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Lower bounds for bootstrap percolation on Galton–Watson trees

Karen Gunderson* Michał Przykucki†

Abstract

Bootstrap percolation is a cellular automaton modelling the spread of an ‘infection’ on a graph. In this note, we prove a family of lower bounds on the critical probability for r -neighbour bootstrap percolation on Galton–Watson trees in terms of moments of the offspring distributions. With this result we confirm a conjecture of Bollobás, Gunderson, Holmgren, Janson and Przykucki. We also show that these bounds are best possible up to positive constants not depending on the offspring distribution.

Keywords: bootstrap percolation; Galton–Watson trees.

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1 Introduction

Bootstrap percolation, a type of cellular automaton, was introduced by Chalupa, Leath and Reich [1] and has been used to model a number of physical processes. Given a graph G and threshold $r \geq 2$, the r -neighbour bootstrap process on G is defined as follows: Given $A \subseteq V(G)$, set $A_0 = A$ and for each $t \geq 1$, define

$$A_t = A_{t-1} \cup \{v \in V(G) : |N(v) \cap A_{t-1}| \geq r\},$$

where $N(v)$ is the neighbourhood of v in G . The closure of a set A is $\langle A \rangle = \bigcup_{t \geq 0} A_t$. Often the bootstrap process is thought of as the spread, in discrete time steps, of an ‘infection’ on a graph. Vertices are in one of two states: ‘infected’ or ‘healthy’ and a vertex with at least r infected neighbours becomes itself infected, if it was not already, at the next time step. For each t , the set A_t is the set of infected vertices at time t . A set $A \subseteq V(G)$ of initially infected vertices is said to *percolate* if $\langle A \rangle = V(G)$.

Usually, the behaviour of bootstrap processes is studied in the case where the initially infected vertices, i.e., the set A , are chosen independently at random with a fixed probability p . For an infinite graph G the *critical probability* is defined by

$$p_c(G, r) = \inf\{p : \mathbb{P}_p(\langle A \rangle = V(G)) > 0\}.$$

This is different from the usual definition of critical probability for finite graphs, which is generally defined as the infimum of the values of p for which percolation is more likely to occur than not.

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In this paper, we consider bootstrap percolation on Galton–Watson trees and answer a conjecture in [3] on lower bounds for their critical probabilities. For any offspring distribution ξ on $\mathbb{N} \cup \{0\}$, let T_ξ denote a random Galton–Watson tree (the family tree of a Galton–Watson branching process) with offspring distribution ξ which we define as follows. Starting with a single root vertex in level 0, at each generation $n = 1, 2, 3, \dots$ every vertex in level $n - 1$ gives birth to a random number of children in level n , where for every vertex the number of offspring is distributed according to the distribution ξ and is independent of the number of children of any other vertex. For any fixed offspring distribution ξ , the critical probability $p_c(T_\xi, r)$ is almost surely a constant (see Lemma 3.2 in [3]) and we shall give lower bounds on the critical probability in terms of various moments of ξ .

Bootstrap processes on infinite regular trees were first considered by Chalupa, Leath and Reich [1]. Later, Balogh, Peres and Pete [2] studied bootstrap percolation on arbitrary infinite trees and one particular example of a random tree given by a Galton–Watson branching process. In [3], Galton–Watson branching processes were further considered, and it was shown that for every $r \geq 2$, there is a constant $c_r > 0$ so that

$$p_c(T_\xi, r) \geq \frac{c_r}{\mathbb{E}[\xi]} \exp\left(-\frac{\mathbb{E}[\xi]}{r-1}\right)$$

and in addition, for every $\alpha \in (0, 1]$, there is a positive constant $c_{r,\alpha}$ so that,

$$p_c(T_\xi, r) \geq c_{r,\alpha} (\mathbb{E}[\xi^{1+\alpha}])^{-1/\alpha}. \tag{1.1}$$

Additionally, in [3] it was conjectured that for any $r \geq 2$, inequality (1.1) holds for any $\alpha \in (0, r - 1]$. As our main result, we show that this conjecture is true. For the proofs to come, some notation from [3] is used. If an offspring distribution ξ is such that $\mathbb{P}(\xi < r) > 0$, then one can easily show that $p_c(T_\xi, r) = 1$. With this in mind, for r -neighbour bootstrap percolation, we only consider offspring distributions with $\xi \geq r$ almost surely.

Definition 1.1. For every $r \geq 2$ and $k \geq r$, define

$$g_k^r(x) = \frac{\mathbb{P}(\text{Bin}(k, 1-x) \leq r-1)}{x} = \sum_{i=0}^{r-1} \binom{k}{i} x^{k-i-1} (1-x)^i$$

and for any offspring distribution ξ with $\xi \geq r$ almost surely, define

$$G_\xi^r(x) = \sum_{k \geq r} \mathbb{P}(\xi = k) g_k^r(x).$$

Some facts, which can be proved by induction, about these functions are used in the proofs to come. For any $r \geq 2$, we have $g_r^r(x) = \sum_{i=0}^{r-1} (1-x)^i$ and for any $k > r$,

$$g_r^r(x) - g_k^r(x) = \sum_{i=r}^{k-1} \binom{i}{r-1} x^{i-r} (1-x)^r. \tag{1.2}$$

Hence, for all distributions ξ we have $G_\xi^r(x) \leq g_r^r(x)$ for $x \in [0, 1]$.

Developing a formulation given by Balogh, Peres and Pete [2], it was shown in [3] (see Theorem 3.6 in [3]) that if $\xi \geq r$, then

$$p_c(T_\xi, r) = 1 - \frac{1}{\max_{x \in [0,1]} G_\xi^r(x)}. \tag{1.3}$$

2 Results

In this section, we shall prove a family of lower bounds on the critical probability $p_c(T_\xi, r)$ based on the $(1+\alpha)$ -moments of the offspring distributions ξ for all $\alpha \in (0, r-1]$, using a modification of the proofs of Lemmas 3.7 and 3.8 in [3] together with some properties of the gamma function and the beta function.

The gamma function is given, for z with $\Re(z) > 0$, by $\Gamma(z) = \int_0^\infty t^{z-1} \exp(-t) dt$ and for all $n \in \mathbb{N}$, satisfies $\Gamma(n) = (n-1)!$. The beta function is given, for $\Re(x), \Re(y) > 0$, by $B(x, y) = \int_0^1 t^{x-1} (1-t)^{y-1} dt$ and satisfies $B(x, y) = \frac{\Gamma(x)\Gamma(y)}{\Gamma(x+y)}$. We shall use the following bounds on the ratio of two values of the gamma function obtained by Gautschi [4]. For $n \in \mathbb{N}$ and $0 \leq s \leq 1$ we have

$$\left(\frac{1}{n+1}\right)^{1-s} \leq \frac{\Gamma(n+s)}{\Gamma(n+1)} \leq \left(\frac{1}{n}\right)^{1-s}. \tag{2.1}$$

Let us now state our main result.

Theorem 2.1. *For each $r \geq 2$ and $\alpha \in (0, r-1]$, there exists a constant $c_{r,\alpha} > 0$ such that for any offspring distribution ξ with $\mathbb{E}[\xi^{1+\alpha}] < \infty$, we have*

$$p_c(T_\xi, r) \geq c_{r,\alpha} (\mathbb{E}[\xi^{1+\alpha}])^{-1/\alpha}.$$

We prove Theorem 2.1 in two steps. First, in Lemma 2.2, we show that it holds for $\alpha \in (0, r-1)$. Then, in Lemma 2.3, we consider the case $\alpha = r-1$.

Lemma 2.2. *For all $r \geq 2$ and $\alpha \in (0, r-1)$, there exists a positive constant $c_{r,\alpha}$ such that for any distribution ξ with $\mathbb{E}[\xi^{1+\alpha}] < \infty$, we have*

$$p_c(T_\xi, r) \geq c_{r,\alpha} (\mathbb{E}[\xi^{1+\alpha}])^{-1/\alpha}.$$

Proof. Fix $r \geq 2$, $\alpha \in (0, r-1)$ with $\alpha \notin \mathbb{Z}$ and an offspring distribution ξ . Set $t = \lfloor \alpha \rfloor$ and $\varepsilon = \alpha - t$ so that $\varepsilon \in (0, 1)$ and t is an integer with $t \in [0, r-2]$. Set $M = \max_{x \in [0,1]} G_\xi^r(x)$ and fix $y \in [0, 1]$ with the property that $g_r^r(1-y) = M$. Such a y can always be found since $G_\xi^r(x) \leq g_r^r(x)$ in $[0, 1]$, $G_\xi^r(1) = g_r^r(1) = 1$ and $g_r^r(x)$ is continuous. Thus, $M = 1 + y + \dots + y^{r-1}$ and so by equation (1.3)

$$p_c(T_\xi, r) = 1 - \frac{1}{M} = \frac{y(1-y^{r-1})}{1-y^r} \geq \frac{r-1}{r}y. \tag{2.2}$$

A lower bound on $p_c(T_\xi, r)$ is given by considering upper and lower bounds for the integral $\int_0^1 \frac{g_r^r(x) - G_\xi^r(x)}{(1-x)^{2+\alpha}} dx$.

For the upper bound, using the definition of the beta function, for every $k \geq r$

$$\begin{aligned} \int_0^1 \frac{g_r^r(x) - g_k^r(x)}{(1-x)^{\alpha+2}} dx &= \sum_{i=r}^{k-1} \binom{i}{r-1} \int_0^1 x^{i-r} (1-x)^{r-2-\alpha} dx \quad (\text{by eq. (1.2)}) \\ &= \sum_{i=r}^{k-1} \binom{i}{r-1} B(i-r+1, r-1-\alpha) \\ &= \sum_{i=r}^{k-1} \frac{i!}{(r-1)!(i-r+1)!} \frac{(i-r)!\Gamma(r-1-\alpha)}{\Gamma(i-\alpha)} \\ &= \sum_{i=r}^{k-1} \frac{i(i-1)\dots(i-t)\Gamma(i-t)}{(i-r+1)\Gamma(i-t-\varepsilon)} \\ &\quad \cdot \frac{\Gamma(r-1-t-\varepsilon)}{(r-1)(r-2)\dots(r-1-t)\Gamma(r-1-t)}. \end{aligned} \tag{2.3}$$

Let $c_1 = c_1(r, \alpha) = \frac{\Gamma(r-1-t-\varepsilon)}{(r-1)(r-2)\dots(r-1-t)\Gamma(r-1-t)}$. Note that by inequality (2.1), for $t < r - 2$, $\frac{\Gamma(r-1-t-\varepsilon)}{\Gamma(r-1-t)} \geq \frac{1}{(r-1-t)^\varepsilon}$ and so $c_1 \geq \frac{1}{(r-1)^{t+\varepsilon}} = (r-1)^{-\alpha}$. On the other hand, if $t = r - 2$, then $c_1 = \frac{\Gamma(1-\varepsilon)}{(r-1)!} = \frac{\Gamma(2-\varepsilon)}{(1-\varepsilon)(r-1)!} \geq \frac{1}{2(r-1)!(1-\varepsilon)}$.

Thus, continuing equation (2.3), applying inequality (2.1) again yields

$$\begin{aligned} & \sum_{i=r}^{k-1} \frac{i(i-1)\dots(i-t)\Gamma(i-t)}{(i-r+1)\Gamma(i-t-\varepsilon)} \cdot \frac{\Gamma(r-1-t-\varepsilon)}{(r-1)(r-2)\dots(r-1-t)\Gamma(r-1-t)} \\ & \leq c_1 \sum_{i=r}^{k-1} \frac{i}{i-r+1} (i-1)(i-2)\dots(i-t)(i-t)^\varepsilon \\ & \leq rc_1 \sum_{i=r}^{k-1} i^{t+\varepsilon} \\ & \leq rc_1 k^{1+t+\varepsilon} = rc_1 k^{1+\alpha}. \end{aligned}$$

Thus, taking expectation over k with respect to ξ ,

$$\int_0^1 \frac{g_r^r(x) - G_\xi^r(x)}{(1-x)^{2+\alpha}} dx \leq rc_1 \mathbb{E}[\xi^{1+\alpha}]. \tag{2.4}$$

Consider now a lower bound on the integral:

$$\begin{aligned} \int_0^1 \frac{g_r^r(x) - G_\xi^r(x)}{(1-x)^{2+\alpha}} dx & \geq \int_0^{1-y} \frac{g_r^r(x) - M}{(1-x)^{2+\alpha}} dx \\ & = \int_0^{1-y} -\frac{(M-1)}{(1-x)^{2+\alpha}} + \sum_{i=0}^{r-2} \frac{1}{(1-x)^{1+\alpha-i}} dx \\ & = \left[-\frac{(M-1)}{(\alpha+1)(1-x)^{1+\alpha}} + \sum_{i=0}^{r-2} \frac{1}{(\alpha-i)(1-x)^{\alpha-i}} \right]_0^{1-y} \\ & = -\frac{(M-1)}{(\alpha+1)} \left(\frac{1}{y^{1+\alpha}} - 1 \right) + \sum_{i=0}^t \frac{1}{\alpha-i} \left(\frac{1}{y^{\alpha-i}} - 1 \right) + \sum_{i=t+1}^{r-2} \frac{1-y^{i-\alpha}}{i-\alpha} \\ & = \frac{1}{y^\alpha} \left(\frac{M-1}{\alpha+1} \left(\frac{y^{\alpha+1}-1}{y} \right) + \sum_{i=0}^t \frac{y^i - y^\alpha}{\alpha-i} + \sum_{i=t+1}^{r-2} \frac{y^\alpha - y^i}{i-\alpha} \right) \\ & = \frac{1}{y^\alpha} \left(\frac{(1+y+y^2+\dots+y^{r-2})(y^{\alpha+1}-1)}{(\alpha+1)} + \sum_{i=0}^t \frac{y^i - y^\alpha}{\alpha-i} + \sum_{i=t+1}^{r-2} \frac{y^\alpha - y^i}{i-\alpha} \right) \\ & = \frac{1}{y^\alpha} \left(\frac{-1}{\alpha+1} + \frac{1}{\alpha} + \sum_{i=1}^t \left(\frac{y^i}{\alpha-i} - \frac{y^i}{\alpha+1} \right) + \sum_{i=0}^{r-2} \frac{y^{\alpha+1+i}}{\alpha+1} - \sum_{i=t+1}^{r-2} \frac{y^i}{\alpha+1} - \sum_{i=0}^t \frac{y^\alpha}{\alpha-i} \right. \\ & \quad \left. + \sum_{i=t+1}^{r-2} \frac{y^\alpha - y^i}{i-\alpha} \right) \\ & \geq \frac{1}{y^\alpha} \left(\frac{1}{\alpha(\alpha+1)} - \frac{y^{t+1}}{\alpha+1} - \sum_{i=0}^t \frac{y^\alpha}{\alpha-i} \right) \\ & \geq \frac{1}{y^\alpha} \left(\frac{1}{\alpha(\alpha+1)} - y^\alpha \sum_{i=0}^{t+1} \frac{1}{\alpha+1-i} \right). \end{aligned}$$

Set $c_2 = c_2(\alpha) = \sum_{i=0}^{t+1} \frac{1}{\alpha+1-i}$ and consider separately two different cases. For the

first, if $y^\alpha c_2 \geq \frac{1}{2\alpha(\alpha+1)}$ then since $\mathbb{E}[\xi^{\alpha+1}] \geq 1$,

$$y^\alpha \geq \frac{1}{2\alpha(\alpha+1)c_2} \geq \frac{1}{2\alpha(\alpha+1)c_2} \mathbb{E}[\xi^{1+\alpha}]^{-1}.$$

Thus, if $c'_2 = \left(\frac{1}{2\alpha(\alpha+1)c_2}\right)^{1/\alpha}$, then $y \geq c'_2 \mathbb{E}[\xi^{1+\alpha}]^{-1/\alpha}$.

In the second case, if $y^\alpha < \frac{1}{2\alpha(\alpha+1)c_2}$, then

$$\int_0^1 \frac{g_r^r(x) - G_\xi^r(x)}{(1-x)^{2+\alpha}} dx \geq \frac{1}{y^\alpha} \frac{1}{2\alpha(\alpha+1)}. \tag{2.5}$$

Combining equation (2.5) with equation (2.4) yields

$$y^\alpha \geq \frac{1}{2\alpha(\alpha+1)} \frac{1}{rc_1} \mathbb{E}[\xi^{1+\alpha}]^{-1}$$

and setting $c'_1 = (2\alpha(\alpha+1)rc_1)^{-1/\alpha}$ gives $y \geq c'_1 \mathbb{E}[\xi^{1+\alpha}]^{-1/\alpha}$.

Finally, set $c_{r,\alpha} = \frac{r-1}{r} \min\{c'_1, c'_2\}$ so that by inequality (2.2) we obtain,

$$p_c(T_\xi, r) \geq \frac{r-1}{r} y \geq c_{r,\alpha} \mathbb{E}[\xi^{1+\alpha}]^{-1/\alpha}.$$

For every natural number $n \in [1, r-2]$, note that $\lim_{\alpha \rightarrow n^-} c_{r,\alpha} > 0$ and, by the monotone convergence theorem, there is a constant $c_{r,n} > 0$ so that

$$p_c(T_\xi, r) \geq c_{r,n} \mathbb{E}[\xi^{1+n}]^{-1/n}.$$

This completes the proof of the lemma. □

In the above proof, as $\alpha \rightarrow (r-1)^-$, $c_1(r, \alpha) \rightarrow \infty$ and hence $\lim_{\alpha \rightarrow (r-1)^-} c_{r,\alpha} = 0$, so the proof of Lemma 2.2 does not directly extend to the case $\alpha = r-1$. We deal with this problem in the next lemma. Using a different approach we prove an essentially best possible lower bound on $p_c(T_\xi, r)$ based on the r -th moment of the distribution ξ . The sharpness of our bound is demonstrated by the b -branching tree T_b , a Galton–Watson tree with a constant offspring distribution, for which, as a function of b , we have $p_c(T_b, r) = (1 + o(1))(1 - 1/r) \left(\frac{(r-1)!}{br}\right)^{1/(r-1)}$ (see Lemma 3.7 in [3]).

Lemma 2.3. *For any $r \geq 2$ and any offspring distribution ξ with $\mathbb{E}[\xi^r] < \infty$,*

$$p_c(T_\xi, r) \geq \left(1 - \frac{1}{r}\right) \left(\frac{(r-1)!}{\mathbb{E}[\xi^r]}\right)^{1/(r-1)}.$$

Proof. As in the proof of Lemma 3.7 of [3] note that for every $k \geq r$ and $t \in [0, 1]$,

$$\begin{aligned} g_k^r(1-t) &= \frac{\mathbb{P}(\text{Bin}(k, t) \leq r-1)}{1-t} = \frac{1 - \mathbb{P}(\text{Bin}(k, t) \geq r)}{1-t} \\ &\geq \frac{1 - \binom{k}{r} t^r}{1-t} \geq \frac{1 - \frac{1}{r!} k^r t^r}{1-t}. \end{aligned} \tag{2.6}$$

Using the lower bound in inequality (2.6) for the function $G_\xi^r(x)$ yields

$$G_\xi^r(1-t) \geq \sum_{k \geq r} \mathbb{P}(\xi = k) \frac{1 - \frac{1}{r!} k^r t^r}{1-t} = \frac{1 - \frac{t^r}{r!} \mathbb{E}[\xi^r]}{1-t}.$$

Evaluating the function $G_\xi^r(1 - t)$ at $t = t_0 = \left(\frac{(r-1)!}{\mathbb{E}[\xi^r]}\right)^{1/(r-1)}$ yields

$$G_\xi^r(1 - t_0) \geq \frac{1 - \frac{t_0^r}{r!} \mathbb{E}[\xi^r]}{1 - t_0} = \frac{1 - \frac{1}{r} t_0}{1 - t_0}.$$

Since the maximum value of $G_\xi^r(x)$ is at least as big as $G_\xi^r(1 - t_0)$, by equation (1.3),

$$\begin{aligned} p_c(T_\xi, r) &\geq 1 - \frac{1}{G_\xi^r(1 - t_0)} = \frac{G_\xi^r(1 - t_0) - 1}{G_\xi^r(1 - t_0)} \\ &= \frac{t_0 \left(1 - \frac{1}{r}\right)}{1 - t_0} \frac{1 - t_0}{1 - \frac{1}{r} t_0} \\ &= \frac{t_0 \left(1 - \frac{1}{r}\right)}{1 - t_0/r} \geq t_0 \left(1 - \frac{1}{r}\right) \\ &= \left(1 - \frac{1}{r}\right) \left(\frac{(r-1)!}{\mathbb{E}[\xi^r]}\right)^{1/(r-1)}. \end{aligned}$$

This completes the proof of the lemma. □

Theorem 2.1 now follows immediately from Lemmas 2.2 and 2.3.

It is not possible to extend a result of the form of Theorem 2.1 to $\alpha > r - 1$, as demonstrated, again, by the regular b -branching tree. For every α , the $(1+\alpha)$ -th moment of this distribution is $b^{1+\alpha}$ and the critical probability for the constant distribution is $p_c(T_b, r) = (1 + o(1))(1 - 1/r) \left(\frac{(r-1)!}{b^r}\right)^{1/(r-1)}$.

As we already noted, Lemma 2.3 is asymptotically sharp, giving the best possible constant in Theorem 2.1 for any $r \geq 2$ and $\alpha = r - 1$. We now show that for $\alpha \in (0, r - 1)$, Theorem 2.1 is also best possible, up to constants. In [3], it was shown that for every $r \geq 2$, there is a constant C_r such that if $b \geq (r - 1)(\log(4r) + 1)$, then there is an offspring distribution $\eta_{r,b}$ with $\mathbb{E}[\eta_{r,b}] = b$ and $p_c(T_{\eta_{r,b}}, r) \leq C_r \exp\left(-\frac{b}{r-1}\right)$ (see Lemma 3.10 in [3]). In particular, it was shown that there are $k_1 = k_1(r, b) \leq (r - 2) \exp\left(\frac{b}{r-1} + 1\right) - 1$ and $A, \lambda \in (0, 1)$ so that the distribution $\eta_{r,b}$ is given by

$$\mathbb{P}(\eta_{r,b} = k) = \begin{cases} \frac{r-1}{k(k-1)} & r < k \leq k_1, k \neq 2r + 1 \\ \frac{1}{r} + \lambda A & k = r \\ \frac{r-1}{(2r+1)2r} + (1 - \lambda)A & k = 2r + 1. \end{cases}$$

For any $\alpha > 0$, the $(\alpha + 1)$ -th moment of $\eta_{r,b}$ is bounded from above as follows,

$$\begin{aligned} \mathbb{E}[\eta_{r,b}^{\alpha+1}] &= \sum_{k=r}^{k_1} \frac{(r-1)}{k(k-1)} k^{\alpha+1} + \lambda A r^{\alpha+1} + (1 - \lambda)A(2r + 1)^{\alpha+1} \\ &\leq 2(r-1) \sum_{k=r}^{k_1} k^{\alpha-1} + 2(2r + 1)^{\alpha+1} \\ &\leq 2(r-1) \left(\int_r^{k_1+1} x^{\alpha-1} dx + r^{\alpha-1} \right) + 2(2r + 1)^{\alpha+1} \\ &\leq \frac{2(r-1)}{\alpha} (k_1 + 1)^\alpha + 3(2r + 1)^{\alpha+1} \\ &\leq \frac{2(r-1)}{\alpha} \left((r-2) \exp\left(\frac{b}{r-1} + 1\right) \right)^\alpha + 3(2r + 1)^{\alpha+1}, \end{aligned}$$

where the $r^{\alpha-1}$ term makes the inequality hold for $\alpha < 1$. In particular, there is a constant $C_{r,\alpha}$ so that for b sufficiently large, $\mathbb{E}[\eta_{r,b}^{1+\alpha}]^{1/\alpha} \leq C_{r,\alpha} \exp\left(\frac{b}{r-1}\right)$. Thus, for some positive constant $C'_{r,\alpha}$,

$$p_c(T_{\eta_{r,b}}, r) \leq C_r \exp\left(-\frac{b}{r-1}\right) \leq C'_{r,\alpha} \mathbb{E}[\eta_{r,b}^{1+\alpha}]^{-1/\alpha}.$$

Hence the bounds in Theorem 2.1 are sharp up to a constant that does not depend on the offspring distribution ξ .

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