DOMINATION BETWEEN TREES AND APPLICATION TO 
AN EXPLOSION PROBLEM
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ABSTRACT:
We define a notion of stochastic domination between trees, where one tree dominates another if when the vertices of each are labeled with independent, identically distributed random variables, one tree is always more likely to contain a path with a specified property. Sufficient conditions for this kind of domination are (1) more symmetry and (2) earlier branching. We apply these conditions to the problem of determining how fast a tree must grow before first-passage percolation on the tree exhibits an explosion, that is to say, infinitely many vertices are reached in finite time. For a tree in which each vertex at distance $n - 1$ from the root has $f(n)$ offspring, $f$ nondecreasing, an explosion occurs with exponentially distributed passage times if and only if $\sum f(n)^{-1} < \infty$.

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1 Introduction

Let $\Gamma$ be any locally finite tree with some vertex $\rho$ chosen as the root and total height $N \leq \infty$. Label the vertices of $\Gamma$ with independent, identically distributed real random variables $\{X(\sigma) : \sigma \in \Gamma\}$, and let $B \subseteq \mathbb{R}^N$ be some Borel set. Let $P(B; \Gamma)$ denote the probability that $(X(\sigma_1), X(\sigma_2), \ldots) \in B$ for some for some non-self-intersecting path $\rho, v_1, \ldots, v_N$ in $\Gamma$. This probability arises in many contexts. The problem as such is studied in [9] and [7]. For first-passage percolation, the probability of reaching level $N$ by time $T$ is $P(B; \Gamma|_N)$, where $B$ is the set $\sum_{i=1}^{N} v_i \leq T$ and $\Gamma|_N$ is the first $N$ levels of $\Gamma$. The same quantity arises when studying diffusion-limited aggregation on trees via the exponential representation [1, 3]. A random walk in a random environment on a tree will be transient when some path is itself transient, which probability can be reduced via the electrical representation to $P(B; \Gamma)$ for independent, identically distributed resistances, the set $B$ being again the summable sequences; see [11] and [12]. The study of tree-indexed Markov chains [5] can be reduced to computations of $P(B; \Gamma)$ by representing the Markov chain as a function of independent, identically distributed uniform $[0, 1]$ random variables.

Let $|v|$ denote the distance from $\rho$ to $v$ and write $w \leq v$ if $w$ is on the path from $\rho$ to $v$. Let $\Gamma_n = \{v : |v| = n\}$ denote the $n^{th}$ level of $\Gamma$ and $\Gamma|_n$ denote the first $n$ levels of $\Gamma$. Say that $\Gamma$ is 	extit{spherically symmetric} with growth function $f$ if every vertex $v \in \Gamma_n$ has $f(|v| + 1)$ children (a child of $v$ is a neighbor $w$ with $v \leq w$). The notion that we examine in this paper of stochastic domination between trees is as follows.

**Definition 1** Let $\Gamma$ and $\Gamma'$ be two finite or infinite trees with roots $\rho$ and $\rho'$ respectively. We say that $\Gamma$ dominates $\Gamma'$ if whenever

$$\{X(v) : v \in \Gamma\} \cup \{X'(w) : w \in \Gamma'\}$$

is a collection of i.i.d. random variables, $n \geq 1$ is an integer, and $B \subseteq \mathbb{R}^n$ is a Borel
set, then
\[ P(B; \Gamma_n) \geq P(B; \Gamma'_n). \]

Spherically symmetric trees are easier to compute with, which is the principal motivation for developing inequalities that compare \( P(B; \Gamma) \) and \( P(B; \Gamma') \) when one of \( \Gamma \) or \( \Gamma' \) is spherically symmetric. Our main comparison result, proved in Section 2, is the following.

**Theorem 1** Let \( \Gamma \) and \( \Gamma' \) be finite or infinite trees with \( \Gamma \) spherically symmetric. Then \( \Gamma \) dominates \( \Gamma' \) if and only if for every \( n \geq 1 \) the \( n \)th generation
\[ \Gamma_n \overset{\text{def}}{=} \{ v \in \Gamma : |v| = n \} \]
of \( \Gamma \) is at least as big as the \( n \)th generation of \( \Gamma' \).

**Remarks:**

1. We understand the domination partial order completely only for spherically symmetric trees and for trees of height two. In the latter case, it reduces to the classical notion of Hardy majorization (Proposition 6).
2. Comparison results for \( P(B; \Gamma) \) as \( B \) varies may be found in [12].

A consequence of Theorem 1 is that among all trees \( \Gamma \) of height \( n \) having \( |\Gamma_n| = k \), the tree \( T(n, k) \) consisting of \( k \) disjoint paths of length \( n \) joined at the root is maximal in the domination order. If the common law of the \( X(\sigma) \) is \( \mu \) and \( B \subseteq \mathbb{R}^n \), then
\[ 1 - P(B; T(n, k)) = (1 - \mu^n(B))^k, \]
thus for any \( \Gamma \) of height \( n \),
\[ 1 - P(B; \Gamma) \geq (1 - \mu^n(B))^k. \]

The definition of \( P(B; \Gamma) \) extends naturally to any graded graph, this being a finite graph whose vertices are partitioned into levels 1, \ldots, \( n \) with oriented edges allowed only from levels \( i \) to \( i + 1 \), \( i = 1, \ldots, n - 1 \). A natural conjecture is then
Conjecture 2 If $G$ is a graded graph of height $n$, let $K(G)$ be the number of oriented paths that pass through every level of $G$. Let $X(\sigma)$ be independent, identically distributed random variables with common law $\mu$ and $B \subseteq \mathbb{R}^n$. Then

$$1 - \mathbf{P}(B; G) \geq (1 - \mu^n(B))^{K(G)}.$$ 

If $B$ is an upwardly closed set then both the conjecture and the inequality (1) follows easily from the FKG inequality. In the case where $n = 2$ ($G$ is bipartite), Conjecture 2 is due to Sidorenko [15, Conjecture 5.2] in the form of an analytic inequality; Sidorenko has proved several related analytic inequalities on graphs in [13, 14] and [15] including the conjecture itself for the special cases where $G$ is bipartite, acyclic, a single cycle or sufficiently small.

The remainder of the paper is devoted to an in-depth application of this theorem to first-passage percolation, which we now describe. Let $\Gamma$ be a spherically symmetric infinite tree with growth function $f$, and let $\{X(v)\}$ be a collection of independent exponential random variables of mean one. (Our results hold for a much more general class of distributions described later.) Think of $X(v)$ as a transit time across the edge connecting $v$ to its parent. Define

$$S(v) = \sum_{\rho < w \leq v} X(w) \quad \text{and} \quad M_n = \min_{|v|=n} S(v).$$

In the context of first-passage percolation, $S(v)$ is the passage time from the root to the vertex $v$ and $M_n$ is the first passage time to the $n^{th}$ generation of $\Gamma$. Say that an explosion occurs if the increasing sequence $M_n$ is bounded.

An explosion is a tail event, so for a given tree the probability of an explosion is zero or one. It is natural to try to determine when the probability is zero and when it is one. This problem was brought to our attention by Enrique Andjel (personal communication) in reference to uniqueness proofs for particle systems via graphical representations. For
arbitrary trees, such questions are difficult to settle without leaving open some “critical” case (see for example [10]). As observed by Andjel, it is elementary, for the case of spherically symmetric trees, to obtain conditions for explosion or non-explosion which are almost sharp. Indeed, if
\[
\liminf_{n \to \infty} \frac{1}{n!} \prod_{i=1}^{n} f(i) < \infty,
\]
then a simple Borel-Cantelli argument shows there is almost surely no explosion before time one, from which it follows easily that with probability one, no explosion occurs at all; on the other hand, the condition \(\sum f(n)^{-1} < \infty\) implies that the greedy algorithm finds an explosion. Thus in particular there is no explosion when \(f(n) = n\) but there is one when \(f(n) = n \ln(n)^{1+\epsilon}\). We show that the summability condition is sharp, the statement being a little more complicated in the case where \(f\) is not an increasing function.

**Theorem 3** Let \(\Gamma\) be a spherically symmetric tree with growth function \(f\) that is nondecreasing. The probability of an explosion is zero or one according to whether \(\sum_{n=1}^{\infty} f(n)^{-1}\) is infinite or finite. Furthermore, if the sum is infinite and \(f(n)\) is unbounded as well, then \(\frac{M_N^n}{\sum_{n=1}^{N} f(n)^{-1}}\) converges almost surely to \(e^{-1}\).

In most integral tests in probability theory, some regularity condition is imposed. An unusual feature of the problem considered here is that it permits a criterion (Theorem 4) valid for arbitrary growth functions \(f\). The condition in Theorem 3 always suffices for explosion but simple examples \((f(2n-1) = 1, f(2n) = 2^n)\) show it is not necessary. For any \(f : \mathbb{N} \to \mathbb{N}\), define a function \(\tilde{f} : \mathbb{N} \to \mathbb{R}^+\) recursively by
\[
\tilde{f}(n+1) = \sup \left\{ a : a^m \prod_{i=1}^{n} \tilde{f}(i) \leq \prod_{i=1}^{n+m} f(i) \text{ for all } m \geq 1 \right\}.
\] (2)
In particular,
\[
\tilde{f}(1) = \inf_{m \geq 1} \left( \prod_{i=1}^{m} f(i) \right)^{1/m}.
\]
It is easy to see that the function \( \tilde{f} \) is always nondecreasing and coincides with \( f \) when \( f \) is nondecreasing.

**Theorem 4** Let \( \Gamma \) be a spherically symmetric tree with growth function \( f \), labeled as before by independent exponentials \( X(v) \) of mean 1. Then the probability of an explosion is zero or one according to whether \( \sum_{n=1}^{\infty} \tilde{f}(n)^{-1} \) is infinite or finite. Furthermore, if the sum is infinite and \( \tilde{f} \) is unbounded then

\[
\limsup_{N \to \infty} \frac{M_N}{\sum_{n=1}^{N} \tilde{f}(n)^{-1}} = e^{-1}.
\]

**Remarks**

1. It is usually easy to calculate \( \tilde{f} \) from \( f \). Informally, if you graph \( \sum_{j=1}^{n} \ln(f(j)) \) against \( n \), and consider the convex hull of the region above this graph, its boundary is the graph of \( \sum_{j=1}^{n} \ln \tilde{f}(j) \). For instance if the two sequences \( \{f(2n-1)\} \) and \( \{f(2n)\} \) are nondecreasing with \( f(2n-1) < f(2n) \) for all \( n \), then

\[
\tilde{f}(2n-1) = \tilde{f}(2n) = (f(2n-1)f(2n))^{1/2}.
\]

2. Theorem 4 is proved by comparing the tree \( \Gamma \) to a tree with nondecreasing growth function, which is where the application of Theorem 1 is needed.

These theorems are proved in Section 3. Versions where the variables \( \{X(v)\} \) are not exponential are given in Section 4. Results completely analogous to Theorems 3 and 4 hold for distributions \( G \) satisfying a power law near zero, i.e. \( \lim_{t \to 0} G(t)/t^\alpha \) is finite and positive for some \( \alpha > 0 \). Obtaining sharp criteria for explosion to occur that are valid for arbitrary transit-time distributions \( G \) seems more delicate, although we cannot find a \( G \) for which the natural criterion (12) in the last section fails.
2 Domination between trees

Proof of Theorem 1: We start by establishing the theorem under the additional assumption that $\Gamma'$ is spherically symmetric. This is the only case that is used in the proof of Theorem 4.

Let $\mu$ be any probability measure on $\mathbb{R}$ and let $D$ be a Borel set in $\mathbb{R}^n$. If $b_1, \ldots, b_n$ are the cardinalities of the generations of a spherically symmetric tree (i.e. the growth function is $f(i) = b_i/b_{i-1}$), let the vertices index i.i.d. random variables $X(v)$ with common law $\mu$ and let $\Psi(b_1, \ldots, b_n; D)$ denote the probability that all paths $\rho, v_1, \ldots, v_n$ in the tree satisfy $(X(v_1), \ldots, X(v_n)) \in D$. Passing to complements in the definition of domination, we must show that

$$\Psi(b_1, \ldots, b_n; D) \leq \Psi(b'_1, \ldots, b'_n; D)$$

whenever both are defined and $b_i \geq b'_i$ for all $i$.

The key to doing this is the following recursive relation, obtained by conditioning on the variables $X(v)$ for $|v| = 1$:

$$\Psi(b_1, \ldots, b_n; D) = \left[ \int \Psi\left(\frac{b_2}{b_1}, \ldots, \frac{b_n}{b_1}; D/v_1 \right) d\mu(x_1) \right]^{b_1}$$

where for $D \subseteq \mathbb{R}^n$ and $(x_1, \ldots, x_k) \in \mathbb{R}^k$, the notation $D/(x_1, \ldots, x_k)$ is used for the cross-section of $D$ given by

$$\{(x_{k+1}, \ldots, x_n) \in \mathbb{R}^{n-k} : (x_1, \ldots, x_n) \in D\}.$$ 

Observe that the relation (3) together with the initial condition $\Psi(b; D) = \mu(D)^b$ for $D \subseteq \mathbb{R}$ uniquely determines $\Psi$ and in fact remains a valid inductive definition when the arguments $b_i$ are positive reals, not necessarily integral. Call this extension $\Psi$ as well, since it agrees with the old $\Psi$ on integral arguments. We verify by induction on $n$ that for any $D \subseteq \mathbb{R}_n$, the function $\Psi(b_1, \ldots, b_n; D)$ is nonincreasing in each of its $n$ arguments.
This is clear for \( n = 1 \) so fix \( n > 1 \) and observe the fact that \( \Psi \) is nonincreasing in \( b_2, \ldots, b_n \) follows directly from the induction hypothesis and (3). It remains to check that 
\[
\Psi(b_1, \ldots, b_n) \leq \Psi(b'_1, b_2, \ldots, b_n)
\]
when \( b_1 \geq b'_1 \). Rewriting this as
\[
\left[ \int \frac{b_2}{b_1}, \ldots, \frac{b_n}{b_1}; D/x_1 \right] \frac{d\mu(x_1)}{x_1} \leq \left[ \int \frac{b_2}{b'_1}, \ldots, \frac{b_n}{b'_1}; D/x_1 \right] \frac{d\mu(x_1)}{x_1}^{b'_1}
\]
we see it is just Hölder’s inequality \( \int h(x_1) d\mu(x_1) \leq [\int h(x_1)^r d\mu(x_1)]^{1/r} \) with \( r = b_1/b'_1 \)
applied to the function
\[
h(x_1) = \left[ \int \frac{b_2}{b_2}, \ldots, \frac{b_n}{b_2}; D/(x_1, x_2) \right] \frac{d\mu(x_2)}{x_2}^{b_2/b_1}.
\]
(For \( n = 2 \) take \( h(x_1) = \mu(D/x_1)^{b_2/b_1} \).)

This proves Theorem 1 for spherically symmetric trees. To obtain the general case requires the following lemma.

**Lemma 5** For any Borel set \( D \subseteq \mathbb{R}^n \), the function \( \Psi(b_1, \ldots, b_n; D) \) defined by (3) is log-convex on the positive orthant of \( \mathbb{R}^n \).

**Proof:** Assume that \( \mu^n(D) > 0 \), since otherwise \( \Psi(\cdot; D) \) is identically zero. For \( n = 1 \), \( \ln \Psi(\cdot; D) \) is linear. Proceeding by induction, fix \( n > 1 \). The relation (3) shows that 
\[
\ln \Psi(\lambda b_1, \ldots, \lambda b_n; D) = \lambda \ln \Psi(b_1, \ldots, b_n; D).
\]
Hence to prove convexity it suffices to verify that \( \ln \Psi(1, b_2, b_3, \ldots, b_n; D) \) is a convex function of the positive variables \( b_2, \ldots, b_n \). But 
\[
\Psi(1, b_2, \ldots, b_n; D) = \int \Psi(b_2, \ldots, b_n; D/x_1) d\mu(x_1)
\]
and the sum or integral of log-convex functions is log-convex (see for example [2, p. 7-10]). \( \square \)
Proof of Theorem 1 completed: We verify by induction that for any tree $T$ of depth $n$ with generation cardinalities $|T_i| = b_i$ for $i = 1, \ldots, n$ and for any $D \subseteq \mathbb{R}^n$, the probability $\phi(T; D)$ that all paths $\rho, w_1, \ldots, w_n$ in $T$ satisfy $(X(w_1), \ldots, X(w_n)) \in D$ is at least $\Psi(b_1, \ldots, b_n; D)$, where as usual, $X(v)$ are i.i.d. with common law $\mu$ and $\Psi$ is defined by (3).

For every $w$ in the first generation $T_1$ of $T$, let $T(w)$ denote the subtree $\{\rho\} \cup \{u \in T : u \geq w\}$. Let $b_i(w)$ denote the cardinality of the $i^{th}$ generation of $T(w)$. (In particular, $b_1(w) = b_0(w) = 1$.) Fix $D \subseteq \mathbb{R}^n$. By the induction hypothesis

$$\phi(T(w); D) \geq \int \Psi(b_2(w), \ldots, b_n(w); D/x_1) \, d\mu(x_1)$$

$$= \Psi(1, b_2(w), \ldots, b_n(w); D).$$

Therefore, since $\sum_{w \in T_1} b_i(w) = b_i$ for $i \geq 1$, using the log-convexity established in the previous lemma gives

$$\phi(T; D) = \prod_{w \in T_1} \phi(T(w); D)$$

$$\geq \prod_{w \in T_1} \Psi(1, b_2(w), \ldots, b_n(w); D)$$

$$\geq \Psi(1, \frac{b_2}{b_1}, \ldots, \frac{b_n}{b_1}; D)^{b_1}$$

$$= \Psi(b_1, \ldots, b_n; D).$$

\[ \square \]

Lemma 5 yields a simple description of domination between trees of height 2.

Proposition 6 Let $\Gamma$ and $\Gamma'$ be trees of height 2. For each vertex in $\Gamma_1$, count its children and order the numbers so obtained in a decreasing sequence $n_1 \geq n_2 \geq \cdots \geq n_b$. Similarly
obtain a sequence \( n'_1 \geq n'_2 \geq \cdots \geq n'_b \) from \( \Gamma' \), appending zeros to one of the sequences if necessary so that both have the same length \( b \). Then \( \Gamma \) dominates \( \Gamma' \) if and only if
\[
\sum_{i>k} n_i \geq \sum_{i>k} n'_i \quad \text{for every } k \geq 0. 
\]

(4)

Remark

Thinking of the numbers \( n_i \) as a partition of \( \sum n_i = |\Gamma_2| \), this is the order gotten by combining the usual majorization order of partitions (in the reverse direction) with the inclusion order (Young’s lattice).

Proof: We start by showing that the condition (4) implies domination. Assume without loss of generality that
\[
\sum_{i=1}^b n_i = \sum_{i=1}^b n'_i
\]
(5)
since otherwise we could increase \( n'_1 \), thereby obtaining a tree dominating \( \Gamma' \), while condition (4) would remain unaffected.

Conditions (4) and (5) together imply that the vector \((n'_i : i = 1, \ldots, b)\) majorizes the vector \((n_i : i = 1, \ldots, b)\) in the sense that the latter is a convex combination of permutations of the former (see [8, Theorem 47]). To show that \( \Gamma \) dominates \( \Gamma' \) it suffices to verify that for any \( D \subseteq \mathbb{R}^2 \)
\[
\prod_{i=1}^b \Psi(1, n_i; D) \leq \prod_{i=1}^b \Psi(1, n'_i; D).
\]
(6)

Fixing \( D \), let \( h(n_1, \ldots, n_b) \) denote the left-hand side of (6). Clearly \( h \) is invariant under permutations of its arguments. By Lemma 5 it is a product of log-convex functions and hence log-convex. Since \((n_1, \ldots, n_b)\) is a weighted average of permutations of \((n'_1, \ldots, n'_b)\), the inequality (6) follows.

For the converse, assume \( \Gamma \) dominates \( \Gamma' \). Let \( r \geq 1 \). For \( 0 < \epsilon < 1 \), take
\[
D_\epsilon = ([0, \epsilon^r] \times [0, 1]) \cup ([0, 1] \times [0, \epsilon])
\]
If \( \{X(v) : v \in \Gamma\} \) are independent and uniform on \([0,1]\), then the probability that \((X(v_1), X(v_2)) \in D\) for every path \(\rho, v_1, v_2\) lies between \(\prod_{i=1}^{b} \epsilon_{\min(n_i, r)}\) and \(\prod_{i=1}^{b} (2\epsilon)_{\min(n_i, r)}\). When \(\epsilon\) is sufficiently small, the assumption that \(\Gamma\) dominates \(\Gamma'\) forces

\[
\sum_{i=1}^{b} \min(n_i, r) \geq \sum_{i=1}^{b} \min(n_i', r).
\]

Choosing \(r = n_k'\) yields

\[
rk + \sum_{i>k} n_i \geq \sum_{i=1}^{b} \min(n_i', r) = rk + \sum_{i>k} n_i'
\]

proving (4).

We end this section with some remarks and questions about domination.

1. Already for trees of height 3 the domination order is somewhat mysterious. Consider the trees \(\Gamma\) and \(\Gamma'\) in figure 1, where \(\Gamma'\) is obtained from \(\Gamma\) by gluing together the vertices in the first generation.

![Diagram](image.png)

**figure 1**

Intuitively, it seems that \(\Gamma\) should dominate \(\Gamma'\), but this is not the case. Let

\[
D = ([0, 1/2] \times [0, 1] \times [0, 2/3]) \cup ([1/2, 1] \times [0, 1/2] \times [0, 1])
\]

and let \(X(v)\) be uniform on \([0, 1]\). The probability that all paths in \(\Gamma\) have \((X(v_1), X(v_2), X(v_3)) \in D\) is \(\frac{1075}{7776}\), while the corresponding probability for \(\Gamma'\) is only \(\frac{698}{7776}\).

2. To verify that a tree \(\Gamma\) dominates another tree \(\Gamma'\), it suffices to consider the case in which the i.i.d. variables \(X(v)\) are uniform on \([0,1]\), since other variables can be
written as functions of these. Along the same lines, Theorem 1 may be seen to sharpen a result from [4]. There it was shown that any tree of height $n$ with $b_n$ vertices in the $n^{th}$ generation is dominated by the tree consisting of $b_n$ disjoint paths of length $n$ from the root. The notion of domination in [4] is for tree-indexed Markov chains; we omit the easy proof that this is an equivalent definition of domination.

3. A counterexample of the type given in figure 1 cannot occur if the set $D$ is restricted to being an upwardly closed subset of $[0, 1]^n$ (i.e. $x \in D$ and $y \geq x$ coordinatewise imply $y \in D$). These subsets occur naturally in percolation problems. Domination for upwardly closed sets subsumes gluing, as may be shown using the FKG inequality. Is this partial order any more tractable for trees of height greater than 2?

3 Exponential transit times and nondecreasing growth functions

Proof of Theorem 3: When $\sum_{n=1}^{\infty} f(n)^{-1} < \infty$, use the greedy algorithm to select the (a.s. unique) random path $\rho = v_0, v_1, v_2, \ldots$ for which $X(v(n))$ is minimal among $\{X(w) : w$ is a child of $v_{n-1}\}$. Then

$$\mathbf{E}X(v_n) = \mathbf{E}[X(v_n) \mid v_{n-1}] = f(n)^{-1},$$

being the minimum of $f(n)$ standard exponentials. Hence $\mathbf{E}\sum_{n=1}^{\infty} X(v_n) < \infty$ so in particular, $\sum_{n=1}^{\infty} X(v_n) < \infty$ almost surely.

The more interesting case is when $\sum_{n=1}^{\infty} f(n)^{-1} = \infty$. In this case, consider the weighted sums

$$S^*(v) = \sum_{\rho \prec w \leq v} Y(w)$$
where \( Y(w) = f(|w|)X(w) \). Define
\[
M^*_n = \min_{|v|=n} S^*(v).
\]
For any vertex \( w \neq \rho \) and any \( \lambda > 0 \) we have
\[
E \left[ e^{-\lambda Y(w)} \right] = \frac{1}{1 + \lambda f(|w|)}.
\]
Multiplying these along a path yields
\[
E \left[ e^{-\lambda S^*(v)} \right] = \prod_{\rho < w \leq v} \frac{1}{1 + \lambda f(|w|)} < \lambda^{-|v|} \prod_{j=1}^{|v|} f(j)^{-1},
\]
and summing over \( \Gamma_n \) gives \( E \left[ \sum_{v \in \Gamma_n} \exp(-\lambda S^*(v)) \right] < \lambda^{-n} \). Now for any \( \epsilon \) with \( 0 < \epsilon < 1 \), Markov’s inequality implies that
\[
P \left[ \exists v \in \Gamma_n : S^*(v) \leq \frac{n}{\lambda} \ln((1-\epsilon)\lambda) \right] < (1-\epsilon)^n.
\]
By Borel-Cantelli, it follows that \( M^*_n \leq \frac{n}{\lambda} \ln((1-\epsilon)\lambda) \) finitely often almost surely; choosing \( \lambda = e \) and letting \( \epsilon \to 0 \) leads to
\[
\liminf_{n \to \infty} \frac{1}{n} M^*_n \geq e^{-1} \text{ almost surely.} \tag{7}
\]

Deriving a lower estimate for \( M_n^* \) from (7) requires no probability theory. Choose \( \epsilon > 0 \); with probability one there exists an integer \( N_\epsilon \) such that \( M^*_k \geq k(e^{-1} - \epsilon) \) for all \( k \geq N_\epsilon \). For any path \( \{v_k : 0 \leq k \leq n\} \) starting from \( v_0 = \rho \), summation by parts yields
\[
\sum_{k=1}^{n} X(v_k) = \sum_{k=1}^{n} f(k)^{-1}[S^*(v_k) - S^*(v_{k-1})]
\]
\[
= S^*(v_n)f(n+1)^{-1} + \sum_{k=1}^{n} S^*(v_k)[f(k)^{-1} - f(k+1)^{-1}]
\]
\[
> S^*(v_n)f(n+1)^{-1} + (e^{-1} - \epsilon) \sum_{k=1}^{n} k[f(k)^{-1} - f(k+1)^{-1}] - C(N_\epsilon)
\]
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where $C(N)$ depends only on $N$ and $f$. The last inequality is the only place we use the assumption that $f$ is nondecreasing. Summing by parts again,

$$\sum_{k=1}^{n} X(v_k) > (e^{-1} - \epsilon) \sum_{j=1}^{n} f(j)^{-1} - C(N) + f(n+1)^{-1} [S^*(v_n) - n(e^{-1} - \epsilon)]$$

$$\geq (e^{-1} - \epsilon) \sum_{j=1}^{n} f(j)^{-1} - C(N).$$

Thus $M_n$ is also greater than the right-hand side of the last inequality, which easily implies that almost surely

$$\liminf_{n \to \infty} M_n/\sum_{j=1}^{n} f(j)^{-1} \geq e^{-1},$$

as long as the series $\sum f(j)^{-1}$ diverges. This completes the proof of the explosion criterion.

To show that

$$\limsup_{n \to \infty} M_n/\sum_{j=1}^{n} f(j)^{-1} \leq e^{-1}$$

when the denominator tends to infinity and $f$ is unbounded it suffices to exhibit, for every $\epsilon > 0$, an infinite path in $\Gamma$ along which the condition

$$S(v)/\sum_{j=1}^{\lfloor |v| \rfloor} f(j)^{-1} \leq e^{-1} + \epsilon$$

fails only finitely often. This may be accomplished by a branching process argument as in [11]; the reader is referred there for greater detail.

Let $k$ be any positive integer. We prune the tree $\Gamma$ to obtain a random subtree $\Gamma'$ as follows. First, take $\rho \in \Gamma'$; next, for $v \in \Gamma'$ and $w$ a child of $v$ in $\Gamma$, let $w \in \Gamma'$ if and only if $X(w)$ has one of the $k$ least values among $\{X(u) : u \text{ is a child of } v \text{ in } \Gamma\}$. Then $\Gamma'$ is a tree which is eventually $k$-ary. Again, let $Y(v) = f(|v|)X(v)$ with partial sums $S^*(v) = \sum_{\rho < w \leq v} Y(w)$ and $M_n^*$ as before. For $v \in \Gamma'$ the variables $\{Y(w) : w \in \Gamma' \text{ is a child of } v\}$ are the first $k$ order statistics of $f(|v|)$ independent exponentials of mean $f(|v|)$. As
$|v| \to \infty$, we have assumed that also $f(|v|) \to \infty$, so these order statistics are converging weakly to the first $k$ hits of a mean 1 Poisson process. [This is an easy consequence of the usual Poisson convergence theorem and the fact that the common distribution of the original random variables $X(v)$ has a density of one at the origin.] In particular, for every $\epsilon > 0$ and $v \in \Gamma'$, the joint distribution of

$$\{Y(w) : w \in \Gamma', w \text{ is a child of } v\}$$

is stochastically dominated by the first $k$ hits of a mean $1 - \epsilon$ Poisson process. In other words, conditional on the selection of the vertices belonging to the subtree $\Gamma'$, we can couple the process $\{Y(v) : v \in \Gamma'\}$ to a process $\{Z(v) : v \in \Gamma'\}$ for which the values of $Z$ over children of $v$ are independent for different $v$ and each distributed as a uniform random permutation of the first $k$ order statistics of a mean $1 - \epsilon$ Poisson process whenever $f(|v| + 1) \geq k$; the coupling will satisfy $Z(v) > Y(v)$ for all $v$ such that $|v| > R$ for some $R$ depending on $k$ and $\epsilon$.

If $Z$ is a random variable distributed as a uniform random selection of one of the first $k$ hits of a mean $1 - \epsilon$ Poisson process, then

$$E e^{-\lambda Z} = k^{-1} \sum_{i=1}^{k} \left( 1 + \frac{\lambda}{1 - \epsilon} \right)^{-i} = \frac{1 - \epsilon}{k \lambda} \left[ 1 - \left( 1 + \frac{\lambda}{1 - \epsilon} \right)^{-k} \right].$$

Choose $k = k(\epsilon)$ so that this is at least $(1 - 2\epsilon)/k\lambda$ for all $\lambda > 1/10$, say. The rate function for $Z$ satisfies

$$m_Z(a) \overset{\text{def}}{=} \inf_{\lambda > 0} e^{\lambda a} E e^{-\lambda Z}$$

$$\geq \frac{1 - 2\epsilon}{k} \inf_{\lambda > 0} e^{\lambda a} \min \left\{ \frac{1}{\lambda}, 10 \right\}$$

$$\geq \frac{1 - 2\epsilon}{k} e a$$

for $0 < a < 10/e$. In particular if we fix $a = (1 - 3\epsilon)^{-1} e^{-1}$ then $m_Z(a) > k^{-1}$.
Now the proof of (2.8) in [11, p. 1237] or the main result of [6] shows that there is some integer $L$ for which the following branching process is supercritical: $v \in \Gamma'$ begets $w \in \Gamma'$ if $|w| = |v| + L$ and $L^{-1} \sum_{v < u \leq w} Z(u) \leq a$. This implies that for such an $L$ there exists almost surely a path $\rho, v_1, v_2, \ldots$ for which

$$\frac{1}{L} \sum_{i = jL+1}^{jL+L} Y(v_i) \leq a$$

for all but finitely many $j$. (9)

From (9) we infer that

$$\sum_{i = jL+1}^{jL+L} X(v_i) \leq aL f(jL + 1)^{-1} \leq a \sum_{i = jL-L+1}^{jL} f(i)^{-1}$$

for all but finitely many $j$, and hence

$$S(v_n) \leq a \sum_{i=1}^{n} f(i)^{-1} + O(1).$$

Since $a = (1 - 3\epsilon)^{-1} e^{-1}$, letting $\epsilon \to 0$ finishes the proof. \qed

4 Exponential transit times, arbitrary growth functions

Proof of Theorem 4: Consider first the case $\sum f(n)^{-1} = \infty$. If $\liminf_{n \to \infty} (\prod_{i=1}^{n} f(i))^{1/n} < \infty$ then clearly there is no explosion since utilizing the gamma distribution shows that $P(M_n < C) \leq \frac{C^n}{n!} \prod_{i=1}^{n} f(i)$. Henceforth we assume that

$$\liminf_{n \to \infty} (\prod_{i=1}^{n} f(i))^{1/n} = \infty,$$

which is equivalent to $\hat{f}(n) \to \infty$. Under this assumption, for every $n \geq 1$ there exists an $m \geq 1$ such that

$$\hat{f}(n+1)^m \prod_{i=1}^{n} \hat{f}(i) = \prod_{i=1}^{n+m} f(i).$$
Consequently, there is an increasing sequence \( \{ n(k) : k = 1, 2, 3, \ldots \} \) for which

\[
\prod_{i=1}^{n(k)} \hat{f}(i) = \prod_{i=1}^{n(k)} f(i)
\]

for all \( k \).

To establish the lower bound on \( \limsup M_n / \sum_{i=1}^{n} \hat{f}(i)^{-1} \), and hence also that no explosion occurs, use the proof of Theorem 3 but with \( Y(v) = \hat{f}(|v|)X(v) \). As in the proof of that theorem, the partial sums \( S^*(v) = \sum_{\rho<w \leq v} Y(w) \) satisfy

\[
E \left[ e^{-\lambda S^*(v)} \right] \leq \lambda^{-|v|} \prod_{i=1}^{\hat{f}(i)}
\]

for \( \lambda > 0 \). When \( |v| = n(k) \), we can substitute \( f \) for \( \hat{f} \) on the right-hand side. It follows for \( \epsilon > 0 \), that with probability one the inequality

\[
M^*_{n(k)} \leq \frac{n(k)}{\lambda} \ln((1-\epsilon)\lambda)
\]

holds for only finitely many \( k \). Taking \( \lambda = e \) we get, as before,

\[
\liminf_{k \to \infty} M_{n(k)} / \sum_{i=1}^{n(k)} \hat{f}(i)^{-1} \geq e^{-1}
\]

so in particular

\[
\limsup_{n \to \infty} M_n / \sum_{i=1}^{n} \hat{f}(i)^{-1} \geq e^{-1}.
\]

Next, let us bound this \( \limsup \) from above, still assuming that \( \sum \hat{f}(n)^{-1} = \infty \) and \( \hat{f}(n) \to \infty \). Let \( g(n) \) denote the integer part of \( \hat{f}(n) \) and let \( \Gamma' \) be a spherically symmetric tree with growth function \( g \). And independent, identically distributed random variables \( \{X(w)\} \). Apply Theorem 3 to \( \Gamma' \). Examining the proof at (8), we see that for any \( \epsilon > 0 \) there exists almost surely some path \( \rho, v_1, v_2, \ldots \) satisfying

\[
X(v_n) \leq (e^{-1} + \epsilon) \sum_{i=1}^{n} g(i)^{-1}
\]

(10)
for all but finitely many \( n \). From Theorem 1 it follows that a path satisfying (10) exists almost surely in \( \Gamma \) as well as \( \Gamma' \). Since \( \sum_{i=1}^{n} g(i)^{-1}/\sum_{i=1}^{n} \tilde{f}(i)^{-1} \rightarrow 1 \), this implies that almost surely \[
\limsup_{n \to \infty} M_n / \sum_{i=1}^{n} \tilde{f}(i)^{-1} \leq e^{-1}.
\]

Finally, in the case when \( \sum \tilde{f}(n)^{-1} \) converges, define \( g \) and \( \Gamma' \) as above and apply Theorem 3 to conclude that for some fixed \( L > 0 \)
\[
\liminf_{n \to \infty} P[\exists w \in \Gamma'_n : S'(w) \leq L] > 0.
\]
This together with Theorem 1 and the zero-one law for explosions proves that the probability of an explosion is 1.

\[\square\]

5 Other transit-time distributions

Theorems 3 and 4 hold without change when the i.i.d. transit times \( X(v) \) have any distribution function \( G \) for which \( G(t)/t = 1 + o(1) \) as \( t \to 0 \). More generally we have

**Proposition 7** Let \( \Gamma \) be an infinite spherically symmetric tree with growth function \( f \). Suppose \( \{X(v) : v \in \Gamma\} \) are i.i.d. with their common distribution function \( G \) satisfying \[
\lim_{t \to 0} G(t)t^{-\alpha} = c > 0
\]
for some \( \alpha > 0 \). With \( S(v) \) and \( M_n \) defined as throughout, we then have the following analogues of Theorems 3 and 4.

(i) If \( f \) is nondecreasing then there is an explosion (sup \( M_n < \infty \)) if and only if \( \sum f(n)^{-1/\alpha} \) converges. If the sum diverges then
\[
\lim_{n \to \infty} M_n / \sum_{j=1}^{n} f(j)^{-1/\alpha} = \alpha e^{-1}[c\Gamma(1 + \alpha)]^{-1/\alpha}.
\]
(ii) For a general growth function $f$, define $\tilde{f}$ as in the preface to Theorem 4. There is almost surely an explosion if and only if $\sum \tilde{f}(n)^{-1/\alpha}$ converges. If the sum diverges then with probability one,

$$\limsup_{n \to \infty} M_n/\sum_{j=1}^{n} \tilde{f}(j)^{-1/\alpha} = \alpha e^{-1} [c \Gamma(1 + \alpha)]^{-1/\alpha}. $$

**Proof:** We only discuss the modifications needed in the proof for the exponential case. For $(i)$, first assume that $\sum f(n)^{-1/\alpha}$ converges. For every vertex $v \in \Gamma_{n-1}$, the number of its children $w \in \Gamma_n$ for which

$$X(w) \leq \left[\frac{c}{2} f(n)\right]^{-1/\alpha}$$

converges in distribution to a Poisson with mean 2. Comparing to a Galton-Watson process shows that (11) holds along an entire infinite path with positive probability and this implies almost sure explosion.

Next, assume that $\sum f(n)^{-1/\alpha} = \infty$ and that $f(n) \to \infty$. Given $\epsilon > 0$, choose $\delta > 0$ so that $G(t) < (c + \epsilon)t^\alpha$ for $0 < t < \delta$. Estimate the moment generating function

$$E[e^{-\lambda X(v)}] = \int_0^\infty e^{-\lambda t} dG(t)$$

$$= \int_0^\infty G(t)\lambda e^{-\lambda t} dt$$

$$\leq \int_0^\delta (c + \epsilon)t^\alpha \lambda e^{-\lambda t} dt + e^{-\lambda \delta}$$

$$\leq (c + \epsilon)\lambda^{-\alpha} \Gamma(1 + \alpha) + e^{-\lambda \delta}$$

$$\leq (c + 2\epsilon)\lambda^{-\alpha} \Gamma(1 + \alpha)$$
for large positive $\lambda$. Define $S^*(v) = \sum_{\rho < w \leq v} f(|w|^{1/\alpha} X(w)$ and $M_n^* = \min_{|v| = n} S^*(v)$. Since $f(n) \to \infty$ it follows that

$$\mathbb{E}[\exp(-\lambda S^*(v))] \leq A(\lambda) \prod_{j=1}^{|v|} (c + 2\epsilon)\lambda^{-\alpha} f(j)^{-1} \Gamma(1 + \alpha)$$

where $A(\lambda) > 0$, and hence

$$\mathbb{E} \left[ \sum_{|v| = n} \exp(-\lambda S^*(v)) \right] \leq A(\lambda) \left[ (c + 2\epsilon)\lambda^{-\alpha} \Gamma(1 + \alpha) \right]^{n}.$$  

As in the proof of Theorem 3, optimizing over $\lambda$ yields

$$\liminf \frac{1}{n} M_n^* \geq e^{-1} \alpha [c \Gamma(1 + \alpha)]^{-1/\alpha}$$

and consequently, with probability one,

$$\liminf \frac{M_n}{\sum_{j=1}^n f(j)^{-1/\alpha}} \geq e^{-1} [c \Gamma(1 + \alpha)]^{-1/\alpha}.$$ 

The rest of the proof proceeds as in Theorem 3.

Finally, the only change in the proof of (ii) is to note that the operation $f \mapsto \tilde{f}$ commutes with taking powers.

**Question:** is there a simple explosion criterion for arbitrary transit time distributions? At least when the growth function $f$ is nondecreasing and $G$ is strictly monotone and continuous, it seems possible that an explosion occurs if and only if

$$\sum_{n=1}^{\infty} G^{-1} \left( \frac{1}{f(n)} \right) < \infty$$ (12)

where $G^{-1}$ is the inverse function to $G$. The technique used to obtain Proposition 7 is powerful enough to verify this criterion for a slightly more general class of distributions, but the general case has eluded us.
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