

ON RANDOM MINIMUM LENGTH SPANNING TREES

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We extend and strengthen the result that, in the complete graph K_n with independent random edge-lengths uniformly distributed on $[0, 1]$, the expected length of the minimum spanning tree tends to $\zeta(3)$ as $n \rightarrow \infty$. In particular, if K_n is replaced by the complete bipartite graph $K_{n,n}$ then there is a corresponding limit of $2\zeta(3)$.

1. Introduction

Suppose that we are given a complete graph K_n on n vertices together with lengths on the edges which are independent identically distributed non-negative random variables. Suppose that their common distribution function F satisfies $F(0)=0$, F is differentiable from the right at zero and $D=F'_+(0)>0$. Let X denote a random variable with this distribution.

Let L_n denote the (random) length of the minimum spanning tree in this graph. Frieze [3] proved the following:

Theorem 1.

(a) If $E(X) < \infty$ then $\lim_{n \rightarrow \infty} E(L_n) = \zeta(3)/D$, where

$$\zeta(3) = \sum_{k=1}^{\infty} k^{-3} = 1.202\dots$$

(b) If $E(X^2) < \infty$ then $\lim_{n \rightarrow \infty} \text{Var}(L_n) = 0$, and so in particular $L_n \rightarrow \zeta(3)/D$ in probability. ■

Recently, Steele [5] has shown that the convergence in probability above holds without assumptions on moments.

In this paper we generalise Theorem 1 to graphs other than K_n . We shall also simplify the proofs and sharpen the results.

Let H be a fixed connected multigraph, with vertex set $V(H) = \{v_1, v_2, \dots, v_h\}$. Corresponding to each edge e of H let F_e be a distribution function of a non-negative

random variable such that $F_e(0)=0$ and F_e has a right derivative D_e at 0. We assume that there exists $D>0$ such that for each vertex v of H ,

$$\sum_{v \in e} D_e = D.$$

(Observe that loops contribute *once* to this sum.)

For each $n=1, 2, \dots$ let H_n be a (loopless) graph obtained as follows. Replace each vertex v_i of H by a set V_i of n new vertices, so that $|V(H_n)|=nh$. Now join two distinct vertices of H_n by the same number of edges as join the corresponding vertices of H . Thus if H has λ loops and ν non-loops then H_n has $\mu = \binom{n}{2} \lambda + n^2 \nu$ edges.

Let the edges of H_n have independent lengths, where the length of an edge e is distributed according to the distribution for the edge of H from which e arose. Let us extend our notation so that the length of $e \in E(H_n)$ has distribution function F_e as well.

For any connected graph G with non-negative edge-lengths let $L(G)$ denote the length of a minimum spanning tree in G .

Theorem 2. *As $n \rightarrow \infty$, $L(H_n) \rightarrow (h/D)\zeta(3)$ a.s.*

This result follows (by a Borel—Cantelli lemma) from

Lemma 0. *For any $\varepsilon > 0$ there exists c , $0 < c < 1$ such that*

$$P(|L(H_n) - (h/D)\zeta(3)| > \varepsilon) < c^{n^{1/4}}.$$

Theorem 1 follows from the case where H has a single vertex and a single loop, so that $H_n = K_n$. Some other interesting cases are the following, where for simplicity we make each edge length uniform on $[0, 1]$.

$$(1) \quad L((K_r)_n) \rightarrow \frac{r}{r-1} \zeta(3) \quad \text{a.s.}$$

(Here $(K_r)_n$ is the complete multipartite graph with r blocks each of size n .) In particular $L(K_{n,n}) \rightarrow 2\zeta(3)$ (see [4]).

$$(2) \quad L((C_k)_n) \rightarrow \frac{k}{2} \zeta(3) \quad \text{a.s.}$$

(Here C_k is a cycle with k vertices.)

$$(3) \quad L((Q_k)_n) \rightarrow \frac{2^k}{k} \zeta(3) \quad \text{a.s.}$$

(Here Q_k is the k -cube.)

We shall prove lemma 0 (and thus Theorem 2) in three stages (sections 3, 4, 5 below), but first we have:

2. Notation and Preliminaries

We use two models of random subgraph of H_n .

For $1 \leq m \leq \mu$ $H_{n,m}$ has the same vertex set as H_n and for its edge set a random m -edge subset of $E(H_n)$.

For $0 \leq p \leq 1$ $H_{n,p}$ has the same vertex set as H_n and each of the μ edges of H_n are independently included with probability p and excluded with probability $1-p$.

We have need of the following simple relation between $H_{n,m}$ and $H_{n,p}$ where $p = \frac{m}{\mu}$: for any property Π

$$(4) \quad P(H_{n,m} \in \Pi) \cong 2\sqrt{\mu} P(H_{n,p} \in \Pi).$$

This follows from

$$P(H_{n,p} \in \Pi) = \sum_{m'=0}^{\mu} P(H_{n,p} \in \Pi | |E(H_{n,p})| = m') P(|E(H_{n,p})| = m')$$

and the fact that (i) $H_{n,p}$ conditional on $|E(H_{n,p})| = m'$ is distributed as $H_{n,m'}$ and (ii) $|E(H_{n,p})|$ has the binomial distribution $B(\mu, p)$.

3. Expected value for uniform [0, 1] case

Our approach to proving theorem 2 is similar to that of [3] but uses martingale inequalities in place of the Chebycheff inequality. We first discuss the case where edge lengths are uniform on $[0, 1]$ and H is r -regular (with loops counting once towards the degree of a node).

Suppose that the edges $E(H_n) = \{u_1, u_2, \dots, u_{\mu}\}$ are numbered so that $l(u_i) \leq l(u_{i+1})$, $i=1, 2, \dots, \mu-1$ where $l(u)$ is the length of edge u .

A minimum length tree may be constructed using the Greedy Algorithm of Kruskal [4]. Let $F_0 = \varnothing$, $F_1 = \{u_1\}$, F_2, \dots, F_{h_n-1} be the sequence of edge sets of the successive forests produced. Here $|F_i| = i$ and F_{h_n-1} is the set of edges in a minimum spanning tree.

Next define $t_i = \max \{j: u_j \in F_i\}$. Then

$$(5) \quad L(H_n) = \sum_{i=1}^{h_n-1} l(u_{t_i}),$$

and thus

$$(6) \quad E(L(H_n)) = \frac{1}{\mu+1} \cdot E\left(\sum_{i=1}^{h_n-1} t_i\right).$$

The subgraph Γ_m of H_n induced by $U_m = \{u_1, u_2, \dots, u_m\}$ is distributed as $H_{n,m}$. Let \varkappa_m denote the number of connected components of Γ_m .

Lemma 1.

$$\sum_{i=1}^{h_n-1} t_i = \sum_{m=1}^{\mu} \varkappa_m + h_n - \mu - 1.$$

Proof.

$$\sum_{m=1}^{\mu} \varkappa_m = \sum_{r=1}^{hn-1} (hn-r)(t_{r+1}-t_r)$$

where $t_{hn} = \mu + 1$. This is because $\Gamma_r, \Gamma_{r+1}, \dots, h_{r+1-1}$ all have $hn-r$ components. Thus

$$\sum_{m=1}^{\mu} \varkappa_m = -(hn-1)t_1 + t_2 + t_3 + \dots + t_{hn-1} + t_{hn},$$

and the result follows on noting that $t_1=1$ and $t_{hn}=\mu+1$. ■

It follows from (6) and the above lemma that

$$(7) \quad E(L(H_n)) = \frac{1}{\mu+1} (E(\sum_{m=1}^{\mu} \varkappa_m) + hn) - 1.$$

We must therefore estimate $E(\sum_{m=1}^{\mu} \varkappa_m)$. It will be easier to work with $H_{n,p}$ and so let \varkappa_p denote the (random) number of components in $H_{n,p}$. The following simplification is from Bollobás and Simon [1].

Lemma 2.

$$\frac{1}{\mu+1} E(\sum_{m=1}^{\mu} \varkappa_m) = \int_0^1 E(\varkappa_p) dp.$$

Proof.

$$\int_0^1 E(\varkappa_p) dp = \int_0^1 \sum_{m=0}^{\mu} \binom{\mu}{m} p^m (1-p)^{\mu-m} E(\varkappa_m) dp = \sum_{m=0}^{\mu} E(\varkappa_m) \binom{\mu}{m} \frac{m!(\mu-m)!}{(\mu+1)!}. \quad \blacksquare$$

Thus to compute $E(L(H_n))$ we need an accurate estimate of $E(\varkappa_p)$.

Lemma 3. *If $p \leq 4 \log n/n$ then*

$$(8) \quad E(\varkappa_p) = hn\varphi(rnp) + o(n^{3/4})$$

where

$$\varphi(a) = \sum_{s=1}^{\infty} \frac{s^{s-2}}{s!} a^{s-1} e^{-as}.$$

(The 'little o' notation in (8) is intended to imply uniformity over relevant p .)

Proof. As we shall see, the most important components from our point of view are small isolated trees. Let therefore τ_p denote the number of components in $H_{n,p}$ which are trees of order $n^{1/3}$ or less. Let $\mathcal{T}_s(G)$ denote the set of s -vertex subtrees of a graph G . For $T \in \mathcal{T}_s(H_n)$ we find

$$P(T \text{ is a component of } H_{n,p}) = p^{s-1}(1-p)^{rns-\alpha(T)}$$

where, rather crudely,

$$0 \leq \alpha(T) \leq r \binom{s}{2} + r.$$

Hence

$$E(\tau_p) = \sum_{s=1}^{n^{1/3}} \sum_{T \in \mathcal{T}_s(H_n)} p^{s-1} (1-p)^{rns-\alpha(T)} = \\ = (1 + o(n^{-1/4})) \sum_{s=1}^{n^{1/3}} |\mathcal{T}_s(H_n)| p^{s-1} e^{-rns}.$$

We must now estimate $|\mathcal{T}_s(H_n)|$.

For each tree T in $\mathcal{T}_s(K_s)$ and each tree T' in $\mathcal{T}_s(H_n)$ let $\mathcal{F}(T, T')$ be the set of bijections f between $E(T)$ and $E(T')$ that correspond to bijections between $V(T)$ and $V(T')$.

Now if $T' \in \mathcal{T}(H_n)$ then

$$\sum_{T \in \mathcal{T}_s(K_s)} |\mathcal{F}(T, T')| = s!$$

since each bijection between $\{1, \dots, s\}$ and $V(T')$ contributes exactly one to the sum on the left hand side. Hence

$$(10) \quad |\mathcal{T}_s(H_n)| = \frac{1}{s!} \sum_{T \in \mathcal{T}_s(K_s)} \sum_{T' \in \mathcal{T}_s(H_n)} |\mathcal{F}(T, T')|.$$

We shall show that for each $T \in \mathcal{T}_s(K_s)$

$$(11) \quad hn \prod_{k=1}^{s-1} r(n-k) \cong \sum_{T' \in \mathcal{T}_s(H_n)} |\mathcal{F}(T, T')| \cong hn \prod_{k=2}^{s-1} rn.$$

Using (11) in (10) and $|\mathcal{T}_s(K_s)| = s^{s-2}$ yields

$$|\mathcal{T}_s(H_n)| = (1 + o(n^{-1/4})) \frac{s^{s-2}}{s!} hr^{s-1} n^s,$$

and then from (9)

$$(12) \quad E(\tau_p) = (1 + o(n^{-1/4})) hn \sum_{s=1}^{n^{1/3}} \frac{s^{s-2}}{s!} (nrp)^{s-1} e^{-rns}.$$

To prove (11) note that when $s=1$ it is correct (if we interpret $\prod_{k=1}^0$ as 1).

Assume that it is true for some $s \geq 1$: we shall show that it is true for $s+1$. Consider a tree T in $\mathcal{T}_{s+1}(K_{s+1})$ and assume without loss of generality that $s+1$ is a leaf of T , with incident edge e . Then having fixed a bijection f on the tree $T-(s+1)$ in $\mathcal{T}_s(K_s)$ there are between $r(n-s)$ and rn choices for the image of e . This completes our proof of (11) and thus of (12).

We observe that since $s! \cong (s/e)^s$

$$\frac{s^{s-2}}{s!} (nrp)^{s-1} e^{-rns} \cong \frac{e}{s^2} (nrpe^{1-rnp})^{s-1} \cong \frac{e}{s^2}.$$

This implies, from (12), that

$$(13) \quad E(\tau_p) = hn\phi(rnp) + o(n^{3/4}).$$

We now look at σ_p = the number of non-tree components of $H_{n,p}$ of order at most $n^{1/3}$. As each such component consists of a tree $T \in \mathcal{T}_s(H_n)$ plus some k extra edges, we deduce that

$$(14) \quad E(\sigma_p) \cong \sum_{s=1}^{n^{1/3}} \sum_{T \in \mathcal{T}_s(H_n)} p^{s-1} (1-p)^{rns-\alpha(T)} \sum_{k=1}^{r\binom{s}{2}-s+1} \binom{s}{k} p^k (1-p)^{-k} =$$

$$= E(\tau_p) \times o(n^{-1/4}).$$

As $H_{n,p}$ contains at most $n^{2/3}$ components of size exceeding $n^{1/3}$, the lemma follows from (13) and (14). ■

For $p \cong 4 \log n/n$ we use the following.

Lemma 4.

(a) If $p = 4 \log n/n$ then

$$P(H_{n,p} \text{ is not connected}) = O(n^{-3}).$$

(b) If $p = n^{-3/4}$ then

$$P(H_{n,p} \text{ is not connected}) = O(ne^{-n^{1/4}}).$$

Proof.

(a) If $H_{n,p}$ is not connected then either

(i) $h=1$

or

(ii) there is a pair of distinct adjacent vertices v_i, v_j in H such that the subgraph of $H_{n,p}$ induced by $V_i \cup V_j$ is not connected.

In case (i) $H_{n,p}$ is the standard model $G_{n,p}$ and in case (ii) the subgraph K induced by $V_i \cup V_j$ contains a random bipartite graph. For brevity we deal with case (ii) and leave case (i) to the reader. Both cases are straightforward.

If K is not connected then there exist $S \subseteq V_i, T \subseteq V_j$ such that $1 \cong |S| + |T| \cong n$ and no edge of $H_{n,p}$ joins $S \cup T$ to $V_i \cup V_j - S \cup T$. Hence

$$P(\text{ii}) \cong \binom{h}{2} \sum_{\substack{k,l=0 \\ 1 \cong k+l \cong n}}^n u(k, l)$$

where

$$u(k, l) = \binom{n}{k} \binom{n}{l} (1-p)^{k(n-l)+l(n-k)} \cong$$

$$\cong n^{k+l-4(k+l)+(8kl/n)} \cong$$

$$\cong n^{-(3-2(k+l)/n)(k+l)}.$$

Part (a) now follows easily, and part (b) may be proved in a similar manner. ■

We can now obtain the limiting value for $E(L(H_n))$ in the special case under consideration.

Lemma 5. *If H is r -regular and edge-lengths are independent and all uniform on $[0, 1]$ then*

$$\lim_{n \rightarrow \infty} E(L(H_n)) = (h/r)\zeta(3).$$

Proof. It follows from (7) and Lemma 2 that

$$E(L(H_n)) = \int_0^1 (E(x_p) - 1) dp + \frac{hn}{\mu + 1}.$$

Now if $p_0 = 4 \log n/n$ then by Lemma 3,

$$\begin{aligned} \int_0^{p_0} E(x_p) dp &= hn \sum_{s=1}^{\infty} \frac{s^{s-2}}{s!} \int_0^{p_0} (rnp)^{s-1} e^{-rnp s} dp + o(n^{3/4} p_0) = \\ &= (h/r) \sum_{s=1}^{\infty} \frac{s^{s-2}}{s!} \int_0^{4r \log n} x^{s-1} e^{-sx} dx + o(\log n/n^{1/4}) = \\ &= (h/r)\zeta(3) + o(\log n/n^{1/4}). \end{aligned}$$

To see the last equation above note that

$$\int_{\omega}^{\infty} x^{s-1} e^{-sx} dx = 0(e^{-\omega/2}) \quad \text{if } \omega = \omega(n) \rightarrow \infty$$

and

$$\int_0^{\infty} x^{s-1} e^{-sx} dx = (s-1)!/s^s.$$

It follows from Lemma 4(a) that for $p \geq p_0$, $E(x_p) = 1 + O(n^{-2})$ and so

$$\int_{p_0}^1 (E(x_p) - 1) dp = O(n^{-2}). \quad \text{Hence}$$

$$(15) \quad E(L(H_n)) = (h/r)\zeta(3) + o(\log n/n^{1/4}). \quad \blacksquare$$

4. Probability inequality for uniform $[0, 1]$ case

Our aim next is to prove that there is a constant $A = A(r, h) > 0$ such that for any $0 < \varepsilon < 2h/r$

$$(16) \quad P(|L(H_n) - (h/r)\zeta(3)| \geq \varepsilon) \leq e^{-A\varepsilon^2 n^{1/4}}$$

for n sufficiently large. We do this in two stages.

Lemma 6. *Let $t_1, t_2, \dots, t_{hn-1}$ be as in (5) and $0 < \varepsilon < 1$ be fixed. Then for n sufficiently large*

$$P\left(\left|\sum_{i=1}^{hn-1} t_i - (h/r)(\mu+1)\zeta(3)\right| \geq \varepsilon n^2\right) \leq e^{-\varepsilon^2 n^{1/4}/r^3 h^3}.$$

Proof. We prove this using a martingale inequality. Let X_1, X_2, \dots, X_N be random variables, and for each $i=1, \dots, N$ let $X^{(i)}$ denote (X_1, X_2, \dots, X_i) . Suppose that the random variable Z is determined by $X^{(N)}$. For each $i=1, 2, \dots, N$ let

$$(17) \quad \delta_i = \sup |E(Z|X^{(i-1)}) - E(Z|X^{(i)})|.$$

Here $E(Z|X^{(0)})$ means just $E(Z)$. The following inequality is a special case of a martingale inequality due to Azuma (see e.g. Stout [6]). For any $u \geq 0$

$$(18) \quad \Pr(|Z - E(Z)| \geq u) \leq 2 \exp \left\{ -u^2/2 \sum_{i=1}^N \delta_i^2 \right\}.$$

To apply (18) we take $N = \lceil \mu/n^{3/4} \rceil$ and let $X_i = u_i$, the i^{th} shortest edge of H_n . Let $Z = \sum_{m=1}^N x_m$. It is not difficult to see that for δ_i as defined by (17) we have $\delta_i \leq N - i + 1$. This follows from the fact (in an obvious notation) that $|x_m(X^{(N)}) - x_m(Y^{(N)})| \leq 1$ if there exists k such that $X_i = Y_i$ for $i \neq k$ or there exist k, l such that $X_k = Y_l$, $X_l = Y_k$ and $X_i = Y_i$ otherwise.

Thus

$$(19) \quad P(|Z - E(Z)| \geq u) \leq 2e^{-3u^2/N(N+1)(2N+1)} \quad \text{for } u \geq 0.$$

Now let $Z' = \sum_{m=N+1}^{\mu} x_m$. It follows from (4) and Lemma 4(b) that

$$(20) \quad P(Z' \neq \mu - N) = O(n^2 e^{-n^{1/4}})$$

and so

$$(21) \quad E(Z') = \mu - N + o(1).$$

Now (7), (15) and (21) imply that

$$E(Z) = (h/r)(\mu+1)\zeta(3) + O(n^{7/4} \log n).$$

We can then use (19) with $u = \frac{1}{2} \varepsilon n^2$ together with Lemma 1, (20) and $\mu \leq \frac{1}{2} r h n^2$ to obtain the Lemma. ■

We must now show that sums of order statistics of a large number of independent uniform $[0, 1]$ random variables usually behave as expected.

Lemma 7. Let u_i , $i=1, 2, \dots, \mu$ denote the order statistics of μ independent uniform $[0, 1]$ random variables. Let $1 \leq t_1 < t_2 < \dots < t_{h-1} \leq \mu$ and $T = \sum_{k=1}^{h-1} t_k$. Then for any fixed $0 < \varepsilon < 1$

$$(22) \quad P \left(\left| \sum_{k=1}^{h-1} u_{t_k} - \frac{T}{\mu+1} \right| > \frac{\varepsilon T}{\mu+1} \right) \leq e^{-(\varepsilon^2 T/16 h n)}.$$

Proof. It is well known (see for example Feller [2]) that if $X_1, X_2, \dots, X_{\mu+1}$ are independent exponential random variables with mean 1 then the variables $Z_i = \frac{Y_i}{Y_{\mu+1}}$,

$i=1, 2, \dots, \mu$ are distributed as u_i , $i=1, 2, \dots, \mu$ where $Y_i = X_1 + X_2 + \dots + X_i$. It suffices therefore to prove (22) with u_{t_k} replaced by Z_{t_k} . Note now that

$$S = \sum_{k=1}^{hn-1} Y_{t_k} = \sum_{j=1}^{\mu+1} a_j X_j$$

where $a_j = |\{k: t_k \cong j\}|$, and that $T = \sum_{j=1}^{\mu+1} a_j$. Now for $\lambda > 0$

$$\begin{aligned} P(S \cong (1+\varepsilon)T) &= P(e^{\lambda S - \lambda(1+\varepsilon)T} \cong 1) \cong \\ &\cong E(e^{\lambda S - \lambda(1+\varepsilon)T}) \\ &= \prod_{j=1}^{\mu+1} \frac{e^{-\lambda(1+\varepsilon)a_j}}{1 - \lambda a_j} \quad \text{if } 0 < \lambda < \min \{1/a_j\} \\ &\cong \prod_{j=1}^{\mu+1} e^{-\varepsilon \lambda a_j + \frac{2}{3}(\lambda a_j)^2} \quad \text{if } 0 < \lambda < \frac{1}{3} \min \{1/a_j\} \end{aligned}$$

and on taking $\lambda = \frac{\varepsilon}{3hn}$

$$\begin{aligned} &\cong \prod_{j=1}^{\mu+1} e^{-\frac{\varepsilon^2 a_j}{3hn} \left(1 - \frac{2}{9} \frac{a_j}{hn}\right)} \\ (23) \quad &\cong e^{-\frac{7\varepsilon^2}{27} \frac{T}{hn}} \quad \text{as } a_j \cong hn. \end{aligned}$$

Similarly, for any $\lambda > 0$,

$$(24) \quad P(S \cong (1-\varepsilon)T) = P(e^{-\lambda S + \lambda(1-\varepsilon)T} \cong 1) \cong e^{-\frac{\varepsilon^2 T}{2hn}}$$

on taking $\lambda = \frac{\varepsilon}{hn}$.

We may argue as above with each $a_j = 1$ (or otherwise) to obtain

$$(25) \quad P(|Y_{\mu+1} - (\mu+1)| \cong \varepsilon(\mu+1)) \cong e^{-\frac{\varepsilon^2}{4} \mu}.$$

The result follows from (23), (24) and (25) after replacing ε by $\varepsilon/2$ throughout the proof. ■

We can now readily establish (16). Let $T = \sum_{i=1}^{hn-1} t_i$, and let

$$A_n = \{|L(H_n) - (h/r)\zeta(3)| \cong \varepsilon\},$$

$$B_n = \{|T/(\mu+1) - (h/r)\zeta(3)| \cong \varepsilon/2\}.$$

Then

$$P(A_n) \cong P(B_n) + P(A_n | \bar{B}_n).$$

Now Lemma 6 gives

$$P(B_n) \leq P\left(|T - (h/r)(\mu + 1)\zeta(3)| \leq (\varepsilon hr/4) \binom{n}{2}\right) \leq \exp(-\varepsilon^2 n^{1/4}/65rh).$$

Furthermore,

$$P(A_n | \bar{B}_n) \leq P(|L(H_n) - T/(\mu + 1)| \leq \varepsilon/2 | \bar{B}_n) \leq \exp(-\tilde{\varepsilon}^2 \tilde{T}/16hn) \text{ by Lemma 7,}$$

where

$$\tilde{\varepsilon} = (\varepsilon/2)/((h/r)\zeta(3) + \varepsilon/2)$$

and

$$\tilde{T} = ((h/r)\zeta(3) - \varepsilon/2)(\mu + 1).$$

The inequality (16) now follows.

5. General case

We will now use the inequality (16) to complete the proof of lemma 0 and thus of Theorem 2 in the general case. We shall assume that $D_e > 0$ for each edge e in $E(H)$. Any edges e with $D_e = 0$ would cause only minor irritation.

We will first use the approach of Steele [5] to relate a random edge-length X_e with distribution function F_e to one which is uniform in $[0, D_e^{-1}]$. Let A_e denote the set of atoms of F_e and define Y_e by

$$Y_e = \begin{cases} D_e^{-1}F_e(X_e) & X_e \notin A_e \\ D_e^{-1}(F_e(X_{e-}) + U_e(F_e(X_e) - F_e(X_{e-}))) & X_e \in A_e. \end{cases}$$

where U_e is a uniform $[0, 1]$ random variable (and we make a suitable assumption of independence).

Observe that Y_e is uniform on $[0, D_e^{-1}]$ and $X_e > X_{e'}$ implies $Y_e \geq Y_{e'}$. It follows that there is always a tree T which is simultaneously of minimum length for edge-lengths $\{X_e\}$ and $\{Y_e\}$.

Our hypotheses for the $F_e, e \in E(H)$ show that we may write $F_e(x) = D_e x + xg_e(x)$ and $F_e(x-) = D_e x + xh_e(x)$ where g_e and h_e go to zero as $x \rightarrow 0$. We then have

$$(27) \quad \sum_{e \in T} D_e^{-1} X_e h_e(X_e) \leq \sum_{e \in T} Y_e - \sum_{e \in T} X_e \leq \sum_{e \in T} D_e^{-1} X_e g_e(X_e).$$

Our immediate task is to bound the probability that either of the outside terms of (27) is significant. Let $g_e^*(x) = \sup \{g_e(y) : 0 \leq y \leq x\}$ for $e \in E(H)$. Now fix $\varepsilon > 0$. For $e \in E(H)$ let

$$\lambda_e = \lambda_e(\varepsilon) = \sup \{\lambda : g_e^*(\lambda) \leq \varepsilon D_e\}.$$

Let

$$\mu = \min \{\lambda_e : e \in E(H)\}$$

and

$$v = \min \{P(X_e < \mu) : e \in E(H)\},$$

and note that $\mu > 0, v > 0$.

Then

$$P\left(\sum_{e \in T} D_e^{-1} X_e g_e(X_e) > \varepsilon \sum_{e \in T} X_e\right) \cong P(X_e \cong \mu \text{ for some } e \in E(H)) \cong \\ \cong P(H_{n,v} \text{ is not connected}).$$

But this last quantity is at most $e^{-nv/3}$ (for n sufficiently large) by an argument similar to that of Lemma 4. An analogous argument yields

$$P\left(\sum_{e \in T} D_e^{-1} X_e h_e(X_e) < -\varepsilon \sum_{e \in T} X_e\right) \cong e^{-nv/3}$$

for some $v' = v'(\varepsilon) > 0$.

Thus if $L(H'_n)$ denotes the length of a minimum spanning tree when the length X'_e of edge $e \in E(H)$ is uniform in $[0, D_e^{-1}]$ then we can write, for small fixed $\varepsilon > 0$.

(28a)
$$P(L(H_n) \cong (1 + \varepsilon)^2 (h/D) \zeta(3)) \cong \\ \cong e^{-nv/3} + P(L(H'_n) \cong (1 + \varepsilon)(h/D) \zeta(3))$$

and

(28b)
$$P(L(H_n) \cong (1 - \varepsilon)^2 (h/D) \zeta(3)) \cong \\ \cong e^{-nv/3} + P(L(H'_n) \cong (1 - \varepsilon)(h/D) \zeta(3)).$$

These results reduce the general case of the theorem to the case of uniform edge-lengths. Thus in particular the inequality (16) holds also when all edge lengths have the negative exponential distribution with mean 1.

However, the above argument works also in the other direction; and we have

(29a)
$$P(L(H'_n) \cong ((1 + \varepsilon)/(1 - \varepsilon))(h/D) \zeta(3)) \cong \\ \cong e^{-nv/5} + P(L(H_n) \cong (1 + \varepsilon)(h/D) \zeta(3))$$

and

(29b)
$$P(L(H'_n) < ((1 - \varepsilon)/(1 + \varepsilon))(h/D) \zeta(3)) \cong \\ \cong e^{-nv/3} + P(L(H_n) < (1 - \varepsilon)(h/D) \zeta(3)).$$

Thus the case of uniform edge-lengths reduces to the case of (negative) exponential edge-lengths.

Now we are almost home. We wish to show that lemma 0 holds when the edge-lengths have exponential distributions.

Let us check first that we may take each D_e rational. Let D' be rational, $0 < D' < D$. We shall show that there exist rational D'_e , $0 < D'_e < D_e$ for $e \in E(H)$ such that $\sum_{v \in e} D'_e = D'$ for $v \in V(H)$. A similar approximation from above may be obtained by the reader.

Suppose then that $0 < \varepsilon < 1$ and $D' = (1 - \varepsilon)D$ is rational. Write $D' = M/N$ where M and N are positive integers such that both $\varepsilon ND_e \cong 1$ and $(1 - \varepsilon)ND_e \cong 1$ for each $e \in E(H)$. Observe next that the polyhedron

$$\sum_{v \in e} x_e = (1 - \varepsilon)D \\ 1/N \cong x_e \cong [(1 - \varepsilon)ND_e]/N$$

is non-empty, since it contains the point $x_e = (1 - \varepsilon)D_e$, $e \in E(H)$. But the polyhedron is rational, and so it contains a rational point, as required.

Finally then we wish to show that lemma 0 holds when each edge e of H has exponential distribution with rational parameter $\lambda_e = D_e = P_e/Q$. Consider the graph \hat{H} obtained from H by replacing each edge e by P_e parallel copies, each with edge-length exponentially distributed with parameter $1/Q$ (mean Q). Then $L(H_k)$ and $L(\hat{H}_n)$ have the same distribution, and we have already shown the required result for $L(\hat{H}_n)$.

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