main result we will force the frogs to take a prescribed sequence of steps initially, and the evolution of each type of frog thereafter will be stochastically dominated by a frog model with a different law  $\mathbb{P}^*$  which we now define. Let  $\mathbb{P}^*$  denote the law of the one-sided frog model with  $\eta'(x)=\eta(x)+\eta(x+1)+\eta(0)$  frogs at  $x\geq 0$  at time 0 (where  $\eta(x)$  are i.i.d.  $\sim \mu$  except that  $\eta(0)$  is conditioned to be strictly positive), all of which are activated (so  $S=\mathbb{Z}_+$  is infinite), and no other frogs. Let  $\mathbb{P}^{*-}$  be defined in the same way but where the frogs are only at negative sites  $x\leq 0$ . By Lemma 4 the range of the frog model under  $\mathbb{P}^*$  dominates that under  $\mathbb{P}^+$ .

Let P(x) denote the original location of the frog that first reaches x (if there are ties, choose one from those that reach x first, uniformly at random).

**Lemma 5.** We have that  $\mathbb{P}^*(A_-(p) = 0) = 1$  for every  $p \in (0,1]$ . Moreover, for all  $\delta, \varepsilon > 0$  there exists  $m_0(\varepsilon, \delta) < \infty$  such that, for all  $m \ge m_0$ , and all  $p \in (0,1]$ ,

$$\mathbb{P}^* \left( \inf_{n \ge 0} \frac{L_n(p) + m}{n} < -\varepsilon \right) < \delta.$$

*Proof.* By Lemma 4 it suffices to prove the result in the case p=1, since the range is stochastically increasing in p.

Let  $T_x$  denote the first time that an active frog reaches x. Below, we use a (non-lazy) simple symmetric random walk  $(S_k)_{k\geq 0}$ , defined under a measure  $\mathbb{P}'$ . Then, for  $n\in\mathbb{N}$  the following holds

$$\begin{split} & \mathbb{P}^*(T_{-n} \leq n^{3/2}) \\ & = \sum_{y=0}^{\infty} \sum_{\ell=0}^{\infty} \sum_{k=0}^{\infty} \sum_{j=1}^{\infty} \mathbb{P}^* \left( T_{-n} \leq n^{3/2}, P(-n) = y, \eta(y) = \ell, \eta(y+1) = k, \eta(0) = j \right) \\ & \leq \sum_{y=0}^{\infty} \sum_{\ell \geq 0} \sum_{k \geq 0} \sum_{j \geq 1} \nu(\{\ell\}) \nu(\{k\}) \mathbb{P}^* (\eta(0) = j) \cdot (\ell + k + j) \cdot \mathbb{P}' \left( \max_{1 \leq k \leq n^{3/2}} S_k \geq n + y \right) \\ & \leq \frac{3\mu}{(1 - \nu(\{0\}))} \sum_{y=0}^{\infty} \mathbb{P}' \left( \max_{1 \leq k \leq n^{3/2}} S_k \geq n + y \right), \end{split}$$

where the first inequality is obtained by conditioning and using a union bound. Using a union bound, Chernoff's bound, and  $\cosh(t) \le e^{t^2/2}$  yields (for any t > 0)

$$\mathbb{P}^*(T_{-n} \le n^{3/2}) \le \frac{3\mu}{(1 - \nu(\{0\}))} \sum_{y=0}^{\infty} \sum_{k < n^{3/2}} \frac{(\cosh(t))^k}{e^{t(n+y)}} \le \frac{3\mu}{(1 - \nu(\{0\}))} n^{3/2} \sum_{y=0}^{\infty} \frac{e^{\frac{1}{2}t^2n^{3/2}}}{e^{t(n+y)}}.$$

Now take  $t = (n + y)n^{-3/2}$  and  $c = 3\mu(1 - \nu(\{0\}))^{-1}$  to see that

$$\mathbb{P}^*(T_{-n} \le n^{3/2}) \le c \cdot n^{3/2} \cdot \sum_{y=0}^{\infty} \exp\left\{-\frac{(n+y)^2}{2n^{3/2}}\right\}.$$

By comparing the sum to an integral and using standard Gaussian tail bounds, we obtain that

$$\mathbb{P}^*(T_{-n} \le n^{3/2}) \le cn^{3/2} \left( \exp\left\{ -\frac{n^2}{2n^{3/2}} \right\} + \int_0^\infty \exp\left\{ -\frac{(n+y)^2}{2n^{3/2}} \right\} dy \right)$$

$$= cn^{3/2} \left( \exp\left\{ -\sqrt{n}/2 \right\} + \int_n^\infty \exp\left\{ -\frac{y^2}{2n^{3/2}} \right\} dy \right)$$

$$\le cn^{3/2} \left( \exp\left\{ -\sqrt{n}/2 \right\} + \int_n^\infty \frac{y}{n} \exp\left\{ -\frac{y^2}{2n^{3/2}} \right\} dy \right)$$

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

$$\le 2cn^2 \exp\left\{ -\sqrt{n}/2 \right\}.$$
(3)

By the Borel-Cantelli Lemma,  $T_{-n} > n^{3/2}$  for all but finitely many n,  $\mathbb{P}^+$ -a.s., which implies that  $n/T_{-n}$  converges to 0  $\mathbb{P}^+$ -almost surely, which implies that  $\mathbb{P}^+(A_-(p)=0)=1$  for every p.

Let us prove the second statement. Since  $\liminf_{n\to\infty} n^{-1}L_n = 0$ , for any  $\delta, \varepsilon > 0$  there exists  $m_0 \in \mathbb{N}$  such that for all  $m \geq m_0$ ,

$$\mathbb{P}^* \left( \inf_{n \ge m_0} \frac{L_n}{n} < -\varepsilon \right) < \delta.$$

As we work in discrete time, for all  $0 \le n \le m$ ,  $L_n + m \ge 0$  almost surely, which proves the second statement.

**Corollary 6.** For every  $\delta > 0$  there exists  $k_{\delta} \in \mathbb{N}$  such that  $\mathbb{P}(P(x) \geq 0, \forall x \geq k_{\delta}) \geq 1 - \delta$ .

Proof. Let  $\bar{A}_+(p) = \limsup_{n \to \infty} n^{-1} R_n(p)$  denote the lim sup of the right-most visited site up to time n. By Lemma 5 and the definition of  $\mathbb{P}^{*-}$  we have that  $\mathbb{P}^{*-}(\bar{A}_+(p)=0)=1$ . By Lemma 4 this implies that for the one-sided model with  $\eta(x)$  frogs per site  $(x \le 0)$  and with all frogs activated at time zero, also  $\bar{A}_+(p)=0$ , a.s.

Now let  $R_n^{(-)}$  denote the right most point reached by frogs originating from sites x < 0 up to time n. For the two-sided model under  $\mathbb{P}$ , this quantity is stochastically dominated by the same quantity when all such frogs are activated at time 0. But the latter this grows sublinearly as above  $(\bar{A}_+(p) = 0 \text{ a.s.})$ . Since  $\mathbb{P}(A_+(p) = A(p)) = 1$  and A(p) > 0, this shows that there exists an a.s. finite random variable K such that  $\mathbb{P}(P(x) \ge 0, \forall x \ge K) = 1$ . The claim now follows since  $\mathbb{P}(K > k_{\delta}) < \delta$  for  $k_{\delta}$  sufficiently large.

Since the two-sided model under  $\mathbb{P}$  dominates the one-sided model under  $\mathbb{P}^+$ , the following implies that both models have the same linear growth rate for the right-most visited site.

**Lemma 7.** For any  $p \in (0, 1]$ ,

$$\mathbb{P}^+(A_+(p) = A(p)) = 1.$$

Proof. Let  $\delta > 0$ . From Corollary 6 there exists  $k_{\delta} \in \mathbb{N}$  such that  $\mathbb{P}(P(x) \geq 0, \forall x \geq k_{\delta}) \geq 1 - \delta$ . Since the activation time  $T_{k_{\delta}}^+$  of  $k_{\delta}$  in the 1-sided model is a.s. finite, there exists  $t_{\delta} > 0$  such that  $\mathbb{P}^+(T_{k_{\delta}}^+ \leq t_{\delta}) \geq 1 - \delta$ . Both of these statements are conditional on  $\eta(0) \geq 1$ , with (only) the origin activated at time 0.

A natural coupling  $\overline{\mathbb{P}}$  of these two processes is as follows. Under  $\overline{\mathbb{P}}$ , the initial configuration for X is  $(\eta(x))_{x\in\mathbb{Z}}$  which are i.i.d.~ $\nu$ , with  $\eta(0)$  conditioned to be strictly positive. The initial configuration  $(\eta^+(x))_{x\in\mathbb{Z}}$  for the one-sided process is defined as  $\eta^+(x) = 0$  if x < 0 and  $\eta^+(x) = \eta(x)$  otherwise. On top of this structure define two frog processes X and  $X^+$  from an i.i.d. collection of steps  $\Delta$  as in Lemma 4, with  $S = S^+ = \{0\}$ , so that X dominates  $X^+$ . On this space we therefore have

$$\overline{\mathbb{P}}\Big(\bigcap_{x>k_{\delta}} \{P(x) \ge 0\}, \, T_{k_{\delta}}^{+} \le t_{\delta}\Big) \ge 1 - 2\delta. \tag{4}$$

Now note that

$$\bigcap_{x\geq k_{\delta}} \{P(x)\geq 0\} \cap \{T_{k_{\delta}}^{+}\leq t_{\delta}\} \subset \bigcap_{x\geq k_{\delta}} \{T_{x}\leq T_{x}^{+}\leq T_{x}+t_{\delta}\},$$

where  $T_x$  denotes the activation time of x by the 2-sided process. Since  $R_{T_n}/T_n = n/T_n \to A(p)$  a.s. this shows that  $T_n^+/n \to A(p)^{-1}$  a.s. on the event in (4), and therefore also  $R_m^+/m \to A(p)$  a.s. on this event (consider  $m \in [T_n^+, T_{n+1}^+)$ ). This shows that  $\mathbb{P}^+(A^+(p) = A(p)) = \overline{\mathbb{P}}(A^+(p) = A(p)) \geq 1 - 2\delta$ . Since  $\delta$  was arbitrary this completes the proof.

**Corollary 8.** For any  $\delta > 0$  and  $p \in (0,1]$  there exists  $m_1(p,\delta) \in \mathbb{N}$  such that, for all  $m \geq m_1$ ,

$$\mathbb{P}^+\left(\inf_{n\geq 0}\frac{R_n+m}{n}>\frac{A(p)}{2}\right)>1-\delta.$$

*Proof.* On the almost sure (by Lemma 7) event  $\{A_+(p) = A(p)\}$ ,  $\inf_{n \geq m} R_n/n \to A(p)$  almost surely as m goes to infinity. Thus there exists  $m_1 < \infty$  such that for all  $m \geq m_1$ ,

$$\mathbb{P}^+\left(A_+(p) = A(p), \inf_{n \ge m} \frac{R_n}{n} > \frac{A(p)}{2}\right) > 1 - \delta.$$

Now, for all  $0 \le n \le m$ ,  $(R_n + m)/n \ge m/n \ge 1$ , which gives the conclusion.

**Remark 9.** Corollary 8 implies that, for any  $k \ge 1$  such that  $\mathbb{P}^+(\eta(0) = k) > 0$ , we have that

$$\mathbb{P}^{+} \left( \inf_{n \ge 0} \frac{R_n + m}{n} > \frac{A(p)}{2} \middle| \eta(0) = k \right) > 1 - \delta,$$

for all  $m \ge m_1(p, \delta \mathbb{P}^+ (\eta(0) = k)) \in \mathbb{N}$ . This is because (letting  $G_m$  denote the event being measured, and  $Q_k$  the event being conditioned on) we have for such m that

$$\mathbb{P}^{+}(G_{m}|Q_{k}) \ge \frac{\mathbb{P}^{+}(G_{m}) - \mathbb{P}^{+}(Q_{k}^{c})}{\mathbb{P}^{+}(Q_{k})} \ge \frac{1 - \delta \mathbb{P}^{+}(Q_{k}) - (1 - \mathbb{P}^{+}(Q_{k}))}{\mathbb{P}^{+}(Q_{k})}.$$

## 4 The two-type frog model

Let  $\mathbb{P}_p^+$  denote the law of the one-sided frog process, conditional on  $\eta(0) > 0$ , where now the laziness parameter p of the walkers appears explicitly as the subscript of the probability measure. We will denote by  $\mathbb{P}_{p_1,p_2}$  the law of a two-type frog model with  $\eta(x) \sim \nu$  dormant frogs at  $x \in \mathbb{Z}$  at time 0, and with  $\eta_i(0) \sim \nu$  active frogs of type  $i \in \{1,2\}$  at 0 at time 0 (with all frog counts independent of each other), conditional on  $\eta_1(0) > 0$  and  $\eta_2(0) > 0$ . Type i frogs have laziness parameter  $p_i \in (0,1]$ . Similarly, we will denote  $\mathbb{P}_p^*$  when we want to make the parameter p more explicit.

*Proof of Theorem 1.* We choose to provide a detailed proof of our main theorem, but the idea is fairly simple and easily obtained from Lemma 5, Corollary 8 and Remark 9.

Let  $p_1, p_2 > 0$  be given. Under the measure  $\mathbb{P}_{p_1, p_2}$ , let the following collection of random variables all be mutually independent:

- $(\eta(x))_{x \in \mathbb{Z} \setminus \{0\}}$ , all  $\sim \nu$  and  $\eta_1(0), \eta_2(0)$  having distribution  $\nu$  conditional on being strictly positive;
- $\Delta = (\Delta_{x,l,j})_{x \in \mathbb{Z}, l,j \in \mathbb{N}}$  satisfying  $\mathbb{P}_{p_1,p_2}(\Delta_{x,l,j} = 1) = \mathbb{P}_{p_1,p_2}(\Delta_{x,l,j} = -1) = 1/2$ ; and
- $(U_{x,l,k})_{x\in\mathbb{Z},l,k\in\mathbb{Z}_+}$  are independent standard uniform random variables.

Define

$$\eta_1(x) = \begin{cases} \eta(x), & \text{if } x > 0 \\ 0, & \text{if } x < 0, \end{cases}$$
 $\eta_2(x) = \begin{cases} 0, & \text{if } x > 0 \\ \eta(x), & \text{if } x < 0. \end{cases}$ 

We will define 3 processes on this space:

- $X^1$  denotes a one-sided one-type frog process with initial configuration  $\eta_1$  (with only the frogs at 0 activated at time 0), such that, once activated, the frog (x, l) with  $1 \le l \le \eta_1(x)$  conducts a lazy walk, stepping on the kth attempt  $(k \ge 1)$  if  $U_{x,l,k} \le p_1$ . Its j-th step is given by  $\Delta_{x,l,j}$ ;
- $X^2$  denotes a one-sided one-type frog process with initial configuration  $\eta_2$  (with only the frogs at 0 activated at time 0, and where the  $\eta_2(0)$  frogs started at 0 are labelled  $(0, \eta_1(0) + 1), \ldots, (0, \eta_1(0) + \eta_2(0))$ , such that, once activated, the frog (x, l) (with  $x \neq 0$  and  $1 \leq l \leq \eta_1(x)$ , or with x = 0 and  $l = \eta_1(0) + 1, \ldots, \eta_1(0) + \eta_2(0)$ ) conducts a lazy walk, stepping on the kth attempt  $(k \geq 1)$  if  $U_{x,l,k} \leq p_2$ . Its j-th step is given by  $\Delta_{x,l,j}$ .
- X denotes a two-sided two-type frog process, with initial configuration  $\eta$  (with only the frogs at 0 activated at time 0), where the frogs labelled  $(0,1),\ldots,(0,\eta_1(0))$  are type 1, and those labelled  $(0,\eta_1(0)+1),\ldots,(0,\eta_1(0)+\eta_2(0))$  are type 2. Once activated, if activated by a frog of type  $i \in \{1,2\}$ , the frog (x,l)  $1 \le l \le \eta(x)$  conducts a lazy walk, stepping on the kth attempt  $(k \ge 1)$  if  $U_{x,l,k} \le p_i$ . If  $k_i$  frogs of type i land on a previously unvisited site y at the same time, we use the random variable  $U_{y,0,0}$  to choose the activator all dormant frogs at y become type 2 if  $U_{y,0,0} \le k_2/(k_1 + k_2)$ , and otherwise they all become type 1.

Let  $\tau$  be the first time (possibly infinite) that in the process X a frog of type 2 activates a site in  $\mathbb{N}$  (i.e. a type 2 frog arrives at a positive site before any type 1 frog has), or a frog

of type 1 activates a site in  $-\mathbb{N}$ . On the event  $\{\tau = \infty\}$  for each  $i \in \{1,2\}$  we have that the moves of all the frogs in the frog process  $X^i$  follow exactly those of the frogs of type i in the frog process X.

Let  $R_n^i$  and  $L_n^i$  be the rightmost and leftmost position visited by active frogs of type  $i \in \{1,2\}$  up to time n for the  $X^i$  process. We will show that

$$\mathbb{P}_{p_1,p_2}\left(\sup_{n\in\mathbb{N}}(R_n^1\wedge -L_n^2) = \infty, \tau = \infty\right) > 0.$$
 (5)

On the event in (5) we have coexistence because type 1 frogs activate every positive site, and frogs of type 2 every negative site. Clearly the first event in (5) has probability 1 since simple random walk has infinite range almost surely (as a Markov chain on an infinite irreducible class:  $\mathbb{Z}$ ), so we need only show that  $\mathbb{P}_{n_1,n_2}(\tau=\infty) > 0$ .

irreducible class:  $\mathbb{Z}$ ), so we need only show that  $\mathbb{P}_{p_1,p_2}(\tau=\infty)>0$ . Let  $\ell_{\min}=\min\{\ell\geq 0:\nu(\{\ell\})>0\}$  and  $\ell_{\min}^+=\min\{\ell>0:\nu(\{\ell\})>0\}$ . Define  $\varepsilon=\min\{A(p_1),A(p_2)\}/4$  and let us fix  $m\in\mathbb{N}$  so that  $m=m_0(\varepsilon,1/5)+m_1(p_1,\nu(\{\ell_{\min}^+\})/5)+m_1(p_2,\nu(\{\ell_{\min}^+\})/5)$ , where  $m_0$  and  $m_1$  are given by Lemma 5 and Corollary 8.

Now, let  $A_m^1$  be the event that  $\eta_1(0) = \ell_{\min}^+$ ,  $\eta(x) = \ell_{\min}$  for all  $1 \le x < 2m$  and that, in the first 2m time units, for the process X, the frogs of type 1 at the origin all take 2m steps to the right while all the activated frogs from the region [1, 2m - 1] take alternate steps left and right (in that order). Let  $A_m^2$  be the event that  $\eta_2(0) = \ell_{\min}^+$ ,  $\eta(x) = \ell_{\min}$  for all  $-2m < x \le -1$  and that, in the first 2m time units, the frogs of type 2 at the origin all take 2m steps to the left while all the activated frogs from the region [-(2m-1), -1] alternate stepping right and left (in that order). Let  $A_m = A_m^1 \cap A_m^2$  (see Figure 2).

Recall that  $\varepsilon = \min\{A(p_1), A(p_2)\}/4$ , and define

$$B_m^{1,l} = \left\{ \inf_{n \ge 0} \frac{L_{n+2m}^1 + m}{n} \ge -\varepsilon \right\} \tag{6}$$

$$B_m^{1,r} = \left\{ \inf_{n \ge 0} \frac{R_{n+2m}^1 - m}{n} > \frac{A(p_1)}{2} \right\}$$
 (7)

$$B_m^{2,l} = \left\{ \sup_{n>0} \frac{L_{n+2m}^2 + m}{n} < -\frac{A(p_2)}{2} \right\}$$
 (8)

$$B_m^{2,r} = \left\{ \sup_{n \ge 0} \frac{R_{n+2m}^2 - m}{n} \le \varepsilon \right\}. \tag{9}$$

Also define

$$B_m^i = B_m^{i,l} \cap B_m^{i,r}, \quad \text{and} \quad B_m = B_m^1 \cap B_m^2.$$

Note that  $A_m \subset \{\tau > 2m\}$  while on  $B_m$  we have for any  $n \geq 0$  that

$$L^1_{n+2m} \geq -n\varepsilon - m > -nA(p_2)/2 - m \geq L^2_{n+2m},$$

and similarly  $R_{n+2m}^1 > R_{n+2m}^2$ , so  $\tau \neq n+2m$  for any n on  $B_m$ . This shows that  $A_m \cap B_m \subset \{\tau = \infty\}$ .

It therefore remains to show that  $\mathbb{P}_{p_1,p_2}(A_m \cap B_m) > 0$ . The event  $A_m$  does not depend on  $\eta(x) = \eta_i(x)$  for  $|x| \geq 2m$ . On the other hand, we have that  $\mathbb{P}_{p_1,p_2}(A_m) > 0$ , for all  $m \geq 1$ , since  $A_m$  only prescribes the value of the environment in a finite box and requires a

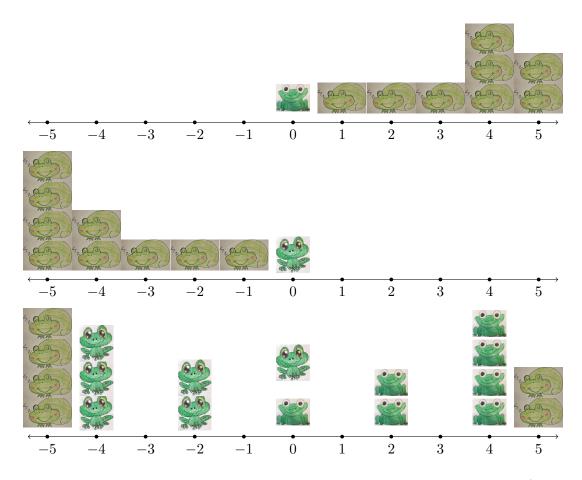


Figure 2: A depiction of the event  $A_m$  when m=2 and  $\ell_{\min}=1$ . The first (resp. second) diagram shows an example of an initial configuration  $\eta_1$  (resp.  $\eta_2$ ) on the event  $A_m$ . The third diagram shows the corresponding configuration of frogs at time 2m=4, in particular, the original type-1 (resp. type 2) frog at the origin at time 0 is now at 2m (resp. -2m).

bounded number of frogs to perform a fixed finite set of moves. Let us briefly justify that, conditional on  $A_m$ , the following dominations<sup>2</sup> hold:

- 1.  $(R_{n+2m}^1 2m)_{n \geq 0}$  stochastically dominates the right-most position of a one-sided frog process  $(R_n)_{n \geq 0}$  under  $\mathbb{P}_{p_1}^+$ , conditioned on  $\{\eta(0) = \ell_{\min}^+\}$ ;
- 2.  $(L_{n+2m}^1)_{n>0}$  stochastically dominates  $(L_n)_{n>0}$  under  $\mathbb{P}_{p_1}^*$ ;
- 3.  $(-L_{n+2m}^2 2m)_{n\geq 0}$  stochastically dominates the right-most position of a one-sided frog process  $(R_n)_{n\geq 0}$  under  $\mathbb{P}_{p_2}^+$ , conditioned on  $\{\eta(0) = \ell_{\min}^+\}$ ;
- 4.  $(-R_{n+2m}^2)_{n\geq 0}$  stochastically dominates the left-most position of a one-sided frog process  $(L_n)_{n>0}$  under  $\mathbb{P}_{p_2}^*$ .

We will prove the first two items, as the other two follow by symmetry. For this purpose, let us denote  $N_1(x)$  the number of active frogs on site x for the  $X^1$  process at time 2m, and see Figure 2.

Note that, on  $A_m$ , at time 2m,  $N_1(x) = 0$  for  $x \le -1$  and x odd,  $N_1(0) = \ell_{\min}$ ,  $N_1(x) = 2\ell_{\min}$  for all even  $x \in [1, 2m - 1]$ , and finally  $N_1(2m) = \eta_1(2m) + \ell_{\min}^+$ . On all  $x \ge 2m + 1$ , there are  $\eta_1(x)$  dormant frogs.

The first item follows by the monotonicity provided by Lemma 4 and using that  $\eta_1(x)$  has the same law as  $\eta_1(x-2m)$  for all x>2m, as there are at least  $\ell_{\min}^+$  active frogs on 2m. To prove item 2, note that for every  $x\geq 0$ ,  $N_1(x)\leq \eta_1(0)+\eta_1(x)+\eta_1(x+1)$ , and use the definition of  $\ell_{\min}^+$  together with Lemma 4.

Thus, by our choice of  $\varepsilon$  and m, using Lemma 5 we obtain

$$\mathbb{P}_{p_1,p_2}\left((B_m^{1,l})^c\Big|A_m\right) \le \mathbb{P}_{p_1}^*\left(\inf_{n\ge 0}\frac{L_n+m}{n} < -\varepsilon\right) \le \frac{1}{5}.$$

Similarly, by Corollary 8 together with Remark 9, we obtain

$$\mathbb{P}_{p_1,p_2}\Big((B_m^{1,r})^c\Big|A_m\Big) = \mathbb{P}_{p_1,p_2}\Big(\inf_{n\geq 0} \frac{R_{n+2m}^{+1} - 2m + m}{n} \leq \frac{A(p_1)}{2}\Big|A_m\Big)$$
$$\leq \mathbb{P}_{p_1}^+\Big(\inf_{n\geq 0} \frac{R_n + m}{n} \leq \frac{A(p_1)}{2}\Big|\eta(0) = \ell_{\min}^+\Big) \leq \frac{1}{5}.$$

Similarly,

$$\mathbb{P}_{p_1, p_2} \left( (B_m^{2,r})^c \middle| A_m \right) \le 1/5,$$

$$\mathbb{P}_{p_1, p_2} \left( (B_m^{2,l})^c \middle| A_m \right) \le 1/5.$$

and thus  $\mathbb{P}_{p_1,p_2}(B_m|A_m) \geq 1/5$ . This proves that  $\mathbb{P}_{p_1,p_2}(A_m \cap B_m) > 0$  as claimed.

<sup>&</sup>lt;sup>2</sup>We say that a  $\mathbb{Z}$ -valued process  $\mathbf{W} = (W_n)_{n\geq 0}$  stochastically dominates  $\mathbf{Y} = (Y_n)_{n\geq 0}$  if there exists a probability space with processes  $\mathbf{W}' \sim \mathbf{W}$  and  $\mathbf{Y}' \sim \mathbf{Y}$  on which  $W_n' \geq Y_n'$  for every n a.s.

## 5 Extremely lazy frogs

In this section we return to studying the one-type frog model.

Proof of Proposition 2. Recall that  $\nu$  is the probability measure on  $\mathbb{Z}_+$  that is the law of  $\eta(x)$ . Let  $\mu = \mathbb{E}[\eta(x)] < \infty$ .

Consider a branching random walk where each individual (independently of other particles) has:

- (i) exactly one offspring, with displacement 0, with probability 1 p,
- (ii) exactly 1+k offspring, all at displacement +1, with probability  $\nu(\{k\})\frac{p}{2}$ ,
- (iii) exactly 1+k offspring at displacement -1, with probability  $\nu(\{k\})^{\frac{p}{2}}$ ,

and then the parent particle dies immediately. The number of offspring is always at least 1, with mean  $1+p\mu$ . It is easy to couple this branching random walk with our lazy frog model such that the set of visited points in the BRW contains that of the frog model. To see this note that in the lazy frog model an active frog at x either doesn't move (with probability 1-p), and then it doesn't activate any new frogs, or it moves left or right with probability p/2 each, and then it activates at most  $\eta(x+1)$  or  $\eta(x-1)$  new frogs.

It is therefore sufficient to show that the speed of the front (the maximal site visited) for the BRW model goes to zero as  $p \downarrow 0$ . We will verify this by applying small modifications to standard results (in particular [8, Lemma 1.5]) that are typically stated for the position of the right-most particle at time n, (so this differs slightly from the quantity we are after).

Firstly note that since  $R_n/n$  converges almost surely and is bounded above by 1 for every n, we have by dominated convergence that the limiting speed A(p) for the lazy frog model with parameter p is equal to  $\limsup_{n\to\infty} \mathbb{E}_p[n^{-1}R_n]$ . By the above coupling it is therefore enough for us to show that

$$\lim_{n \to \infty} \sup_{n \to \infty} \mathbb{E}_p[n^{-1}M_n^*] \to 0, \text{ as } p \to 0,$$
(10)

where  $M_n^*$  is the largest site visited by the Branching random walk up to (and including) time n. Let |x| denote the generation of a particle x in the BRW, and V(x) denote the location of x. Then

$$\frac{1}{n}\mathbb{E}_p[M_n^*] = \frac{1}{n}\mathbb{E}_p\left[\max_{x:|x| \le n} V(x)\right] \le \frac{1}{n}\log\left(\mathbb{E}_p\left[e^{\max_{x:|x| \le n} V(x)}\right]\right) \tag{11}$$

$$\leq \frac{1}{n} \log \left( \mathbb{E}_p \left[ \sum_{x:|x| \leq n} e^{V(x)} \right] \right) = \frac{1}{n} \log \left( \sum_{k=0}^n \mathbb{E}_p \left[ \sum_{x:|x|=k} e^{V(x)} \right] \right). \tag{12}$$

By conditioning on generation k-1 of the branching process, one has that

$$\mathbb{E}_{p}\left[\sum_{x:|x|=k} e^{V(x)}\right] = \mathbb{E}_{p}\left[\sum_{x:|x|=k-1} e^{V(x)}\right] e^{\psi_{p}(1)},\tag{13}$$

where, for t > 0,

$$\psi_p(t) = \log \mathbb{E}_p \left[ \sum_{x:|x|=1} e^{V(x)} \right] = \log \left( (1-p) + \frac{p}{2} \sum_{k=0}^{\infty} \nu(\{k\})(k+1)(e^{-t} + e^t) \right). \tag{14}$$

Thus, by induction, we obtain  $\mathbb{E}_p\left[\sum_{x:|x|=k}e^{V(x)}\right]=e^{k\psi_p(1)}$ . Alternatively, we could have applied the Many-To-One Lemma (see [8, Theorem 1.1]).

Thus we have

$$\frac{1}{n}\mathbb{E}_p[M_n^*] \le \frac{1}{n}\log\left(\sum_{k=0}^n e^{k\psi_p(1)}\right),\tag{15}$$

where

$$\psi_p(1) = \log\left(1 + p\left[\frac{1}{2}(e^{-1} + e)(1 + \mu) - 1\right]\right) \ge 0.$$
 (16)

Thus

$$\frac{1}{n}\mathbb{E}_p[M_n^*] \le \frac{1}{n}\log\left((n+1)e^{n\psi_p(1)}\right) = \frac{\log(n+1)}{n} + \psi_p(1). \tag{17}$$

Since  $\log(1+x) \le x$ , we see that  $\psi_p(1)$  can be made arbitrarily small by making p arbitrarily small, which completes the proof.

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